

Beyond Uncertainty

Reasoning with Unknown Possibilities

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Contents

1	Introduction	3
1.1	Roadmaps to the unknown	3
1.2	Internal consistency and its limits	5
1.2.1	Introducing probabilities	5
1.2.2	Rationality as internal consistency	7
1.2.3	Beyond internal consistency	10
1.3	Limited awareness in perspective	11
1.3.1	Towards a truly <i>subjective</i> expected utility theory . . .	11
1.3.2	A limit to what can be modelled?	13
1.4	How we will proceed	15
2	Fully subjective decision models and the test of time	19
2.1	Introduction	19
2.2	Situating oneself in the stream of events	20
2.3	Feedback from the world	24
2.3.1	The two extremes	25
2.3.2	The middle ground	27
2.4	Concluding remarks on chapter 2	29
3	Modelling (Un)Awareness	31
3.1	Introduction	31
3.2	Rent or buy?: Types of awareness change	31
3.3	Modelling awareness growth	34
3.3.1	First pass: the catch-all model	34
3.3.2	Our preferred model	37
3.4	Awareness growth in the literature	41
3.5	Concluding remarks on chapter 3	44

4	Responding to Awareness Growth	45
4.1	Introduction	45
4.2	Traditional Bayesianism	46
4.2.1	The rule of conditionalisation	47
4.2.2	Interpreting the norm	48
4.3	Reverse Bayesianism	50
4.4	Counterexamples to Reverse Bayesianism	55
4.5	Concluding remarks on chapter 4	59
5	Awareness Rigidity	61
5.1	Introduction	61
5.2	Awareness Rigidity defined	62
5.3	Against Awareness Rigidity	64
5.4	<i>Restricted</i> Reverse Bayesianism	67
5.5	Concluding remarks on chapter 5	71
6	Anticipating Awareness Growth	73
6.1	Introduction	73
6.2	Example: Solar Radiation Management	75
6.3	Modelling anticipated awareness growth	77
6.4	Not such extraordinary reasoning?	82
6.5	Concluding remarks on chapter 6	86
7	Awareness Reflection	87
7.1	Introduction	87
7.2	Informal argument	88
7.3	Formal argument	91
7.3.1	Awareness Reflection formalised	91
7.3.2	A Dutch book argument for Awareness Reflection	94
7.4	Awareness Reflection vs. Reverse Bayesianism	96
7.5	Preference Awareness Reflection	99
7.6	Concluding remarks on chapter 7	102
8	Conclusion	105
8.1	‘Whereof one cannot speak, thereof one must be silent’	105
8.2	Norms for limited awareness	107
8.3	Two challenges revisited	109

CONTENTS

v

8.4	Connection to applied work	111
8.5	Further research	112
8.6	Closing remarks	114
	References	114

List of Figures

2.1	A roadmap as per table 2.1 with predicted learning	21
5.1	Expansion to include STR	61
5.2	Refinement into high/low language	65

List of Tables

1.1	A simple roadmap for navigating the unknown, a.k.a. a decision model.	4
1.2	A more detailed roadmap for navigating the unknown	17
1.3	A yet more detailed roadmap for navigating the unknown . .	17
2.1	Conservationist’s roadmap	20
2.2	feedback on our reasoning roadmaps	25
2.3	Conservationist’s roadmap after awareness of pest contingencies	26
3.1	Rent or buy?	32
3.2	Expanded rent-or-buy	33
3.3	Expanded and refined rent-or-buy	33
3.4	Less aware state with “catch-all” proposition(s)	36
4.1	State space for rent-or-buy	52
4.2	Expanded state space for rent-or-buy	52
4.3	Expanded and refined state space for rent-or-buy	52
6.1	Decision with no anticipated awareness growth	78
6.2	Decision with anticipated awareness growth	79
7.1	Beach or Home	88

Preface

When we reason about what to do we try to include everything that we think might affect the outcome of our decision. When deciding between taking the bus and the subway, for instance, one may take into account things such as the traffic, the reliability of the bus and the subway, how (un)comfortable the two options are, and so on. But often we fail to include something in our deliberation, even though it could affect the outcome of our decision in an important way. Sometimes we fail to include such things due to a momentary lack of perspective—a failure to consider some otherwise familiar contingency that bears on the decision at hand. Other times we fail in this way due to a more far-reaching lack of perspective. For instance, when early industrialists reasoned about their actions, they failed to take into account the possibility that their actions would lead to a hotter climate; this was well beyond their scientific comprehension of the world at the time. While the latter may be more striking and interesting, we regard both kinds of failures as ones of *limited awareness*.

The main aim of this book is to introduce the topic of limited awareness, and changes in awareness, to those interested in the philosophy of decision-making and uncertain reasoning. While it has long been of interest to economists and computer scientists, this topic has only recently been subject to philosophical investigation. Indeed, at first sight limited awareness seems to evade any systematic treatment: it is *beyond the uncertainty* that can be managed. On the one hand, an agent has no control over what contingencies she is and is not aware of at a given time, and any awareness growth takes her by surprise. On the other hand, agents apparently learn to identify the situations in which they are more and less likely to experience limited awareness and subsequent awareness growth. How can these two sides be reconciled? That is the puzzle we confront in this book.

We propose a way of conceiving limited awareness that does justice to its elusive character. While we build on earlier work of others, our analysis departs from this previous work in various ways. We accept that awareness growth can have radical and unpredictable effects on an agent's beliefs. But we argue that this does not preclude *anticipating* awareness growth. Moreover, we argue that unlike the effects of experiencing "unexpected" awareness growth, the effects of anticipating awareness growth are both quite predictable and can be captured without too radical a departure from the standard (Bayesian) model of rational preference and belief.

Acknowledgements

to be completed...

1

Introduction

1.1 Roadmaps to the unknown

This book is about our plight as reasoning agents in the world. That is, our plight as agents who seek to understand the world and how we can change it to best align with our ends. This requires some ingenuity because our perspective on the world is inherently limited. Think of it this way: our experience is confined to a more or less tiny patch of the world's history, so we can be certain of relatively little. The best we can do is try to account for all the contingencies, that is, all the ways the world *might* be, in at least as much detail as is relevant for our purposes. According to standard decision theory, at any rate, this is the kind of roadmap that we assemble for ourselves to navigate the unknown.

Consider, for instance, the reasoning of a single-minded conservationist who cares only about eradicating weeds and pests. At a particular juncture, our conservationist perceives she has a limited set of options: she can release a moth that will hopefully eat the non-native cactus plant known as "prickly pear", or she can continue with the status quo, whereby all resources are devoted to manually uprooting the pear. Our conservationist perceives that which of the two options will best realise her ends depends on whether or not the world is such that the moth will eat (and kill) the prickly pear, and this she is unsure about. Thus her enumeration of the relevant contingencies is as per the second and third columns in table 1.1. In decision-theory parlance, these are the *states of the world*. We see that, depending on the state of the world, the options yield different outcomes,

and indeed, for this choice problem, which option is best depends on what state of the world is the actual one.

	moth eats pear	moth fails to eat pear
Release moth	pear eradicated	pear thriving , wasted resources
Status quo	pear thriving	pear thriving

Table 1.1: A simple roadmap for navigating the unknown, a.k.a. a decision model.

The prickly pear decision is a highly stylised one, but it exemplifies the general predicament we reasoning agents face, day in and day out. We are condemned to live as gamblers. By our own lights, our choices are nearly always risky ventures—we are not assured that the world will turn out one way or another, and thus whether our ends will be served more or less well by any given choice of option. Not only are we limited by our practical circumstances—the options we have to change the world—but we are limited also by our epistemic circumstances—the ability we have to discern what is true of the world and thus which of our options serve us best.

It is these trying epistemic circumstances—our unavoidably parochial view of the world—that is the particular concern of this book. We have little to say about an agent’s basic values or ends; we simply take them as given. (That is, we leave the analysis of values or ends for others to address.) Nonetheless these ends have an important bearing on our epistemological project. We will nearly always represent an agent’s epistemic circumstances or perspective in the context of a choice problem, as per table 1.1. That is because, as suggested above, we are ultimately interested in how an agent reasons about what to do to further her ends. Our conservationist, for instance, may well have many (at least implicit) ideas about the way the world is, concerning the weather, her family and friends, and so on. But these ideas are in a sense idle, at least in the context of her current options and ends, which concern the eradication of pests. Generally speaking, there may or may not be a richer story to tell about an agent’s epistemic life. This book, however, aims only to capture a part of this story. When we talk of an agent’s epistemic perspective, we mean her current views about the possible contingencies, or ways the world might be, **in so far as those contingencies** play a role in her reasoning about what to do *now* to further

her ends. In other words, an agent's epistemic perspective is relative to a decision-problem, on this picture.

1.2 Internal consistency and its limits

The standard decision-theoretic account of our reasoning does not quite end with simple (qualitative) roadmaps. Our roadmaps not only account for the possible contingencies or ways the world might be, but also, typically, their relative plausibility. To be *rational*, i.e., to reason well, is for one's judgments of relative plausibility to be (at least) *internally consistent*. Indeed, decision theory can be understood as a theory of internal consistency. It tells how our epistemic and evaluative judgments or attitudes must "hang together" so as to yield clear choices of action that are not self-defeating with respect to our ends.

This book is about the limits of internal consistency, in particular due to an agent's (*limited*) *awareness*, or what she perceives to be the possible contingencies or ways the world might be. But we need an understanding of the guidance that internal consistency can provide in order to see what are the shortcomings of this guidance. In what follows, we start by articulating the guidance (1.2.1), before looking more closely at how arguments from internal consistency work (1.2.2) and what are their inherent limitations (1.2.3).

1.2.1 Introducing probabilities

We said that agents consider the relative plausibility of the possible ways the world might be. Put differently, agents have varying *degrees of confidence*—also known as *degrees of belief*, or as *credences*, which is the term we shall mostly use—in ways the world might be. It is as if we weigh the competing possibilities on a set of scales with multiple arms. The common wisdom is that, as an arm gets more weight, the others should collectively get less weight. To be more precise: credences are rational only if they can be represented as probabilities. This norm is often referred to as *probabilism*. For instance, if our conservationist assigns much weight, or has relatively high credence, say, of 0.9, in the moth eating the pear, then on pain of inconsistency she must assign little weight, or have relatively low credence,

here 0.1, in the moth *not* eating the pear.

Let us more thoroughly describe our conservationist's credences, as pertinent to the choice problem depicted in table 1.1. As noted, what matters for determining how well her options realise her ends is whether the released moth will eat the prickly pear or not—that is, which of these states of the world is actual—which we can denote M and $\neg M$ respectively. Strictly speaking, our conservationist is also unsure about what she will do, whether she will release the moth or not, denoted R and $\neg R$ respectively. This yields four relevant possibilities for how the world might be: $R \& M$, $R \& \neg M$, $\neg R \& M$ and $\neg R \& \neg M$. We assume that our conservationist has credences in each of these fine-grained possibilities or *sure outcomes* that are each non-negative and together sum to one. Her credences in all other propositions can be derived in conformity with the probability calculus. Moreover, presumably our conservationist's credences in M versus $\neg M$ do not depend on her credences in R versus $\neg R$. That is, $P(M|R) = P(M|\neg R)$, where P represents the agent's credences, and $P(M|R)$ denotes her *conditional* credence in M given R . That is, in this case we have *act-state probabilistic independence*, but this need not always be so.¹

Table 1.1 is the most economical depiction of our conservationist's choice problem. But note that the view of the world she brings to bear on this choice problem, and her associated credences, may be somewhat more complicated. Perhaps she entertains other potential properties of the world in an effort to form judgments about the relevant states of the world. For instance, perhaps our conservationist recognises that there may or may not be a drought during the year following the release of the moth, denoted D and $\neg D$ respectively. She does not care about droughts. Our assumption is that she cares only about the eradication of weeds and pests. So in a sense whether or not there is a drought does not matter to her. Nonetheless, the consideration of whether there will be a drought may assist our conservationist in forming her credences in M and $\neg M$. After all, by the law of total probability, $P(M) = P(M \& D) + P(M \& \neg D)$. Plausibly, our conservationist arrives at a settled credence in M by considering her "component" credences in $P(M \& D)$ and $P(M \& \neg D)$. This is to say that our conservationist's roadmap may look more like table 1.2 (which, along with table 1.3, is produced at the

¹The knowledgeable reader may discern that our presentation of the agent's decision model follows that of Jeffrey (1965), as opposed to Savage (1954).

very end of this introduction).

In general, there is a *space of propositions describing ways the world could be* about which the agent has an opinion that bears on her practical reasoning at some given time. This space of propositions about which she has an opinion is assumed to have a certain completeness in structure. In technical terms, it forms an algebra \mathcal{F} with the following characteristics (which means that it is what is called a *Boolean algebra*):

- \mathcal{F} contains a contradictory proposition (\perp).
- \mathcal{F} contains a tautologous proposition (\top).
- \mathcal{F} is closed under disjunction, conjunction, and negation. That is, if A and B are in \mathcal{F} , then $A \vee B$, $A \& B$ and $\neg A$ and $\neg B$ are also in \mathcal{F} .

The rational agent has credences in the propositions in \mathcal{F} that can be represented by a probability function P . That is, $P(A) \in [0, 1]$ for all A in \mathcal{F} ; $P(\perp) = 0$; $P(\top) = 1$; $P(A \vee B) = P(A) + P(B)$ for all mutually exclusive A and B in \mathcal{F} .

1.2.2 Rationality as internal consistency

Why think that rational credences are probabilities? There are various arguments for this position. A relatively straightforward one is known as the “Dutch book argument”.² It turns on the claim that an agent’s credences are effectively her “betting odds” or the proportion of the stakes she’d be willing to pay for a bet that yields the stakes if the proposition in question turns out true but yields nothing otherwise. It is shown that if and only if her betting odds over the space of propositions conform to the probability calculus, the agent is *not* vulnerable to accepting a set of bets that would guarantee her a sure loss (measured in monetary terms). Positioning oneself for a sure loss is considered a marker of inconsistency, albeit of a *pragmatic* kind. So one’s credences had better be probabilities. Note that other arguments for credences being probabilities turn on inconsistencies of a *non-pragmatic* kind. For instance, credences that do not conform to the probability calculus are shown to be *accuracy dominated* in the sense that

²The first suggestion of the Dutch book argument is due to Ramsey (1926).

some alternative probabilistic credence function would be more accurate (roughly, closer to the truth) no matter how the world turns out.³

We have been emphasising the role of rational credences in deliberating about what to do. Indeed, the standard wisdom is that probabilistic credences, together with a cardinal value or *utility* function over sure outcomes (in our example: the cell entries in tables 1.1 and 1.2) determine the *expected utility* of risky options, which is the basis for their relative desirability and thus choice-worthiness. The expected utility of an option (or indeed any prospect or claim about the world represented by a proposition) is the sum of the probability of each possible way in which the option or prospect may be true multiplied by the utility of that way it may be true. The higher the expected utility, the better, according to expected utility theory.

To make the above more concrete, recall the choice problem of our conservationist, as represented by the original table 1.1. The conservationist is considering two options: release the moth (R), and not release the moth ($\neg R$). There are only two states of the world that she considers relevant to the outcome of her options: the moth eats the pear (M) or it does not ($\neg M$). Now let U be the conservationist's utility function over sure outcomes. Then the expected utility (EU) of the conservationist's options, according to her, are given by:

$$EU(R) = U(R\&M)P(M | R) + U(R\&\neg M)P(\neg M | R)$$

$$EU(\neg R) = U(\neg R\&M)P(M | \neg R) + U(\neg R\&\neg M)P(\neg M | \neg R)$$

By the aforementioned assumption that act-state probabilistic independence holds in this case, the above equations reduce to:

$$EU(R) = U(R\&M)P(M) + U(R\&\neg M)P(\neg M)$$

$$EU(\neg R) = U(\neg R\&M)P(M) + U(\neg R\&\neg M)P(\neg M)$$

There are again various arguments for why one ought to evaluate and rank risky options according to their expected utility.⁴ (Call this the *expected utility principle*.) One kind of argument appeals to the infinite long run: expected utility matches what one would be more or less sure to gain were the choice repeated over and over. The more prominent kind of

³See, e.g., Joyce (1998).

⁴For overviews of these arguments, see Briggs (2017) and Steele and Stefánsson (2015).

argument is known as the *expected utility representation theorem*, which is all-encompassing in that it supposedly justifies rational credences being probabilities together with the expected utility principle, all in one hit. There are several different versions of this theorem, but they have a similar form. They appeal to consistency in the ranking of options comprising a specially engineered rich set of options. (Note that the “ranking” of options here means how they are ordered in terms of [the agent’s judgement](#) of their relative desirability or “choice-worthiness”.) One consistency requirement, for instance, is *transitivity*, which requires that if an agent ranks option *A* over *B* and *B* over *C*, then she ranks *A* over *C*. In short, the expected utility theorem is the result that if and only if an agent’s ranking of the relevant options satisfies a number of consistency constraints like transitivity, she can be represented as having credences measured by a probability function and judgments of relative desirability measured by a cardinal function that conforms with the expected-utility principle.⁵

What we have just presented is the orthodox position in decision theory. Indeed, the arguments for rational credences being probabilities and for rational evaluations of options satisfying the expected utility principle are core results in decision theory. This is not to say, however, that these arguments have not been challenged. Different (generally weaker) constraints on what counts as rational credence have been fruitfully explored, where “rational” is still understood in terms of internal consistency.⁶ And different (again, generally weaker) constraints on rational evaluations of options have also been fruitfully explored, where, again, “rational” is understood as internal consistency.⁷ We do not pursue these debates in this book, however. Rather, we stick with the orthodoxy, at least to the extent that it is applicable. But this should not be interpreted as strong endorsement of expected utility theory. We build on the orthodox theory for reasons of simplicity. The concerns raised in this book concern the limits of internal consistency, and as such, are orthogonal to the debate between expected utility theory and alternative formalisations of internal consistency.

⁵For some classical representation theorems, see Ramsey (1926), Savage (1954), and Bolker (1967).

⁶In particular, that rational credences need not be *precise* is a popular view. For an overview, see Bradley (2019).

⁷Lara Buchak (2013) has developed an influential theory with weaker constraints on the evaluation of options than those in orthodox decision theory. Stefánsson and Bradley (2019) criticise Buchak’s theory and defend an alternative view.

1.2.3 Beyond internal consistency

So internal consistency provides standards for our roadmaps. But one might have further concerns about whether an exemplary roadmap, in this respect, will really be a good guide to the world. What if the roadmap is not very faithful to the world, despite being internally consistent? There are two issues that might come to mind. The first is that the agent's credences may not be very sensible. After all, we would not take very seriously an agent who is extremely confident that the moon is made of green cheese, even if the agent's credence function overall conformed to the probability calculus.

The other issue is that the agent may have *limited awareness* of the various contingencies, or possible ways the world might be, such that her perception of her own decision problem, and what matters, is rather naive. For instance, perhaps our conservationist simply does not take account of whether or not there is a drought, and so her credences lack the sophistication that would come from accounting for this contingency (as per table 1.2 rather than table 1.1). Worse still perhaps, she might not recognise contingencies that have a more obvious bearing on her evaluation of outcomes. For instance, our conservationist might not realise that the released moth could possibly eat native plant species, thereby itself becoming a pest. Table 1.3 includes this contingency in addition to the drought contingency that was already introduced in table 1.2.⁸

This book is concerned with the problem of limited awareness. We do not try to address the problem of nonsensical credences. For what it's worth, we do not think the two issues are entirely unrelated, since the more possible ways in which the world might be that an agent is aware of, the more checks and balances there are on her credences. For instance, if one is to maintain high credence in the moon being made of green cheese, without there being a conflict within one's overall credal state, then one needs to also maintain some other odd credences, like that telescopes are generally misleading, at least when it comes to the composition of the moon. And so on. There may be no substantial norms governing the credences that an

⁸Table 1.3 is what we later call a *refinement* of the agent's possibility space described in table 1.1. Alternatively, our conservationist might have implicitly assumed that the moth would not be a pest, and failed to realise that this is not exhaustive of the ways the world could be (in which case the awareness growth would be what we later refer to as an *expansion*).

agent has at any particular point in time, beyond the probability calculus. But we leave that as an open question.

1.3 Limited awareness in perspective

So our focus is on how an agent's roadmap to the unknown may fail to capture all the important contingencies or ways the world might be. Standard decision theory is silent on how to respond to this failure. The primary reason for this is that standard decision theory does not acknowledge the failure! It is simply assumed that the space of possibilities is public knowledge, or else is an objective part of the model that illuminates an agent's subjective attitudes. But surely a general account of reasoning should not **assume that all share the same view of the ways the world might be.** To see that the move to modelling limited awareness is an important one, let us situate it in a series of developments towards a more thoroughly subjective decision theory. We are then in a better position to determine whether this is an important next move.

Katie: I think we may have to clarify what we mean by "public" knowledge, etc., i.e., to emphasise that we simply mean what is and what is not specifically part of the decision-problem rather than the agent's mind

1.3.1 Towards a truly *subjective* expected utility theory

Let us return to the development of the standard model. The history of standard expected utility theory can be understood in terms of social scientists' efforts to provide operational definitions of key reasoning attitudes like credence and desire in more and more realistic settings. We want to position our inclusion of growing awareness as the natural next step in the trajectory towards a fully general model of reasoning. To see roughly how this goes, we will describe an earlier, somewhat analogous move along the trajectory: the move from the von Neumann-Morgenstern (1947) model to that of, say, Savage (1954) or Jeffrey (1965).

Start with the expected utility theory of von Neumann and Morgenstern (vNM). The theory establishes the conditions under which an agent ranks risky options in accordance with their expected utility. But the model vNM introduce effectively has only one free variable or subjective dimension: the extent the agent desires or values various outcomes. Other aspects of the options are fixed since these are assumed to be public knowledge that is not specific to the agent in question. In particular, the nature of the final

outcomes is assumed to be transparent to all, as well as the probabilities by which the various options may yield these outcomes (and, by implication, the options themselves). The theory establishes that if and only if the agent's ranking of these public options satisfies some proposed internal consistency constraints, then the agent's strength of desire can be measured by a cardinal utility function (unique up to positive linear transformation) such that she evaluates options according to their expected utility. (Recall our earlier remarks about the expected utility theorems.)

The vNM model offers great insights about the structure of reasoning; in particular, it provides a powerful way to measure and thus conceptualise strength of desire. The problem is that the model is only applicable when its assumptions hold. And they rarely do hold. In most choice settings, the probabilities with which the available options yield the different outcomes is not public knowledge. Hence the development of expected utility theories that allow for more subjectivity in the characterisation of the options. The theories of Savage and Jeffrey, for instance, accommodate desires that are specific to the agent, as vNM do, but in addition allow for probabilistic credences that are specific to the agent (*subjective* desire *and* credence). The ingenuity of these theories is that they still allow a way of measuring and thus conceptualising these attitudes in terms of the agent's ranking of options (insofar as this ranking is internally consistent or rational). But note that there remains an important "objectivity" assumption: the nature of the options is still largely a matter of public knowledge. In Savage's model, for instance, the options are functions from states of the world to outcomes. It is not that the probabilities for states of the world are public knowledge. That is where the agent's own credences come in. But the states of the world themselves and the nature of the outcomes are assumed to be transparent to all involved.

The subjective expected utility theories of Savage and Jeffrey are more generally applicable than that of vNM. They propose a way of measuring and thus making sense of subjective desire and credence. But these theories too have limited applicability. In particular, they are not applicable when the very states of the world and outcomes (or more generally, the ways the world might be) cannot be assumed to be public knowledge but are rather specific to the agent's own perspective, in particular, her specific level of awareness. That is the kind of scenario we will focus on in this book.

1.3.2 A limit to what can be modelled?

Now one might think there is nevertheless good reason for not trying to model limited awareness. To begin with, this arguably makes for one too many free variables in our model of an agent's reasoning that cannot, even in principle, be empirically settled. We lose sight of what our concepts mean in an operational sense, and providing this meaning was one of the great advances of decision theory. The ingenuity of Savage's expected utility theory, for example, is that it offers a way to understand credence and desire in terms of the evidence of choice behaviour. But there might seem to be no conceivable choice scenario that could reveal an agent's attitude to an outcome of which she is unaware without thereby making her aware of the outcome.

Secondly, one might think it is in any case pointless to model limited awareness in a normative decision model. This is not something that an agent can correct. Compare this to, say, non-transitive preferences. An agent can check for herself whether her preferences are transitive or not; and if they are not, she can make them transitive. By contrast, an agent cannot check whether she is in fact unaware of something; and even if she suspects that she is unaware of something, she cannot simply correct for this lack of awareness by becoming more aware.

These are two important challenges for the project of modelling limited awareness. We will not, however, respond to either of them directly in a satisfying way. We rather acknowledge, here at the outset, reasonable scepticism about the project we embark upon. Our hope is that the scepticism is mitigated by our treatment of limited awareness in the remainder of the book.

Moreover, we are not alone in the quest to understand limited awareness and how it affects an agent's reasoning. A small but strong cohort of economists, computer scientists, and philosophers have already made important progress towards this goal, and we are indebted to the foundations they have laid. The very notion of limited awareness—or *unawareness*, as they typically call it⁹—is due to early work by economists and computer

⁹Since awareness is typically a matter of degree, we find the term "limited awareness" to be more apt, in most cases, than "unawareness". And indeed that is the terminology we shall typically use. However, since "unawareness" has come to be widely used, in particular by economists, for what we think should be called "limited" awareness, we will occasionally use the term "unawareness" for limited awareness, for instance, when discussing the works

scientists. An extensive review of this work can be found in Schipper (2015). Philosophers too have explored the challenge that growing awareness poses for the traditional probabilist model of belief and decision, typically under the guise of “the problem of new theories” (Earman 1992; the problem was originally raised by Glymour 1980 as the counterpart to “the problem of old evidence”). Richard Bradley (2017) has recently turned philosophical attention to the general problem that (un)awareness poses for rationality; his own work draws on a series of recent papers by the economists Karni and Vierø (2013, 2015, 2017).

We will expand on these earlier contributions in relevant places throughout the book. For now, let us simply give a taste of how the challenges raised above have been at least partially met by others. For instance, on the first: Piermont (2017) illustrates how we can, by observing a person’s choices between what he calls “contingent plans” (that is, conditional options of the form: if state s obtains, then choose c), tell whether or not a person anticipates her awareness to grow. In short, the idea is that a person anticipates awareness growth just in case she is willing to take on some cost to postpone a choice between contingent plans, even when all such plans that she can conceive of are available. Moreover, Karni and Vierø (2017) show that even if a person anticipates awareness growth, as long as her preferences satisfy certain consistency constraints, then she can be represented as maximizing expected utility. The representation even allows us to infer how (un)desirable the decision-maker predicts a currently unknown outcome to be.

Turning to the second challenge: Although it is true that one cannot determine whether or not one is unaware of something, nor immediately become aware of that something in case one is unaware, one can take steps to increase one’s level of awareness if one suspects that there is something one is unaware of. That is, one can conduct (formal or informal) experiments that can be expected to reveal contingencies that one is unaware of, if there are such contingencies. Moreover, one can make plans for how one will adjust one’s attitudes, and what choices one will make, if one does become more aware. In fact, it would seem that one *should* in certain choice situations suspect that there is something that one is unaware of; hence, one arguably should, in some situations, make plans for how to respond to

of these economists.

growing awareness.

The issues just raised—concerning how one should respond to awareness growth and the extent to which one can and should plan for such growth—will be explored in a systematic fashion throughout the book (more on how we will proceed shortly). For the moment, the idea is just to get a feel for how (un)awareness, as amorphous as it may sound, nonetheless admits of structured treatment according to usual decision theoretic principles. That said, we are sympathetic to the general worry that there is only so much that can be said or done about limited awareness. It is just that, notwithstanding the significant and pioneering contributions to date, not enough has yet been said. Limited awareness and its perils is an interesting aspect of our epistemic predicament, and it has not yet been addressed in a fully general and comprehensive way. This book is intended as a step towards remedying that situation.

1.4 How we will proceed

As noted, we will approach the problem of (un)awareness in an incremental fashion. Even if it turns out that there is little to say about one's limited awareness at a time, there is plausibly a lot to say about *changes in awareness*, specifically *awareness growth*, over time. After all, our roadmaps for navigating the unknown are not finished deeds but rather works in progress. They are subject to feedback from the world itself, as events unfold and the agent comes to new realisations. That is, in any case, the starting insight for the approach taken in this book. We will approach the issue of limited awareness—what it is and whether there is anything to be done about it—by considering first *changes* in awareness.

In fact we begin in chapter 2 at an even more preliminary point. We consider, in qualitative terms, the kinds of feedback that the world may provide on our reasoning roadmaps. As said, these roadmaps are not finished deeds. Consider our conservationist. Subsequent to her deliberations, let's say she releases the moth. The story does not end there of course. Presumably, after some time, it will be apparent to her whether or not the prickly pear has been eradicated, and thus which state of the world is actual. A more thorough roadmap than those we have considered thus far would in fact seek to anticipate such learning events. That is, in addition to other

sorts of properties of the world, such as whether a moth population will eradicate a prickly pear population, it is important to also anticipate one's interactions with the world, in particular, what one will learn and when, and what options one may choose between and when. The popular format for these more thorough roadmaps is the *sequential-decision* model. Chapter 2 proposes a way to read such models. But no matter how thorough, a roadmap that attempts to account for all relevant future contingencies comes up against the world as time unfolds. Actual events will either be consistent with the roadmap or not.

In chapter 3 we elaborate on the kind of feedback from the world that is the focus of this book—the realisation of unfamiliar contingencies, or in other words, the realisation of limited awareness. We articulate the different types of awareness growth that such a realisation may herald, and go on to consider how such growth is best modelled. We finally reflect on how our approach to awareness growth borrows insights from the work of others across different disciplines.

Chapters 4 and 5 go on to consider the impact of awareness growth on rational credences. While others who have investigated this question have a rather sanguine view of the impact of awareness growth on credences, we argue that awareness growth may have highly disruptive, far-reaching impacts, at least in some cases. We proceed to offer a characterisation of the better behaved cases: when awareness growth has a more conservative impact on one's credences.

We then turn, in chapters 6 and 7, to the question of whether there is anything to be done in advance to stave off radical changes in one's credences due to awareness growth. We argue that there is a sense in which one can and indeed *should* plan ahead for awareness growth, even if, at the end of the day, there are no assurances that awareness will not change in unforeseen ways. Encouraging decision-makers to plan for awareness growth, as well as providing them with the tools for such planning, is arguably decision theory's most important contribution to the problem of limited awareness.

Chapter 8 summarises the findings of the book and suggests avenues for future research.

1.4. HOW WE WILL PROCEED

Release moth	moth eats pear; drought	moth eats pear; no drought	moth fails to eat pear; drought	moth fails to eat pear; no drought
Status quo	pear eradicated pear thriving	pear eradicated pear thriving	pear thriving, wasted resources pear thriving	pear thriving, wasted resources pear thriving

Table 1.2: A more detailed roadmap for navigating the unknown

Release moth	moth eats pear; drought; pest	moth eats pear; drought; no pest	moth eats pear; no drought; pest	moth eats pear; no drought; no pest	...
Status quo	pear eradicated, pest pear thriving	pear eradicated, no pest pear thriving	pear eradicated, pest pear thriving	pear eradicated, no pest pear thriving	...

Table 1.3: A yet more detailed roadmap for navigating the unknown

2

Fully subjective decision models and the test of time

2.1 Introduction

As we explained in the introductory chapter, one can view the introduction of limited awareness into decision models as the natural next stage in decision theory's historical trajectory, from the "objective" expected utility theory of von Neumann and Morgenstern (1947) to the "subjective" expected utility theories of Savage (1954) and Jeffrey (1965). In this chapter we begin the task of articulating what a fully subjective decision model that can accommodate changes in awareness looks like.

The starting point for our investigation is the experience of failure. In the course of time, it often becomes apparent that one's roadmap or decision model has failed to account for all contingencies. The recognition of failure, we suggest, is the first step to i) acknowledging limited awareness and subsequent growth in awareness and ii) reflecting on when such growth might happen again. These are the major themes to be developed later in this book.

There are in fact several kinds of feedback that the world may provide on one's roadmap. The failure to account for all contingencies is just one kind of feedback. This chapter will consider the other kinds of feedback as well, by way of putting limited awareness in perspective. First, however, we need to introduce a more detailed kind of roadmap, known as a *sequential-decision model*. Such a model makes explicit an important kind of future contingency:

20 2. FULLY SUBJECTIVE DECISION MODELS AND THE TEST OF TIME

the agent’s own interactions with the world—not only when and what she will be able to *choose* but also when and what she will *learn* about the world. An important aspect of the feedback that the world provides is whether an agent learns the things about the world that she expects to learn.

This chapter thus proceeds, in section 2.2, to lay out sequential-decision models. We can then identify, in section 2.3, four major kinds of feedback the world may provide on such models. Once we have limited awareness in perspective, we can elaborate on the particular forms it may take and how it is best modelled. But that task must wait until the next chapter.

2.2 Situating oneself in the stream of events

Return to the plight of our single-minded conservationist. Assume, for starters, that the contingencies she is tracking (apart from her own choice of action) concern whether or not the moth eats the pear and whether or not there is a drought. That is, her roadmap, or personal decision model, is as per the second decision problem in the introductory chapter, reproduced below in abbreviated form.

	eats; drought	eats; no drought	no eats; drought	no eats; no drought
Release moth	eradicated	eradicated	thriving, waste	thriving, waste
Status quo	thriving	thriving	thriving	thriving

Table 2.1: Conservationist’s roadmap

Let us now add some further detail to our conservationist’s roadmap, concerning her other interactions with the world. As noted above, such interactions may include i) when she receives information from the world, which we may refer to as “learning events”, and ii) when she makes a choice that impacts on the world, which we may refer to as “choice events”.¹ Besides her choice of whether or not to release the moth, let us assume that our conservationist predicts only one relevant interaction with the world: she predicts that she will learn, prior to making her choice about whether to

¹In game theory, the terms “Nature’s moves” versus “agents’ moves” are used. This is along the lines of the distinction we are drawing here, but does not quite coincide, since “Nature’s moves” may or may not be learnt by the agent.

release the moth, whether or not there will indeed be a forthcoming drought. To keep things simple, imagine that she predicts she will somehow learn this with certainty, even though that is rather implausible—forthcoming weather is not ordinarily the sort of thing one can learn with certainty.

Our conservationist's sequential-decision model is shown in figure 2.1. As per convention, the circle nodes represent learning events, when the world presents new evidence, the possibilities for which are represented by the branches emanating from the node; the square nodes represent choice events, when the agent will have the opportunity to choose amongst options, normally also represented as branches emanating from the node. Here we abbreviate the model slightly by inserting the relevant decision table at the choice nodes. (Recall that M denotes that the moth eats the pear, while D denotes that there is a drought.)

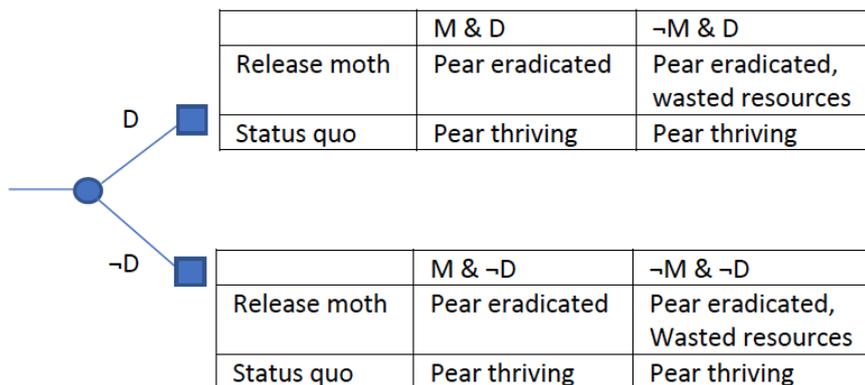


Figure 2.1: A roadmap as per table 2.1 with predicted learning

As per their *static* (or *single-decision*) counterparts, sequential-decision models are not typically interpreted in a fully subjective way. The space of contingencies, including the agent's own interactions with the world, are taken to be public knowledge. For instance, with respect to figure 2.1, it is typically taken to be a matter of public knowledge that the agent will learn whether or not there will be a drought, and will subsequently choose whether or not to release the moth. But we pursue a fully subjective interpretation of sequential-decision models. It is not merely that the agent

22.2. FULLY SUBJECTIVE DECISION MODELS AND THE TEST OF TIME

assigns her own subjective probabilities to whether or not there will be a drought, but it is moreover her own fallible prediction that she will learn this information at this time and then will subsequently be presented with a choice as to whether or not to release the moth. Of course, the finer details of “subjectivising” the space of contingencies in this way will need to be ironed out. For now we ask that the reader simply go along with our reading of sequential-decision models such as that depicted in figure 2.1, and our “subjectivising” project more generally.

We make one further point before continuing: If learning and choice events are a matter of the agent’s own predictions, then strictly speaking, these should feature in the space of contingencies. That is, the relevant propositions should feature in the algebra \mathcal{F} over which the agent has opinions. For instance, as well as entertaining the contingencies of drought, D , and no drought, $\neg D$, figure 2.1 shows that our agent also entertains learning, prior to acting, of drought, or else of no drought, which we might denote L_D and $L_{\neg D}$ respectively. Learning is typically understood to mean that the proposition in question is indeed true, in which case L_D entails D and $L_{\neg D}$ entails $\neg D$.² The agent may also make predictions about her credences at future times, given how she plans to update or change her beliefs upon learning new information. But including all this in the agent’s algebra would make it rather complicated. So the convention is not to include these propositions, unless for some special purpose (such as arises in chapter 7). One can think of a sequential-decision model as a relatively simple way to represent all these extra predictions about how the agent will interact with the world, without expanding the agent’s algebra (instead using circle and square nodes in a tree-like structure).

We assume that our conservationist, described now by figure 2.1, predicts that she will continue to be her own rational self throughout time. She does not take her future self to be a stranger whose credences and desires or ends are completely disconnected, for no good reason, from her current attitudes. She rather predicts, for instance, that any changes in credence will accord with her rational plans. In other words, our conservationist predicts

²The structure of sequential-decision models imposes a further assumption that need not generally be the case: that learning some fact at a given time is inevitable, such that $P(L_D|D) = 1$ and $P(L_{\neg D}|\neg D) = 1$, where, as usual, P is the agent’s credence function. So often an event and the learning of that event at a particular time are treated as logically equivalent.

she will be *stable* over time. Indeed, the assumption of agent stability over time will run throughout the book. The task of reasoning is difficult enough for agents who take themselves to be continuous with their future selves.

It is generally accepted that the rational plan for updating or changing one's credences in response to learning that some proposition (in which one had positive credence) is certain, as befalls our conservationist in figure 2.1, is to *conditionalise*, as it were, on what one has learnt. The new credence in any proposition is just the old credence, conditional on what is learnt.³ Accordingly, our conservationist predicts that her credence in the moth eating the pear, M , in the case that she learns of a drought at the circle node, L_D (or equivalently, D), will be $P(M|L_D)$ (or equivalently, $P(M|D)$). Likewise, she predicts that her credence in M , in the case that she learns that a drought will not occur, $L_{\neg D}$ (or equivalently, $\neg D$), will be $P(M|L_{\neg D})$ (or equivalently $P(M|\neg D)$). Note that for the remainder of this chapter we will suppress the propositions describing the agent's own learning, since we are assuming that the agent treats these propositions (for instance, L_D) as equivalent to those describing what is learnt (in this case, D).

It follows from the above that our conservationist's current credence in M equals a weighted average of what she predicts her credence in M to be in the future, where the weighing is determined by her current credence in D vs. $\neg D$. (In other words, the agent satisfies a principle called *Reflection*, which we discuss in detail in chapter 7.) To see this, note that by the law of total probability, for any M and D :

$$P(M) = P(M | D)P(D) + P(M | \neg D)P(\neg D)$$

Therefore, by our assumption that, first, our conservationist predicts that she will learn either D or $\neg D$, and, second, she predicts that her credence in M , in the case that she learns of D , will be equivalent to $P(M|D)$, and similarly for $\neg D$, it follows that her current credence in M equals what she *expects* (i.e., her weighted prediction) her credence in M to be in the future. To take a concrete numerical example, let's say that our conservationist's

³ The rule of conditionalisation is sometimes referred to as *Bayesian learning*. Indeed, the term *Bayesianism* incorporates the norm that credences should be probabilities (probabilism) as well as the norm that credences should be revised in accordance with conditionalisation (or generalisations thereof). Moreover, subjective expected utility theory, as described in chapter 1, is sometimes referred to as *Bayesian decision theory*. In later chapters there will be reason to use the *Bayesian* terminology.

24.2. FULLY SUBJECTIVE DECISION MODELS AND THE TEST OF TIME

credences are such that $P(M|D) = 0.6$ and $P(M|\neg D) = 0.9$. Then, since $P(D) = 0.5 = P(\neg D)$, it follows from the above that $P(M) = 0.75$, and this, moreover, is her expected future credence in M .

Now it may well be that the learning of the drought is irrelevant for our conservationist's choice, because whether her credence in $P(M)$ increases from 0.75 to 0.9, or else decreases to 0.6, her preferred option is still to release the moth. (This will be so if eradicating the pear is sufficiently better than the status quo relative to how much worse the wasted resources are compared to the status quo.) Or it may be that she predicts the learning to be relevant in the sense that it will affect her choice. Either way, that is not our focus here. The point is rather to show how an agent's roadmap may be rather sophisticated in keeping track of the possible ways the world may be, including when information or choice opportunities will arrive. These ways in which the agent interacts with the world are typically represented in sequential-decision format. Strictly speaking, they should be included in the algebra of propositions of which the agent is aware.

Finally, we note that the sequential-decision format has brought to light some controversies amongst decision theorists concerning how a rational agent should identify and evaluate options.⁴ But we can put these controversies aside for the time being, at least, since we are thus far concerned with orthodox agents who maximise expected utility, plan to learn in accordance with the rule of conditionalisation, and moreover expect their plans to be carried through. Under these assumptions, the various approaches to choosing in the sequential-decision context coincide. If and when debates about *sequential choice* become relevant, we will expound on the details.

2.3 Feedback from the world

As time progresses, the world provides feedback on an agent's roadmap or (sequential-)decision model. Think of feedback as itself a learning event that was either accounted for or not. The learning may concern a contingency with which one was either familiar or unfamiliar. This gives rise to four major kinds of feedback from the world, of which Table 2.2 provides a summary. The more obvious are the two extreme kinds of feedback. The happy case (top left) is where the world effectively vindicates one's model.

There was some confusion in the Oxford group about whether unfamiliar was just another way of describing awareness or not. I am not if we need to add anything to address that.

⁴For a summary of the controversies, see Steele (2018).

All learning was accounted for and concerns familiar events. The less happy case (bottom right) involves learning that was not accounted for and moreover concerns completely unfamiliar events. This is the paradigmatic case of limited awareness and subsequent awareness growth. Finally, there is the middle ground (top right and bottom left).

	familiar	unfamiliar
accounted for	planned learning	planned awareness growth
unaccounted for	unplanned learning	unplanned awareness growth

Table 2.2: feedback on our reasoning roadmaps

We will expand on table 2.2 in what follows. The point to note from the outset is that awareness growth amounts to the recognition of unfamiliar contingencies. While this kind of learning about the world is often unaccounted for, table 2.2 floats the possibility that this need not be so. We can plan for awareness growth, so to speak. (And as such there may be a case for representing such plans in a sequential-decision model by way of some further symbol, say a diamond.) This issue will be taken up later in the book.

2.3.1 The two extremes

Sometimes the world affirms one's roadmap or model, in the sense that one expects to receive some kind of input from the world, that is, to learn something, and does indeed receive this input. We referred to this as the happy case. Consider our conservationist as described in figure 2.1. It may well be that she learns about the drought—whether it is surely happening or surely not happening (and nothing more)—prior to making her choice, just as she predicted. So she learns as planned; her learning was accounted for and familiar.

Of course, part of the reason that she does not perceive other events that her roadmap did not account for may be that she is not looking for these events. Her observations of the world are no doubt "theory-laden" or rather "roadmap-laden". We are not suggesting that, when an agent perceives that the world affirms her roadmap, it has passed some objective test and is shown to be faithful to reality. It is purely the agent's own perspective that we are considering here. What we draw attention to is just

26 2. FULLY SUBJECTIVE DECISION MODELS AND THE TEST OF TIME

that sometimes an agent’s roadmap appears to serve her well. The feedback from the world does *not* call for a restructuring of her roadmap, but rather affirms it. That need not always be the case, as we will explore now.

What does the less happy case look like? Assume that our conservationist reasons as before, as described by figure 2.1. But the feedback she receives from the world is rather different. She learns about the drought, as before, but then a little later, she realises there are further contingencies that are pertinent to her choice problem. Some interaction with the world, perhaps simply the sight of a moth landing on a native plant, prompts the realisation that the moth may or may not itself become a pest, eating and destroying native plants.

So in this latter case our conservationist’s roadmap turns out not to serve her very well. She has a learning experience that was not accounted for, namely, the realisation sparked by the moth landing on the leaf of a native plant. Moreover, this realisation or learning event itself concerns potential properties of the world that are unfamiliar, in the sense of not being already articulated by her roadmap. In short, in this case our conservationist’s learning was not accounted for and is unfamiliar. It prompts a restructuring of her roadmap, presumably such that her impending decision problem now has the complexity of the third decision problem of the introductory chapter, reproduced below in abbreviated form.

	eats; drought; pest	eats; drought; no pest	eats; no drought; pest	...
Release moth	eradicated, pest	eradicated, no pest	eradicated, pest	...
Status quo	thriving	thriving	thriving	...

Table 2.3: Conservationist’s roadmap after awareness of pest contingencies

We suggest that this sort of roadmap failure and subsequent restructuring is the natural way to think about limited awareness and subsequent awareness growth. That is, it is natural to think that awareness growth is unplanned (although it need not always be, as we shall soon see). Note that, in the case we described, the newly realised potential properties of the world are in a sense unfamiliar because they are newly realised, but they are not *radically* unfamiliar. It is not as if our conservationist lacks the concept of either moths or pest species. After all, we stipulated that she cares single-mindedly about eradicating weeds and pests and is contemplating whether

to release a moth. Moreover, it is presumably not a giant conceptual leap to imagine moths being pest species, in addition to prickly pears being weeds. Our conservationist simply had not considered, in the context of the choice problem at hand, whether the moth might itself become a pest. Perhaps for very mundane reasons, this simply did not cross her mind, and so was not part of her reasoning, or roadmap, at the time in question.

Later in chapter 4 we will appeal to examples of limited awareness and subsequent growth that do in fact involve radically unfamiliar properties of the world. The examples concern shifts in scientific world view, for instance, the shift in the early 1900s to accommodate Einstein's theories concerning the relationship between space and time, amongst other things. These more dramatic examples are useful for highlighting that, at the earlier time, the contingencies in question are truly beyond the agent's grasp of the ways the world might be. We can all appreciate that someone in the early 1900s simply does not have access to the notion that space may be curved, or that someone in the 1970s does not have access to the notion of data sharing over the internet. These are very obvious limitations in awareness. What we draw attention to is that the more ordinary limitations in awareness, like that experienced by our conservationist, may be similarly unavoidable. For whatever reason, sometimes quite ordinary properties of the world are simply inaccessible to an agent at a given time.

2.3.2 The middle ground

We suggested above that there are yet two further kinds of feedback that the world may present to an agent, as regards her reasoning roadmap (refer back to table 2.2). Our contention is that the accounted for/unaccounted for distinction comes apart from the familiar/unfamiliar distinction. Awareness growth has to do with the latter distinction: it is when the agent is exposed to unfamiliar as opposed to familiar properties of the world. Such learning may nonetheless be in some sense planned. Before we get to this case, however, let us warm up with the other middle-ground case: when the agent learns something familiar, but this was nonetheless *not* accounted for. This case should strike the reader as very common, even if it is not explicitly acknowledged or discussed in the literature.⁵

⁵Learning is often discussed in the literature in a way that abstracts from whether or not the learning experience was predicted or planned. Sequential-decision models *do* include

28 2. FULLY SUBJECTIVE DECISION MODELS AND THE TEST OF TIME

To illustrate unplanned learning of the familiar, let us again appeal to the plight of our conservationist. Assume this time that she does not expect to learn anything relevant before making her choice. Her roadmap is simply table 1.2, or figure 2.1 *without* the circle node representing the learning experience regarding the drought. Now, it may turn out that she does not learn anything; the world affirms her roadmap. Another possibility, however—the one we draw attention to now—is that she unexpectedly learns about the drought, that is, she either learns D or $\neg D$. These are familiar contingencies in that they are represented in her roadmap. But the learning experience was not accounted for, so in this sense, her roadmap let her down. Presumably, it is easy for her to adapt, however, to her unplanned circumstances. For starters, she can simply proceed to update her credences in the rational way, by adopting her “old” credences conditional on what she now knows to be true, whether D or $\neg D$. So, the idea is that when it comes to unexpected learning of the familiar, the agent does have credences conditional on what she learns; it is just that she had not anticipated that she would revise her beliefs in line with these conditional credences in the roadmap she had been using for the decision at hand.

We turn now to the case of planned awareness change: when an agent predicts or plans that she will come to recognise unfamiliar contingencies. We suggest that such occurrences can be represented in a suitably embellished sequential-decision model, even if the contingencies in question cannot be articulated in advance. For instance, our conservationist might anticipate that she will (or at least may) learn something pertinent to her choice problem, in addition to whether or not there is a drought. But she cannot put her finger on what this might be. Her characterisation of the contingency may be more or less abstract. We will not say too much more about this sort of case for now, as the focus of chapters 6 and 7 is what an agent can anticipate and how it affects her current credences. The answers to these questions are not obvious. The “hard” position is that limited awareness cannot be well understood by the agent and it is incoherent to regard her as anticipating awareness growth. As mentioned, we will argue for the “soft” position that awareness growth can in some sense be anticipated and

learning experiences, but these models are often treated as objective representations of a temporally-extended choice problem, rather than a subjective representation that may or may not turn out to be correct.

planned for. But the details matter.

2.4 Concluding remarks on chapter 2

So far we have introduced (un)awareness and changes in awareness in general terms, with reference to the experience of unfamiliar contingencies. We have moreover situated changes in awareness within a general picture of how one's (sequential-)decision model, or roadmap, may be vindicated or else undermined by the feedback one receives from the world. We have not, however, yet said much about how to formally represent limited awareness or how to formally model changes in awareness. That is the main topic of the next chapter.

30 2. *FULLY SUBJECTIVE DECISION MODELS AND THE TEST OF TIME*

3

Modelling (Un)Awareness

3.1 Introduction

We turn now to a closer examination of the forms that limited awareness and changes in awareness might take. We initially, in section 3.2, introduce a new example which we use to illustrate two intuitively quite different ways in which awareness might grow. Subsequently, in section 3.3, we again use this example to describe two approaches to modelling limited awareness and growing awareness; first a model that may seem natural but which we nevertheless reject, next our own preferred model. Section 3.4 reflects on how our discussion has been informed by others; we briefly survey the diverse treatments of awareness growth in the literature.

3.2 Rent or buy?: Types of awareness change

Suppose that you are contemplating buying an apartment and moving out of your rental apartment. The reason is that you have heard that the rent might go up, and you are primarily concerned with whether you will be able to make ends meet. So, you are trying to figure out whether you will be more likely to make ends meet in your current rental apartment or in an apartment that you own. Your decision problem can be represented by table 3.1.

Now you realise that an additional important possibility, that you should factor into your decision, is that the owner of the apartment you currently rent decides to sell; in which case you will find yourself suddenly homeless.

	Rent higher	Rent same or lower
Rent	Rent & Rent higher	Rent & Rent same or lower
Buy	Buy & Rent higher	Buy & Rent same or lower

Table 3.1: Rent or buy?

This transforms your decision problem¹ into one that can be represented by table 3.2.

Before you make your choice, you hear speculations about the central bank planning to raise interest rates. As you realise that this would affect whether you are able to make ends meet after having bought an apartment, you want to factor this possibility into your deliberation too. To simplify the table, let us assume that the interest rate only affects you if you decide to buy. Then the decision-problem you are now faced with can be represented by table 3.3.

Note that in the shift from the epistemic state represented by Table 3.1 to the one represented by Table 3.2, you have *expanded* or *extended* the possibilities you entertain to include the possibility that the apartment is sold. By contrast, in the shift from the epistemic state represented by Table 3.2 to the one represented by Table 3.3, you have *refined* some of the possibilities you entertain to accommodate the possibility of a changed interest rate. In this case, your old possibilities are effectively split into more fine-grained ones, allowing for new partitions of the possibility space.

The examples above are special cases of awareness growth: what we might dub *pure* expansion and *pure* refinement respectively. Throughout the book we will often appeal to these sorts of cases. But it is also possible for awareness growth to be a *mixture* of expansion and refinement. For example, one may recognise that the apartment might be sold at the same time as recognising that the interest rate might change (a transition from table 3.1 directly to table 3.3).

¹In keeping with our subjectivist approach, we take an agent's *decision problem* to be (at least) partly defined by the agent's own epistemic state, in particular, what she takes to be her options and the relevant states of the world.

	Rent higher	Rent same or lower	Rental apartment sold
Rent	Rent & Rent higher	Rent & Rent same or lower	Rent & Rental apartment sold
	Buy & Rent higher	Buy & Rent same or lower	Buy & Rental apartment sold

Table 3.2: Expanded rent-or-buy

	Rent higher	Rent same or lower	Rental apartment sold
Rent	Rent & Rent higher	Rent & Rent same or lower	Rent & Rental apartment sold
	Buy & Rent higher Interest higher	Buy & Rent same or lower Interest higher	Buy & Rental apartment sold Interest higher

Table 3.3: Expanded and refined rent-or-buy

3.3 Modelling awareness growth

The tables above give us a fair idea of what limited awareness and subsequent awareness growth looks like. But in order to get a firm grip on whether an evolving agent like the one described by the transition from table 3.1 through to table 3.3 is rational, both *at* a time and *over* time, we need to go into a little more detail. Ultimately, we need to describe and assess the agent's evolving credences.

In the introductory chapter it was noted that a reasoning agent may be depicted as entertaining a set of propositions at the time in question. The set of propositions should form what is known as a *Boolean algebra*, that is, a set of propositions \mathcal{F} , that contains both the contradiction and the tautology, and is moreover closed under disjunction, negation and conjunction. Moreover, it is assumed that a rational agent has credences in the propositions in \mathcal{F} that can be represented by a probability function P (see p. 7 for the details.)

We want to preserve this way of conceiving a rational agent's epistemic state at any given time, but also introduce the possibility of awareness growth. To that end, let us refer to an agent's state of awareness at a given time as her *awareness context*. For any given awareness context, then, the agent entertains a set of propositions—a Boolean algebra—over which she has probabilistic credences. The question is: how does the agent's set of propositions change upon awareness growth? More precisely: what is the relationship between her Boolean algebras, so to speak, from one awareness context to the next? We respond to this question in our own way in what follows, and then, in section 3.4, we compare our model of awareness growth with those others have proposed.

3.3.1 First pass: the catch-all model

In the philosophical tradition, propositions are typically interpreted as sets of *possible worlds*, where these worlds are understood to be maximally detailed world histories or ways the world might be. For instance, the proposition "Oswald killed Kennedy" is just the set of possible worlds for which that particular proposition is true.² The tautologous proposition is thus

²If propositions are identified with sets of possible worlds, the reader might wonder why we characterise the agent's algebra for any given awareness context sententially rather than

identified with the set of all possible worlds, while the contradictory proposition is identified with the empty set of possible worlds. Note that, for most purposes, it is not useful to quantify over *all* possible worlds; rather, one can simply refer to *the possible worlds relative to \mathcal{F}* , which are specific enough just to assign truth values to each of the propositions in \mathcal{F} and thus amount to a coarsening of the set of all possible worlds.

This way of conceiving an agent's proposition space, however, does not seem to leave room for awareness growth by expansion. On this model, at any given time, the rational agent supposedly grasps the full set of possible ways the world might be, since she assigns probability one to the tautology. That does not seem to square with transitions, e.g., the shift from table 3.1 to table 3.2 above, in which the agent apparently comes to recognise possibilities that are inconsistent with all those she previously entertained, such that there is a genuine enlargement of her (subjective) set of possibilities.

The way to proceed, if one wants to stick with this kind of underlying model, is to engineer the agent's proposition space for any given awareness context so that all instances of awareness growth can be treated, at least formally, as refinements. The thought would be that an agent's algebra includes an abstract *catch-all* proposition signifying "Other ways the world might be" or "None of the above"; it represents all those further possible ways the world might be that the agent cannot (yet) articulate. For instance, table 3.4 is identical to table 3.1 above except that it allows for a catch-all proposition that may be interpreted as "Other ways in which the landlord affects my living arrangement".³

The idea is that what we earlier called awareness growth by expansion is really a special case of refinement—it is a refinement of the catch-all. In the case that the agent realises the landlord might sell the apartment (what was formerly the transition from table 3.1 to table 3.2), the catch-all proposition is effectively divided into "Rental apartment sold" and, say, "Other (yet unarticulated) ways in which the landlord affects my living arrangement."

set-theoretically. The reason is that we want to leave open the possibility that propositions are not sets of possible worlds (see e.g. footnote 6).

³The catch-all will typically be best interpreted as a disjunction, since it must account for *all* other ways that the world might be that are inconsistent with those of which one is already aware. For instance, even for the simple proposition space depicted in table 3.4, the catch-all should strictly speaking be "Other ways in which the landlord affects my living arrangement *or* other housing options that I might choose".

	Rent higher	Rent same or lower	???
Rent	Rent & Rent higher	Rent & Rent same or lower	Rent & ???
Buy	Buy & Rent higher	Buy & Rent same or lower	Buy & ???

Table 3.4: Less aware state with “catch-all” proposition(s)

The worry with the catch-all model is that it seems not to be an apt characterisation of an agent’s reasoning. Nothing that we have said thus far suggests that an agent at all times entertains a catch-all by way of accounting for her inevitably limited awareness. Indeed, this proposal may not even be cogent, since, in order for an agent to make sense of a catch-all, she would presumably need to entertain some universal set of possibilities relative to which the catch-all can be defined as the complement of those possibilities she can properly articulate. But it is hard to see how the agent could have access to this universal set of possibilities (which might in fact not even be a coherent notion), given that, by assumption, some of these possibilities cannot be articulated. So, it is hard to see how the catch-all could be well-defined for the agent.⁴

By way of response, one might resist these cogency worries. Indeed, far from being incoherent, an agent who accounts for her limited awareness in entertaining a catch-all may be regarded praiseworthy. Alternatively, one might argue that the cogency worries are moot because a catch-all model need not entail that the agent herself entertains such a proposition. The inclusion of a catch-all proposition, to which the agent implicitly assigns zero probability, may simply be the most elegant way for the modeller to account for an agent’s limited awareness. While we are sympathetic to these lines of argument, we do not find them sufficiently convincing (to be elaborated shortly) to continue pursuing a catch-all model of awareness growth.

As for the first: We admit that a certain portrayal of the cogency worry would, as it were, prove too much. The portrayal we have in mind is one that trades on the catch-all being too “abstract”, in that the agent has no idea how to specify the proposition’s content. After all, the abstractness of a

⁴We thank Alan Hájek for suggesting this way of putting the problem. He elaborates on worries along these (and other) lines in his unpublished manuscript “Omega”.

proposition would seem to be a matter of degree rather than an on/off affair. And one can surely represent an agent as having credences in propositions of varying abstractness without thereby being committed to her being able to precisely articulate what these propositions mean. Moreover, arguably the wiser agents do routinely entertain propositions at the more extreme levels of abstractness—propositions that are intended to capture a wide range of contingencies. Indeed, this point will become important later, in chapters 6 and 7. That said, the catch-all does not simply capture a wide range of contingencies; it represents *all* other ways the world might be that are inconsistent with those of which the agent is aware. But how can an agent conceptualise *all other ways* the world might be? Therein lies the real cogency problem.

Turning now to the second line of argument: The idea is that the catch-all is accessible only to the modeller (or perhaps to the agent herself at a later time), not the agent in question (at that time), who is modelled as implicitly assigning the catch-all zero probability.⁵ Refer back to table 3.4: the catch-all proposition indicates that the agent has limited awareness. She entertains only two ways in which her landlord may affect her living arrangements in ways relevant to her ends. She fails to see the other possibilities, and thus these are represented by a catch-all. The agent implicitly assigns the catch-all zero probability. The model in a sense captures a wiser person's perspective and how that person interprets the reasoning of the agent in question. But again, therein lies the problem. For some applications, there may well be special reason to capture a wiser perspective on an agent's limited awareness. But we contend that that is not the case for our application. We are interested in the reasoning of a single agent, and how her perspective or awareness changes with time. For this purpose, there is no need to keep track of how the agent's awareness looks from some more expansive point of view. Doing so only detracts from the simplicity of the model.

3.3.2 Our preferred model

The way forward, we suggest, is to divorce the *agent's possibilities* from objective possible worlds. While it is useful to depict an agent's epistemic

⁵We thank Richard Bradley for this suggestion.

outlook in terms of atomic possibilities that are the ultimate bearers of probability, these atomic possibilities need not be the objective possible worlds that many think give meaning to propositions. Indeed, a model of an agent's epistemic outlook need not offer an interpretation of propositions. They can simply go uninterpreted; the agent's possibilities being defined as truth functions over these uninterpreted propositions, or more accurately, over the *basic* propositions. We proceed now to spell out this idea more carefully.

We say that an agent's awareness context is defined by a set X of *basic* propositions of which she is aware (which we assume to be finite). We take basic propositions to be primitive propositions, representing simple facts about the world, that do not involve any logical connectives. So, for instance, in the awareness context described by Table 3.1, "Rent" and "Rent higher" are basic, while "Rent & Rent higher" and " \neg Rent higher" are not. The basic propositions are not themselves given an interpretation in our model; they are simply the primitive facts that the agent is aware of. (In other words, any deeper interpretation of these propositions, whether in terms of objective possible worlds or some other kind of structure, is not explicitly modelled here. We do not here take a stance on whether propositions should be identified with sets; see footnote 6.)

Let the *possibilities* that the agent is aware of be truth functions, ω_i , that return "true/false" for each of the basic propositions. Note that below we will occasionally use $\omega_1, \omega_2, \dots, \omega_n$ to denote individual possibilities. The *putative* set of possibilities are all the distinct truth functions that take this form, that is, effectively all the different combinations of truth values for the basic propositions. This is merely the *putative* or *first-pass* set of possibilities, since some will be deemed inconsistent by the agent (to be explained shortly) and thus excluded from the *real* set of possibilities (as recognised by the agent). We may describe the possibilities in terms of conjunctions of the basic propositions for which the ω_i function in question returns "true". So, in the awareness context represented by Table 3.1, the possibility $\{\omega_i(\text{Rent}) = \text{true}, \omega_i(\text{Buy}) = \text{false}, \omega_i(\text{Rent higher}) = \text{true}, \omega_i(\text{Rent same or lower}) = \text{false}\}$ can be described as "Rent & Rent higher". From now on, we will use this latter way of describing possibilities.

For the set of basic propositions X , let W_X be the agent's (real) set of possibilities, which is a subset of the *putative* set of possibilities, containing

only the possibilities that the agent regards as consistent. A possibility is *consistent*, by the agent's lights, if all its conjuncts *could* be true, that is, if the agent does not take the conjuncts to be mutually inconsistent. What an agent takes to be the set of consistent possibilities will depend on what she regards as partitions of the proposition space (corresponding to properties or categories for which one and only one value can be assumed). For instance, for the agent described by Table 3.1, one partition of the space is {"Rent", "Buy"}, these being the candidate values for what we might call the "action property"; a necessary condition for being a consistent possibility, then, is that the conjuncts include only one of "Rent", "Buy".

So an agent's awareness context \mathbf{X} may be just as well defined in terms of her possibility space, $\mathbf{W}_\mathbf{X}$. Any given basic proposition X_i can now be associated with a set of possibilities in $\mathbf{W}_\mathbf{X}$: the $\omega_i \in \mathbf{W}_\mathbf{X}$ for which the proposition X_i is true. For simplicity, we refer to this set as $\{X_i\}$.^{6,7} We can now also generate a Boolean algebra, $\mathcal{F}_\mathbf{X}$, in the usual way: $\neg X_i$ is associated with the set $\mathbf{W}_\mathbf{X} \setminus \{X_i\}$, $X_i \vee X_j$ is associated with the set $\{X_i\} \cup \{X_j\}$, and $X_i \& X_j$ is associated with the set $\{X_i\} \cap \{X_j\}$. For reasons that will become apparent in the next chapter, the same proposition can be associated with different sets of possibilities in different awareness contexts. So, more formally, we can think of a proposition as a function from the awareness contexts in which the proposition plays a role to the corresponding sets of possibilities.⁸

For simplicity, we will model only *growth* in awareness over time; our model will not countenance the *shrinking* or *contraction* of awareness over time. Inclusion of the latter possibility would complicate the model and its presentation; moreover, there is a tradition in modelling *rational* belief change to consider only incremental learning (gains in information) rather than forgetting (losses of information).⁹ That said, contraction of one's

⁶We are not here suggesting that the basic propositions are identical to, or defined in terms of, the relevant set of possibilities. After all, the possibilities were themselves constructed from propositions that had some prior meaning. One can retain the traditional notion of propositions being identified with sets of objective possible worlds, as per, e.g. Stalnaker (1984), although this is not explicitly represented in our model. The relation of "association" that we appeal to here is intended to be weaker than "identity".

⁷Strictly speaking, the set in question should be thought of as being indexed to the relevant awareness context. If we wanted to make the index explicit, we could, for instance, write $\{X_i\}_\mathbf{X}$. But to simplify the notation, we omit making the index explicit.

⁸To clarify: For awareness contexts where the proposition does not play a role, it is not associated with any set of possibilities.

⁹For notable exceptions, see Titelbaum (2012) and Bradley (2017).

concepts—that is, contraction of the set \mathbf{X} —may in some cases not be due to “forgetting” but rather due to considerations that make it an important aspect of rational learning; we leave further exploration of this phenomenon for future work, however.

Now let us address the dynamics of awareness. We say that the agent’s *awareness grows* when the awareness context shifts from \mathbf{X} to $\mathbf{X}^+ = \mathbf{X} \cup \mathbf{X}_j$. Note that by the assumptions we made above, when the awareness context shifts from \mathbf{X} to \mathbf{X}^+ there is a corresponding shift from $\mathbf{W}_\mathbf{X}$ to $\mathbf{W}_{\mathbf{X}^+}$ and from $\mathcal{F}_\mathbf{X}$ to $\mathcal{F}_{\mathbf{X}^+}$. Strictly speaking, $\mathbf{W}_\mathbf{X}$ and $\mathbf{W}_{\mathbf{X}^+}$ do not have any possibilities in common; after all, the possibilities in each are truth functions that have a different number of propositions in their domain. If, however, we allow that the possibilities may be described in terms of the proposition that they are each associated with—the conjunction of all basic propositions for which the function in question returns “true”—then $\mathbf{W}_\mathbf{X}$ and $\mathbf{W}_{\mathbf{X}^+}$ may in certain cases (as we will see shortly) have possibilities in common.

Now we can characterise the difference between awareness growth by expansion and awareness growth by refinement. Let us measure the *length* of a possibility by the number of propositions for which the function in question returns “true”. (Recall that we assume that the set of basic propositions is finite.) we say that the awareness growth was (purely) due to *expansion* if the *number* of possibilities in $\mathbf{W}_{\mathbf{X}^+}$ is greater than in $\mathbf{W}_\mathbf{X}$, without any possibilities becoming *longer* in the sense given. In contrast, we say that the awareness growth was (purely) due to *refinement* if the *number* of possibilities in $\mathbf{W}_{\mathbf{X}^+}$ is greater than in $\mathbf{W}_\mathbf{X}$, and moreover, all possibilities in $\mathbf{W}_{\mathbf{X}^+}$ are *longer* (in the sense just described) than the possibilities in $\mathbf{W}_\mathbf{X}$. Moreover, in the case of pure expansion there are some possibilities common to $\mathbf{W}_\mathbf{X}$ and $\mathbf{W}_{\mathbf{X}^+}$, while in the case of pure refinement, there are no possibilities common to $\mathbf{W}_\mathbf{X}$ and $\mathbf{W}_{\mathbf{X}^+}$.

Return again to our rent-or-buy example, and suppose now that in the least-aware context (Table 3.1), the only possibilities that the agent of interest is aware of and considers consistent can be characterised as: “Rent & Rent same or lower”, “Rent & Rent higher”, “Buy & Rent same or lower”, “Buy & Rent higher”. In other words, she regards any possibility that involves “Rent & Buy”, and likewise “Rent same or lower & Rent higher”, inconsistent. Now, when awareness grows due to an expansion, e.g., when the agent becomes aware of the possibility that the owner sells the apartment—which

the agent takes to be inconsistent with the owner keeping the rent the same and also inconsistent with the owner increasing the rent—the possibilities do not become longer. Instead, we simply add “Rent & Rental apartment sold”, “Buy & Rental apartment sold” to the original four possibilities. This is represented by the shift from the awareness context represented by Table 3.1 to the one represented by Table 3.2.

In contrast, when awareness grows due to refinement of the possible interest rate—as represented by the shift from the awareness context represented by Table 3.2 to the one represented by Table 3.3—some of the new possibilities are longer; for instance, “Buy & Rent same or lower & Interest higher” compared to “Buy & Rent same or lower”. The number of possibilities also grows, since e.g. “Buy & Rent same or lower” becomes “Buy & Rent same or lower & Interest higher”, “Buy & Rent same or lower & Interest same or lower”.

3.4 Awareness growth in the literature

We noted in Chapter 1 that there is already a small but significant literature on (un)awareness and awareness growth, dating back 30 years or so. A select group of economists and computer scientists have introduced this phenomenon to their various abstract models of reasoning and knowledge (and indeed, as noted earlier, the term *(un)awareness* comes from this literature).

For the most part, at least until very recently, philosophers have tended to consider awareness growth in relation to very specific questions in the philosophy of science concerning theory change. In the last few years, however, occasional philosophers have turned their attention to awareness growth more generally. Richard Bradley’s (2017) recent work for instance pursues the more general challenge that (un)awareness poses for decision theory. Anna Mahtani’s (2020) is primarily a critical response to Bradley, as is Chloé de Canson’s (ms). Aron Vallinder’s (2018) examines the problems that changing awareness poses for moral epistemology. Joe Rousoss (2020) discusses the relationship between expert testimony and awareness growth. And Edward Elliott (ms) discusses the problems that lack of awareness creates for possible world models of belief and develops a multi-agent model based on neighbourhood structures.

Our treatment of (un)awareness draws on various aspects of earlier approaches. Like Bradley (2017) and economists such as Karni and Vierø (2013, 2015), our aim is to characterise a rational agent's practical reasoning in the face of limited awareness and subsequent awareness growth.¹⁰ As is typical of philosophers, we prefer the generality of a Jeffrey (1965) inspired decision framework, rather than the Anscombe and Aumann (1963) framework (based on that of Savage 1954) employed by Karni and Vierø. What is distinctive about the the Jeffrey framework is that all events are treated as propositions towards which an agent has both an epistemic and a desire attitude. That is, no distinction is made between the objects of actions, desires, and beliefs.¹¹ But our model has a further subjectivity that is more in keeping with the economists' approach: we treat the relevant set of possibilities as dependent on an agent's state of awareness, in that the set changes as awareness grows.

Philosophers have not embraced this kind of "subjective state space" (as Schipper (2015) puts it, in an important review of economists' and computer scientists' work on (un)awareness). For the reasons discussed in section 3.3.1 above, the natural move when working with objective possible worlds is to introduce a catch-all proposition standing for "the other ways the world might be", where this catch-all need not be something the agent herself entertains. That is the approach taken by various philosophers seeking to model growing awareness about new scientific theories, including Shimony (1970), Earman (1992), and recently Wenmackers and Romeijn (2016). Bradley's (2017) model does not appeal to a catch-all *per se*, but the agent's point of view is described in contrast to the modeller's point of view, such that the former can be seen as presupposing some proposition to be true and its negation false. Canson (ms) has argued in favour of the agent herself entertaining a catch-all. Indeed, she claims that rationality requires this kind of epistemic humility. While we have some sympathy with these approaches, we question the cogency and/or the usefulness of including a catch-all in a model of an agent's reasoning (recall our arguments above in section 3.3.1).

¹⁰Note that this need not be the aim. The pioneering work on (un)awareness by Fagin and Halpern (1987), for instance, was more about characterising another agent's state of (un)awareness (e.g., in the context of a game).

¹¹See e.g. Joyce (1999) for a discussion of how to translate between Jeffrey's framework and Savage's.

Other philosophers interpret catch-alls in a less open-ended way (or else avoid this terminology)—in that they appeal to abstract propositions that stand in for classes of contingencies whose members are yet to be better articulated. As a result, however, these models are highly constrained in terms of the (un)awareness and awareness growth that is permitted. For instance, Maher (1995) assumes that the agent's algebra contains variable propositions for *each* of the yet-to-be-formulated theories, and he moreover assumes that the agent assigns a (non-zero) probability to each such proposition. Henderson et al. (2010) propose something similar, although with the added sophistication that the propositions in the agent's algebra form a hierarchy that remains fixed throughout the investigation, that is, remains unchanged even when the agent becomes aware of new theories that effectively fill in this hierarchy.

It is also worth mentioning the well known proposal of Zabell (1992) in this context. Zabell extends statistical inference to cases where previously unsuspected phenomena of a given kind may occur (such as in the so-called *sampling of the species problem*). As in Henderson et al., the probability function is defined over a set of hypotheses that are sufficiently abstract to accommodate all the possible phenomena of the given kind, whatever they turn out to be. Moreover, by construction, the probability function does not depend on how exactly the abstract hypotheses are instantiated.

We will argue in chapters 6 and 7 that the appeal to abstract propositions that are placeholders for yet-to-be-fully-articulated contingencies is an apt way to model *only* the special case whereby an agent *anticipates* her own awareness growth. Note that economists too have appealed to “catch-alls” of this sort. For instance, Grant and Quiggin (2013a, 2013b) incorporate such a catch-all in their model; it is assigned a probability based on the agent's past experience of limited awareness. In one of their more recent papers, Karni and Vierø (2017) introduce a “catch-all” consequence to allow for the agent anticipating her awareness growth, which they also call “awareness of unawareness”. This psychological phenomena is also sometimes dubbed “conscious unawareness” (Walker and Dietz 2011), or “introspective unawareness” (Piermont 2017). We suggest that these models do not incorporate what is strictly speaking a catch-all, i.e., a proposition standing for *all* other possibilities of which the agent is unaware (since this is not obviously cogent), but rather appeal to an abstract proposition standing for

a broad class of contingencies that the agent expects to “fill in” later. The same is true of the “catch-all” model we introduce in chapter 6.

An agent need not always anticipate her own awareness growth, however, such that it can be treated formally as a refinement. We insist that there can be genuine cases of awareness growth by expansion. That is why the notion of a subjective and changing possibility space is important. In this respect, our model draws on the work of economists Heifetz et al. (2006, 2008). In particular, we draw on the idea that an agent’s possibility space is constructed from combinations of the basic propositions of which she is aware, which may change over time. Heifetz et al. (2006: 80) model progressive awareness change in terms of “a complete lattice of disjoint spaces”. Similar to our characterisation of propositions above, Heifetz et al. also refer to a “surjective projection” from more aware to less aware spaces, i.e., every atomic possibility in the more aware state maps onto a single atomic possibility in the less aware state, whereas some possibilities in the less aware state map onto more than one atomic possibility in the more aware state. The preoccupations of these economists—and indeed the finer details of their model—differs, however, from ours. The particular problems of awareness growth that we are concerned with are the content of the remainder of this book.

3.5 Concluding remarks on chapter 3

The main aim of this chapter, as the reader will recall, was to develop a model of reasoning and choice that is more subjective than even traditional subjective expected utility theory, in that our model makes room for the possibility that agents differ in the extent to which they are aware. Although we have now developed such a model, and explained how it can be used to characterise intuitively different ways in which awareness can grow, we have not yet considered how an agent should change her credence when awareness grows. That will be the topic of the next two chapters: 4 and 5. Moreover, while we have suggested that agents sometimes can, and arguably should, anticipate that their awareness will grow, we have not examined how they should factor this anticipation into their reasoning. That will be the topic of chapters 6 and 7.

4

Responding to Awareness Growth

4.1 Introduction

So we have introduced awareness growth informally and formally. How should an agent respond to such growth? That is what we turn to now. We consider what sort of impact awareness growth has on the reasoning of a rational agent.

We know that awareness growth is different from the type of learning that philosophers and decision theorists typically consider. The typical case is where the agent comes to have a new credence in propositions representing familiar contingencies. (Recall our characterisations of different types of learning in chapter 2, table 2.2.) With awareness growth we are talking about coming to entertain propositions that represent previously unfamiliar contingencies. This is not a learning experience that can be characterised in the usual way: as a constraint on the agent's probability function over a given possibility space. It rather involves a revision of this very possibility space.

This chapter and the next investigate the extent of the analogy between awareness growth and traditional kinds of learning. One way to put the question is as follows: What is the parallel between norms for belief or credence revision under awareness growth and regular so-called *Bayesian* norms for belief revision?¹ (Recall from chapter 2, in particular footnote 3,

¹Throughout this book, we focus on *degrees* of belief, rather than on outright (all-or-

that the probabilist model of belief, together with the norm of belief revision known as *conditionalisation*, is collectively referred to as the *Bayesian* model of belief, or simply as *Bayesianism*.²)

The standard answer in the literature—at least to the extent that there can be said to be a “standard” answer when it comes to awareness growth—is that there is a strong resemblance between awareness growth and ordinary learning, and similarly between the norm for awareness growth and Bayesian conditionalisation. In the case of awareness growth, the norm has become known as *Reverse Bayesianism*, a term coined by economists Karni and Vierø, who have devised influential decision-theoretic arguments in favour of the norm (2013, 2015). Informally, Reverse Bayesianism states that when a person becomes aware of new contingencies, she should update her credences “in such a way that likelihood [probability] ratios of events in the original [epistemic] state space remain intact” (2013: 2801). Insofar as they have addressed belief revision under awareness growth, philosophers too (notably, Wenmackers and Romeijn 2016, and Bradley 2017) have endorsed what is effectively Reverse Bayesianism.

In this chapter we investigate the Reverse Bayesianism response to the question posed above. We initially, in sections 4.2 and 4.3, explain why Reverse Bayesianism has a strong resemblance to regular Bayesian belief revision. We go on, however, in section 4.4, to pose putative counterexamples to Reverse Bayesianism. Our diagnosis of these cases leads us to conclude that they are indeed genuine counterexamples to Reverse Bayesianism, which thus cannot be a general requirement of rationality.

4.2 Traditional Bayesianism

Let us first lay out in detail the traditional Bayesian account of how credences should be revised in response to learning.³ We note upfront that the finer details of Bayesian norms of belief revision are controversial. For instance, there is disagreement about whether the norms govern the *actual* transitions in an agent’s credences through time, or merely an agent’s *plans*,

nothing) belief which some epistemologists fancy. Thus, when we talk about “belief revision” we mean revision of degrees of belief, that is, revision of one’s credences.

²Note that *Bayesianism* is often taken to incorporate the expected utility principle as well.

³A very brief account of Bayesian learning was given in chapter 2, specifically in section 2.2.

at some given time, for how to revise her credences in response to learning new things. For many purposes, such finer interpretative questions arguably do not matter. We hope that interpretative issues do not matter for our purposes either, but that is not something we can take for granted. We seek belief revision norms for awareness growth that are the counterpart of norms for ordinary learning. Our response will be more compelling to the extent that it does not depend on a specific and controversial interpretation of these norms for ordinary learning.

4.2.1 The rule of conditionalisation

We will get to the nuances shortly. Let us first present the basics. The rule of conditionalisation—one of the core theses of Bayesian epistemology—states that for any proposition B , the agent's degrees of belief or credence in B , after learning A (and nothing stronger), should equal her (prior) conditional credence in B given A , i.e., $P(B | A)$, which, according to the standard definition of conditional probabilities, equates to $P(A \& B) / P(A)$ whenever $P(A) > 0$. (It is typically assumed that the agent would never learn something to which she had assigned zero probability.⁴) More formally, let P_A represent our agent's credences after she has learned A . Then the rule of Conditionalisation states that:

Conditionalisation. *For any $A, B \in \mathcal{F}$ and according to any rational agent:*

$$P_A(B) = P(B | A) \quad \text{assuming that } P(A) > 0$$

Given the standard definition of conditional probabilities stated above, Conditionalisation is logically equivalent to the conjunction of the following two principles:

Certainty. $P_A(A) = 1$

Rigidity. $P_A(B | A) = P(B | A) \quad \text{assuming that } P(A) > 0$

Informally, Certainty says that the agent is certain of whatever she has learned. Rigidity, on the other hand, says that whatever proposition the

⁴We leave it as an open question whether or not the relevant conditional probabilities can otherwise be defined for cases where the proposition conditioned on is assigned zero probability (see Hájek 2003 for discussion). In general, there is a question which we do not address in this book of how an agent should revise her credences if she were to learn that a proposition which she had assigned zero probability is in fact true.

agent may learn, her degrees of belief conditional on this proposition are rigid, or unchanged by the learning experience. These two principles thus reflect a neat division between those beliefs directly affected by the learning experience (described by Certainty), and those beliefs that are not affected by the learning experience and are thus unchanged (as per Rigidity). In this way, the rule can be described as a “conservative-change” maxim: “hold fixed any beliefs that are not directly affected by the learning experience”.

It has been well noted that the Certainty condition does not encompass all kinds of learning. For starters, it does not fit well with an intuitive notion of learning according to which one could take oneself to have learned something without having become certain of some proposition. Fortunately, the Bayesian framework can be straightforwardly extended to learning experiences where an agent does not learn anything with certainty, without giving up Rigidity, as Richard Jeffrey (1965) proposed.⁵

4.2.2 Interpreting the norm

We said that conditionalisation articulates conservative belief change. One might defend such a rule on the basis that an agent should adjust her credences in response to learning only to the extent that the learning is revelatory. Any further change would be arbitrary. For instance, when an agent learns a proposition *A* with certainty, the only change that this licences is, first, a credence of 1 in *A*, and, second, the corresponding credence change in other propositions brought about by conditionalising on *A*. All other credences should stay the same. In particular, all conditional credences should stay the same (as per Rigidity), and so should the credence in any proposition *B* that the agent takes to be independent of *A* in the sense that conditionalising on *A* does not change her credence in *B*.

One might wonder why an austere approach to learning is rationally required, and if so, whether an austere approach amounts to conditionalisation. But let us put this big substantive question to the side. We focus rather on the interpretative question introduced above: Is the relevant norm about transition or planning? That is, are we talking about whether an agent should *actually* or rather merely *plan to* change her credences according to

⁵The maxim has also been extended to cases where the information one gains affects one’s conditional probabilities (in the form of the rule known as *Adams conditionalisation*) (Bradley 2005).

the rule of conditionalisation?⁶ The hope would be that it does not matter: either interpretation of Bayesian conditionalisation will have an analogue in the case of awareness growth.

Indeed, defenders of the Bayesian model (so-called *Bayesians*) have taken different sides on questions of interpretation. The standard reading of conditionalisation is arguably the transition reading: that it is a genuinely diachronic norm governing the relationship between an agent's credences at different times, before and after learning. The planning interpretation is more attractive, however, to those who are uncomfortable with the idea of genuinely diachronic norms of rationality, and who doubt that such norms can be defended.

The problem, when it comes to remaining ecumenical about the interpretation of belief revision norms, is that any planning norm does not seem to have an obvious analogue in the context of awareness growth. There is no sense in which an agent can plan for a *particular* awareness growth, as that would contradict the very nature of the phenomenon. That said, one might suppose that an agent can plan for some generically-described awareness growth. Indeed, we explore such plans in detail in chapters 6 and 7, under the auspices of "anticipated awareness growth". Accordingly, one might regard the norm we seek, the analogue of conditionalisation, to be a constraint on one's planned credence change in the event of some generically-described awareness growth. While we do return to this idea in chapter 7.4, for now it is worth noting that there are certain limitations to the kinds of awareness growth that one can anticipate. For instance, one cannot anticipate what is truly awareness growth by expansion. It seems that, insofar as we are looking for a parallel norm for *all kinds of* awareness growth, it is more straightforward to have in mind the actual-transition version of Bayesian conditionalisation.

It is however worth asking: Is there yet another way to interpret norms of belief change in the standard setting that can also be extended to awareness growth? Let us suggest one that also explains the conservative nature of conditionalisation. The main idea is that conditionalisation is not a

⁶Others have suggested that there is no direct norm governing credence change at all. Rather, conditionalisation is a mere consequence of rational agents having credences that fit the evidence at any given time (Hedden 2015). But this picture involves substantial evidential constraints on credence and lies well outside the subjectivist picture we are pursuing here.

substantive norm at all but rather a mere description of what it means to be a stable agent upon learning. That is, an agent who learns some proposition for sure changes her credences in accordance with conditionalisation *if* she remains (epistemically speaking, at least) the same agent. After all, a stable agent is presumably one who changes only in very conservative ways. When it comes to the actual transitions in an agent's credences then, she is not irrational in any sense if her changes in credence do not accord with conditionalisation. She is simply not the same agent. Similarly, in the case of planning, it is not irrational for an agent to predict that she will not revise her credences in accordance with conditionalisation. There is nonetheless something regrettable about this kind of self-knowledge: in effect, the agent predicts she will not be a stable agent through time, and she may furthermore predict that there will be costs associated with her instability.

We have introduced this latter interpretation of Bayesian conditionalisation to allow for at least two ways to think about what a "norm" of belief change for growing awareness amounts to. That is to say that one can read the quest of this chapter in at least two ways. One option is that we seek a genuinely diachronic norm for belief change under growing awareness. That was the presumption in our introductory remarks. We ask: how should an agent change her credences in response to a given growth in awareness? Another option, for those sceptical of genuinely diachronic norms (and who moreover doubt the prospects for the relevant "planning" norm), is that we seek a mere description of, or criteria for, what it means to be a stable agent through time, having experienced awareness growth. We will proceed using language that is more fitting to the former interpretation, but everything we say can rather be read along the lines of the latter.

4.3 Reverse Bayesianism

The "standard" response to the question of how one should change one's credences in response to awareness growth, is, as mentioned, known as Reverse Bayesianism. We note upfront a fixed point in our discussion of Reverse Bayesianism and potential rivals. We take *probabilism* to be non-negotiable: we are assuming that a rational agent's credences *in any given awareness context* must satisfy the probability axioms. Recall from chap-

ter 3.3 that an *awareness context* is defined by the set of *basic propositions* of which the agent is aware, from which a Boolean algebra can be generated.⁷ A brief comment is in order here: The tautology, which has probability one according to the probability axioms, must be interpreted such that it depends on the awareness context: it is associated with the set of all (epistemic) possibilities in that context, which corresponds, for instance, to $A \vee \neg A$, for any A in the context.

The question is whether and how an agent's subjective probability function for one awareness context constrains or relates to her subjective probability function once she has experienced a growth in awareness. Let's consider a variant of the rent-or-buy example from the last chapter. This new example involves a slightly more complicated set of contingencies. In your least aware epistemic situation, represented by table 4.1, you perceive the outcome of your decision as to whether to continue to rent or instead buy an apartment to depend not just on how the rental price will change, but also on whether the new neighbours would turn out to be noisy or quiet. Note that in order to make these two sorts of properties of the world vivid, table 4.1 suppresses the possible actions of "Rent" and "Buy"; the table is rather a two-dimensional representation of the state space.

Now suppose that having found yourself in the least aware epistemic situation, you become aware that your rental apartment might be sold, as per table 4.2. How should you revise your credences in the various other propositions in light of this expansion? Or, instead, suppose that in the situation represented by Table 4.2, you realise that changes to the interest rate affects the outcome of buying an apartment, as per table 4.3. How should this refinement affect your credences in other propositions?

Traditional Bayesianism is silent on these two questions. As we have seen, this is not a type of learning experience that the traditional Bayesian framework incorporates. But recently, Karni and Vierø have defended a unified answer to these two questions (at least for the particular kind of decision problem and awareness growth that they represent) in the form of a principle that they call "Reverse Bayesianism".

Let us state Reverse Bayesianism as if it were a general principle transcending the particular type of decision model formulated by Karni and

⁷In chapter 3.3 we specified that we understand *basic propositions* to be primitive propositions (representing simple facts about the world) that do not involve any logical connectives.

4. RESPONDING TO AWARENESS GROWTH

	Rent higher	Rent same or lower
Neighbours noisy	Neighbours noisy & Rent higher	Neighbours noisy & Rent same or lower
Neighbours quiet	Neighbours quiet & Rent higher	Neighbours quiet & Rent same or lower

Table 4.1: State space for rent-or-buy

	Rent higher	Rent same or lower	Rental apartment sold
Neighbours noisy	Neighbours noisy & Rent higher	Neighbours noisy & Rent same or lower	Neighbours noisy & Rental apartment sold
Neighbours quiet	Neighbours quiet & Rent higher	Neighbours quiet & Rent same or lower	Neighbours quiet & Rental apartment sold

Table 4.2: Expanded state space for rent-or-buy

	Rent higher		Rent same or lower		Rental apartment sold	
Neighbours noisy	Neighbours noisy & Rent higher & Interest higher	Neighbours noisy & Rent higher & Interest same or lower	Neighbours noisy & Rent same or lower & Interest higher	Neighbours noisy & Rent same or lower & Interest same or lower	Neighbours noisy & Rental apartment sold & Interest higher	Neighbours noisy & Rental apartment sold & Interest same or lower
Neighbours quiet	Neighbours quiet & Rent higher & Interest higher	Neighbours quiet & Rent higher & Interest same or lower	Neighbours quiet & Rent same or lower & Interest higher	Neighbours quiet & Rent same or lower & Interest same or lower	Neighbours quiet & Rental apartment sold & Interest higher	Neighbours quiet & Rental apartment sold & Interest same or lower

Table 4.3: Expanded and refined state space for rent-or-buy

Vierø (more on which below). We use P (P^+) to represent the probabilistic degrees of belief of the agent before (after) awareness grows. As per the notation in chapter 3.3, \mathbf{X} (\mathbf{X}^+) is the set of basic propositions of which the agent is aware before (after) awareness grows. Reverse Bayesianism holds that the ratio between the probabilities of any two inconsistent basic propositions in the old epistemic state (that each had positive probability) should not change when awareness grows. More formally:

Reverse Bayesianism. *For any $A, B \in \mathbf{X}$ (where $P(A \& B) = 0$, $P(A) > 0$ and $P(B) > 0$) and according to any rational agent:*

$$\frac{P(A)}{P(B)} = \frac{P^+(A)}{P^+(B)}$$

Versions of this principle have more recently been endorsed by Wenmackers and Romeijn (2016) and Bradley (2017).

Consider what Reverse Bayesianism requires in the rent-or-buy variant presented above. Suppose you find “Neighbours noisy” to be twice as likely as “Neighbours quiet” before realising that the apartment could be sold. Then after this realisation, you should still find “Neighbours noisy” to be twice as likely as “Neighbours quiet”. Similarly, after you realise that the central bank might change the interest rate, you should still find “Neighbours noisy” to be twice as likely as “Neighbours quiet”. On the face of it, these implications of Reverse Bayesianism might seem intuitive. For why should the prospect of the the apartment being sold, say, change one’s relative credence in whether the new neighbours will be noisy vs. quiet?

One might surmise that Reverse Bayesianism is compelling because it precisely captures conservative belief change for the learning experience in question—awareness growth. Indeed, its defenders take it to be the consequence of something akin to the Rigidity condition for this kind of learning experience. Bradley, for instance, says as much:

Within the Bayesian framework, conservation of the agent’s relational beliefs is ensured by the rigidity of her conditional probabilities. So we can conclude that conservative belief change [when faced with growing awareness] requires [that] the agent’s new conditional probabilities, given the old domain, for any members of the old domain should equal her old unconditional probabilities for these members. (2017: 258)

Wenmackers and Romeijn similarly suggest that the conservation of “probability ratios among the old hypotheses” follows from the relevant conditional probabilities remaining constant:

In analogy with Bayes’ rule, one natural conservativity constraint is that the new [i.e., more aware] probability distribution must respect the old [i.e., less aware] distribution on the pre-existing parts of the algebra [i.e., on the distributions’ shared domain]. (2016: 1235)

Karni and Vierø also appeal to the constancy of conditional attitudes by way of defending Reverse Bayesianism. In the behaviourist economics tradition, they appeal to constraints on preferences, and only indirectly on beliefs:

... as the decision-maker’s awareness of consequences grows and his state space expands, his preference relation conditional on the prior state space remains unchanged. (2013: 2801)

The above defences of Reverse Bayesianism are arguably sound given the models of awareness growth to which they pertain. As mentioned earlier, however, these models place limitations on the kinds of awareness growth and/or the beliefs that are subject to the Reverse Bayesianism rule. Karni and Vierø, for instance, employ an Anscombe and Aumann (1963) framework, which, (like Savage’s 1954 framework, on which it is based), consists of acts, maximally specific consequences, and states amounting to act-consequence pairs. For Karni and Vierø, in cases of awareness growth by expansion, the agent ultimately comes to be aware of states that are by their very nature inconsistent with the states that define her old awareness context.

Philosophers tend to prefer a more general Jeffrey (1965)-inspired propositional framework, but nonetheless introduce similar restrictions to Karni and Vierø in their discussions of belief change under growing awareness. Wenmackers and Romeijn (2016) focus on changes to sets of scientific theories that are assumed to be mutually inconsistent. Bradley’s interests are more general, but he too, in his endorsement of Reverse Bayesianism for awareness growth by expansion, at least, focuses on propositions that are inconsistent with those the agent comes to be aware of:

... the key to conservative attitude change in cases where we become aware of prospects *that are inconsistent with those that*

we previously took into consideration is that we should extend our relational attitudes to the new set in such a way as to conserve all prior relational beliefs . . . (2017: 257, emphasis added)

Moreover, we hold that Bradley implicitly assumes only “vanilla” kinds of awareness growth by refinement, as our discussion in the next section will reveal.

We allow that Reverse Bayesianism may be defensible in the limited setting that the above authors consider. But the question remains as to whether this learning rule is defensible in a more general setting (as Bradley, at least, suggests). In the next section, we show that the answer to this question is negative: we offer some informal counterexamples to Reverse Bayesianism.

4.4 Counterexamples to Reverse Bayesianism

It is not hard to see that Reverse Bayesianism cannot generally be true once we move beyond the constrained models of its defenders. That is, one can devise examples where Reverse Bayesianism is violated without irrationality on behalf of the agent in question. All we need are examples where awareness grows since an agent becomes aware of a proposition that she takes to be evidentially relevant, intuitively speaking, to the comparison of propositions of which she was already aware. For in that case, the ratio between probabilities of propositions of which the agent was already aware will not stay the same; one proposition will become more probable compared to the other, just like in ordinary cases where one learns evidence relevant to the comparison of hypotheses.

In fact, the history of science is full of examples that undermine Reverse Bayesianism, for the above reason. Here is a particularly prominent such example:

Example 1. *Nineteenth century physicists were unaware of the Special Theory of Relativity (STR). That is, not only did they not take the theory to be true; they had not even entertained the theory. We can suppose, however, that they had entertained various propositions for which the theory was regarded evidentially relevant, once Einstein brought the theory to their attention. In particular, they did (rightly) take the theory to be evidentially relevant to various propositions about the speed*

of light, such as whether the speed of light would always be measured at 300,000 km/s independently of how fast the investigator is moving or whether the measured speed would differ, depending on how fast the investigator is moving. But then the awareness and subsequent acceptance of the STR changed their relative confidence in such propositions.

Not all examples where Reverse Bayesianism fails come from the history of science. Here is a more mundane, or everyday, example:

Example 2. *Suppose you happen to see your partner enter your best friend's house on an evening when your partner had told you she would have to work late. At that point, you become convinced that your partner and best friend are having an affair, as opposed to their being warm friends or mere acquaintances. You discuss your suspicion with another friend of yours, who points out that perhaps they were meeting to plan a surprise party to celebrate your upcoming birthday—a possibility that you had not even entertained. Becoming aware of this possible explanation for your partner's behaviour makes you doubt that she is having an affair with your friend, relative, for instance, to their being warm friends.*

There is an important difference between the two examples. In the first example, the awareness growth consists in the recognition of a completely new idea that the scientists had never heard or thought about before. In the second example, by contrast, the agent has, we can assume, heard of surprise parties before; it is just that in the situation and the moment in which he/she finds him/herself, the possibility is not part of his/her awareness. As we however pointed out in the introductory chapter, our interest is the role that an agent's epistemic state plays in her deliberation about how to act. And since "unawareness" due to never having heard about an idea generally plays the same role in deliberation as "unawareness" merely in the moment of deliberation, we treat these two types of epistemic limitations in the same way.

A defender of Reverse Bayesianism might argue that the above two examples do not undermine their thesis, since, for instance, the proposition picked out by the sentence "the speed of light will always be measured at 300,000 km/s independently of how fast the investigator is moving" is different before and after the speaker becomes aware of the Special Theory of Relativity. (Similarly, the proposition picked out by the sentence "my partner and best friend are having an affair" is different before and after

the speaker realises that their partner and best friend might be organising a surprise party.) That is, it is not just that the propositions in question are *understood differently*, given a change in the underlying possibility space (as per our own approach, detailed in chapter 3.3); rather what appear to be the same propositions across awareness contexts are in fact entirely different propositions. For instance, the physics case might be spelled out as follows: despite appearances, the agent's growth in awareness is not simply an expansion of the "fundamental physical theory" partition to include the STR; there is also an expansion of the "light hypothesis" partition to include the STR versions of the (speed-of-) light hypotheses. As a result, the addition of the STR has no bearing on the original (speed-of-) light hypotheses, in conformity with Reverse Bayesianism. It might be added that, if the new propositions of which the agent becomes aware *were* apparently evidentially relevant to the basic propositions in the old awareness context, then we would *not* have a case of *genuine* awareness growth, to which Reverse Bayesianism is limited.⁸

This way of saving Reverse Bayesianism however seriously weakens the commonsense appeal and normative interest of the thesis, and seems rather *ad hoc*, as the examples under consideration are surely as genuine cases of awareness growth as any. Moreover, if the aim is to represent the internal perspective of an agent, then it is surely more natural to take the individuation of propositions at face value, such that, with respect to our example above, the speed-of-light hypothesis corresponds to the same proposition before and after recognition of the Special Theory of Relativity. But that means that new propositions may well have a bearing on the relative probabilities of old basic propositions. Better to modify the Reverse Bayesian principle itself than to modify what counts as genuine awareness growth.

So, we can conclude that we should not impose Reverse Bayesianism as a general constraint on how a rational agent can revise her credences when her awareness grows. The above counterexamples, however, both involve what we called awareness growth by *expansion*. But as previously mentioned, proponents also want to impose Reverse Bayesianism as a constraint on how a rational agent can revise her credences when her awareness grows

⁸The implication is that we would rather have a case of irrational and/or poorly represented belief change.

due to *refinement* (see e.g. Karni and Vierø 2013: 2803). And one might well hope that despite the above counterexamples, the principle could be retained for belief revision due to refinement.

Unfortunately, counterexamples similar to those discussed above also undermine Reverse Bayesianism understood in this latter way. Consider a third example:

Example 3. *Suppose you are deciding whether to see a movie at your local cinema. You know that on the day in question, the cinema only shows “international” (non-English) movies. You realise that both the movie’s language and genre will affect your viewing experience. The possible languages you consider are French and German and the genres you consider are thriller and comedy. But then you realise that, due to your poor French and German skills, your enjoyment of the movie will also depend on the level of difficulty of the language. Since you know the owner of the cinema to be simple-minded, you are, after this realisation, much more confident that the movie will have low-level language than high-level language. Moreover, since you associate low-level language with thrillers, this makes you more confident than you were before that the movie on offer is a thriller as opposed to a comedy.*

The important feature of the above example is that the original awareness context is partitioned according to some property (the language level) that is taken to be evidentially relevant to the comparison of some pair of inconsistent basic propositions—that the movie is a thriller and that it is a comedy—in the old awareness context.

There is one remaining potential objection to, in particular, examples 2 and 3 that we should address before proceeding. The objection is that what happens in these examples is that, in example 2, you realise that you did not to begin with have as much evidence as you had thought for your partner having an affair, whereas in example 3 you realise that you had evidence all along that the movie would be a thriller. So, in both cases, you realise that your previous epistemic state was flawed.⁹

Now, we need not reject the above characterisation of these examples for them to be counterexamples to Reverse Bayesianism. After all, the reason you realise that your previous epistemic state was flawed, is that you become aware of new contingencies that show that you either had more or else less evidence than you thought. In the movie example you

⁹Thanks to Teru Thomas for suggesting this interpretation.

become aware of a fact that connects what you already knew to the movie being a thriller, whereas in the surprise party example you become aware of a new possible explanation that convinces you that the evidence you took yourself to have for your partner's affair wasn't strong. So, even though you do realise that your previous epistemic state was flawed, you do so by becoming more aware, and this growth in awareness need not align with the requirements of Reverse Bayesianism to be rational.

In sum, the above examples show that Reverse Bayesianism cannot hold in full generality, neither as a constraint on belief revision due to expansion nor as a constraint on belief revision due to refinement. Before closing, however, we note a different potential criticism of our analysis. It might be argued that our examples are not illustrative of a simple learning event (a simple growth in awareness); rather, our examples illustrate and should be expressed formally as complex learning experiences, where first there is a growth in awareness, and then there is a further learning event that may be represented, say, as a Jeffrey-style or Adams-style learning event.¹⁰ In this way, one could argue that the awareness-growth aspect of the learning event always satisfies Reverse Bayesianism (the new propositions are in the first instance evidentially irrelevant to the comparison of the old basic propositions). Subsequently, however, there may be a revision of probabilities over some partition of the possibility space, resulting in more dramatic changes to the ratios of probabilities for the old basic propositions. The reason we reject this way of conceiving of the learning events described by our examples is that the two-part structure is ultimately unmotivated. The second learning stage is an odd, spontaneous learning event that would be hard to rationalise. Hence, this would again seem to us to be an artificial and *ad hoc* way to save Reverse Bayesianism.

4.5 Concluding remarks on chapter 4

In this chapter we set out to determine what are the norms for belief revision under growing awareness that are the suitable parallel to or extension of the traditional Bayesian norms for belief revision. We have seen that the arguably most-worked-out proposal to this effect, namely, Reverse Bayesianism, does not hold in general. That is, there are examples where an

¹⁰This suggestion resonates with the discussion in Hill (2010).

agent's awareness grows in a way that conflicts with Reverse Bayesianism, without any apparent irrationality or undue lack of conservatism on behalf of the agent. The further question, to be pursued in the next chapter, is whether there is some alternative norm of belief change for growing awareness. More generally, what are the features, if any, of conservative belief change when one becomes aware of new contingencies?

5

Awareness Rigidity

5.1 Introduction

In the last chapter we argued that Reverse Bayesianism fails with respect to propositions A and B if the awareness growth favours one of these propositions over the other. What happens in these cases is that the “new” propositions that the agent comes to be aware of change the relationships between the “old” propositions. Figure 5.1 provides a stylised (*not to scale!*) pictorial illustration of this phenomenon with respect to the STR example of awareness growth by expansion. The figure makes clear that the propositions concerning whether or not the speed of light is relative to the observer, denoted “Relative Light” and “Non-relative Light” respectively, are associated with different sets of possibilities when awareness about fundamental scientific theories grows to include the Special Theory of Relativity, “STR”.

Figure 5.1: Expansion to include STR

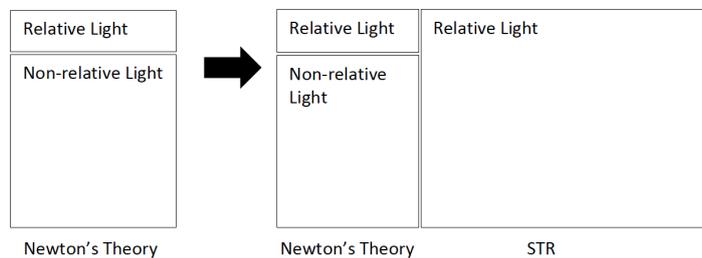


Figure 5.1 moreover suggests a retreat from Reverse Bayesianism to the kind of rigidity principle defenders of Reverse Bayesianism apparently

take as fundamental. (Recall the quotes from chapter 4.3 which reveal that defenders of Reverse Bayesianism take this principle to be the consequence of appropriate conditional attitudes remaining constant or rigid under growing awareness). Informally, the idea is that the probabilities of the old propositions, conditional on, roughly speaking, “how things were before” (in our example, the proposition “Newton’s Theory”), should be rigid or unaffected by the awareness growth (in our example, the expansion of the fundamental theory space to include “STR”). Such a rigidity principle looks to be the appropriate basis for conservative belief change in the awareness-growth setting.¹ It is just that the condition does not entail Reverse Bayesianism when stated in general terms, namely, for all pairs of inconsistent basic propositions. Or so the argument might go.

Even if this position were roughly right, the relevant rigidity condition would need to be properly spelled out. For instance, what precisely is meant by “how things were before”, which should be conditioned on according to this view? In section 5.2, we appeal to the model we introduced in chapter 3.3 in order to meet this challenge. We give a precise definition of the intended rigidity principle, which we dub *Awareness Rigidity*. In section 5.3, we go on to argue against Awareness Rigidity. The prospects for a general principle for belief change under growing awareness are thus dim. Nonetheless, we proceed, in section 5.4, to formulate what we call *Restricted Reverse Bayesianism (RRB)*. Even if RRB is not a substantive norm for belief revision under growing awareness, it may yet serve as a guide for detecting cases where we can expect the type of conservatism enshrined in Bayesianism to hold.

5.2 Awareness Rigidity defined

Let us start then by spelling out the rigidity condition that plausibly constrains belief change under growing awareness. The hope would be that this rigidity principle does not imply Reverse Bayesianism in the situations in which we do not want the latter to hold, that is, in situations where the

¹In fact, Karni and Vierø (2013) formulate the preference analogue of such a rigidity principle, which they call *Awareness consistency*, and use it in their derivation of Reverse Bayesianism. (Karni and Vierø 2015 however use a somewhat weaker axiom.) Moreover, as we point out below, Bradley (2017) explicitly assumes what is essentially Awareness Rigidity (although giving it a different name).

awareness growth is intuitively evidentially relevant to the comparison of some propositions.

Our suggestion for specifying a rigidity condition for awareness growth is to identify the smallest set of possibilities in the new awareness context that corresponds to what used to be the tautology in the old awareness context. For the example described in figure 5.1, this proposition will indeed be “Newton’s theory”, as per our informal discussion above. With respect to our rent-or-buy example, when awareness grows by expansion to incorporate the new contingency “Apartment is sold”, as per the shift from table 4.1 to table 4.2, the proposition corresponding to the smallest set of new possibilities and which used to be the tautology in the old awareness context is the disjunction of all the old Landlord contingencies, i.e., “Rent higher \vee Rent same or lower”. In special cases of awareness growth by refinement where *all* possibilities are effectively refined (what we refer to as *complete* refinement), the relevant proposition will simply correspond to the entire set of possibilities constituting the new awareness context. So, for instance, when awareness grows by refinement to incorporate the new contingencies concerning the interest rates, as per the shift from table 4.2 to table 4.3, the proposition corresponding to the smallest set of new possibilities and which used to be the tautology in the old awareness context is, for instance, “Rent higher \vee Rent same or lower \vee Apartment is sold”. This proposition corresponds to the tautology in the new awareness context.

This allows us to specify a rigidity condition that one might take to be the appropriate extension of Bayesian (i.e., conservative) belief change to the case of growing awareness:

Awareness Rigidity. *Let T^* in \mathcal{F}_X be the proposition that, amongst those associated with the full set \mathbf{W}_X , is associated with the smallest subset of \mathbf{W}_{X^+} . For any rational agent and for any $A \in \mathcal{F}_X$:*

$$P^+(A | T^*) = P(A)$$

This, we think, captures the rigidity condition that defenders of Reverse Bayesianism take as more fundamental than Reverse Bayesianism itself. In fact, setting aside subtle differences when it comes to the interpretation of the background algebra (in particular, whether parts of it remain constant during awareness change, which we deny), this is precisely the norm

that Bradley (2017: 258) endorses as the appropriate extension of Bayesian conservatism to situations where awareness changes.²

Figure 5.1 makes vivid that Awareness Rigidity does not generally entail Reverse Bayesianism, at least for cases of awareness growth by expansion. Awareness Rigidity requires that the probabilities for old propositions—“Relative Light” and “Non-relative Light”—*conditional on “Newton’s theory”*, remain constant when awareness grows. One can see just by looking at the figure that it does not follow that the ratio of the absolute probabilities for “Relative Light” and “Non-relative Light” remain constant when awareness grows. For this particular case of awareness growth, then, Awareness Rigidity does seem to precisely capture what aspect of the agent’s credences should stay constant as awareness grows.

Note that we might have otherwise defined Awareness Rigidity in a different but logically equivalent way, by concentrating on those propositions that correspond to the agent’s atomic possibilities in the old awareness context. Awareness Rigidity requires that the relative probabilities of these propositions remain constant. For instance, with respect to the awareness growth by expansion described in figure 5.1, the original possibilities correspond to the propositions “Newton’s Theory & Relative Light” and “Newton’s Theory & Non-relative Light”. It is the ratio of probabilities for these propositions that must remain constant, according to Awareness Rigidity, when the agent becomes aware of a further fundamental theory: “STR”.

5.3 Against Awareness Rigidity

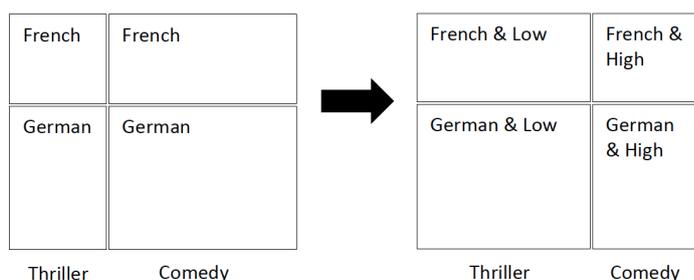
Unfortunately, Awareness Rigidity is not a compelling rationality requirement, especially for cases of refinement, which we turn to now. The problem is that Awareness Rigidity entails Reverse Bayesianism in those special cases where awareness grows by refinement of *all* possibilities, as per the shift from table 4.2 to table 4.3 (since it effectively requires that the probabilities for all propositions in the old awareness context remain unchanged). We previously argued that Reverse Bayesianism is not plausible even in these

²Bradley however uses the term “rigid extension” rather than Awareness Rigidity. More precisely, he calls one (more aware) probability function a rigid extension of another if the two are related by what we call Awareness Rigidity.

special cases of complete refinement. By modus tollens, Awareness Rigidity is then not a plausible principle for belief change.

Is there any reason to retract this position? We think not. Figure 5.2 is a pictorial representation of our third counterexample to Reverse Bayesianism—the refinement-counterexample to Reverse Bayesianism concerning the international movie that will be shown at the local cinema. Figure 5.2 makes vivid how awareness of an entirely new property or kind of contingency may intuitively cause adjustment of the relative probabilities of “old” propositions. Specifically, the relative probabilities of the thriller and comedy genres changes when the agent becomes aware of the language-level contingencies, denoted “High” and “Low”.

Figure 5.2: Refinement into high/low language



While we hesitate to read too much into a formal representation of awareness growth, we suggest that our model introduced in chapter 3.3 offers some explanation for why Awareness Rigidity may be reasonably violated. In short, the set of possibilities associated with any proposition completely changes when awareness grows. Initially the set is constituted by possibilities that are effectively truth functions over a given set of basic propositions; after awareness growth, the set is constituted by possibilities that are truth functions over an enlarged set of basic propositions.

So, it is not surprising that the relationships between propositions, even conditional on “how things were before”, may in some cases dramatically change. This is what happens in the movie example depicted in figure 5.2: when the set of possibilities associated with the movie genres “Thriller” and “Comedy” change, so too do their relative probabilities conditional on “how things were before”, which, in this case, equate to their unconditional probabilities. Examples like this highlight that there is no clear delineation

between those propositions and associated credences that are and are not affected by the awareness growth.

Mahtani (2020) presents an example that also serves as a striking counterexample to Awareness Rigidity in the case of refinement.³ The example goes as follows:

You know that I am holding a fair ten pence UK coin which I am about to toss. You have a credence of 0.5 that it will land HEADS, and a credence of 0.5 that it will land TAILS. You think that the tails side always shows an engraving of a lion. So you also naturally have a credence of 0.5 that (LION) it will land with the lion engraving face-up: relative to your state of awareness, TAILS and LION are equivalent. [...] Now let's suppose that you somehow become aware that occasionally ten pence coins have something other than a lion engraving on the tails side. In particular, you become aware that [STONE] there are some ten pence coins that have an engraving of Stonehenge on the tails side. Let's assume that no other possibilities occur to you. [...]

The awareness growth in Mahtani's example is one of refinement (albeit *partial* refinement as opposed to *complete* refinement, since not all possibilities are refined): you realise that TAILS can be further refined into the disjunction LION \vee STONE. To see that this is a counterexample to Awareness Rigidity, note that the smallest set of possibilities in the new awareness context that corresponds to what used to be the tautology in the old awareness context is LION \vee HEADS. So, according to Awareness Rigidity, we should find that, say, $P^+(\text{LION} \mid \text{LION} \vee \text{HEADS}) = P(\text{LION}) = 0.5$. But that is counterintuitive. By learning that some instances of TAILS are not instances of LION, and nothing more, your credence in LION should diminish. In other words, intuitively, we should find that $P^+(\text{LION} \mid \text{LION} \vee \text{HEADS}) < P(\text{LION})$, in violation of Awareness Rigidity. So, we have another refinement-counterexample to Awareness Rigidity.

Awareness Rigidity might nevertheless be compelling for awareness growth by expansion. After all, when awareness grows by expansion,

³Mahtani proposed the example rather as a counterexample to a restricted version of Reverse Bayesianism that we proposed in earlier draft work. We follow up on this in the next section (in particular, see our reference to Mahtani in footnote 5).

some old propositions are no longer equivalent to the tautology, and hence, Awareness Rigidity does not entail Reverse Bayesianism in cases of expansion. Perhaps Awareness Rigidity is plausible when restricted to expansion, even if Reverse Bayesianism is not. The example involving an expansion of fundamental theories depicted in figure 5.1 suggests as much.

Even in cases of expansion, however, it may be that the gained awareness shakes things up sufficiently in one's old awareness state that Awareness Rigidity is violated. Again, our formal model underscores the fact that *all* propositions are in some way affected by a growth in awareness, whether it be awareness growth due to expansion or refinement. So it is certainly not off the table that Awareness Rigidity may be violated in cases of awareness growth by expansion. We leave as an open question whether that is something that can rationally happen.

5.4 *Restricted Reverse Bayesianism*

Finally, we note that while Reverse Bayesianism must be abandoned as a general principle for belief revision in the case of growing awareness, it is arguably still an interesting relation that *may sometimes* hold between pairs of propositions in the transition from one belief state to another. As we have seen with the earlier rent-or-buy examples, there are circumstances where Reverse Bayesianism does intuitively hold. We suggest that these are cases in which what is learnt is *evidentially irrelevant* to the pairs of propositions at issue. In other words, Reverse Bayesianism is a useful relationship for *characterising* cases of awareness growth that is evidentially irrelevant for the pair of propositions at issue.

One might say that a restricted version of Reverse Bayesianism holds. The relevant principle can be stated as follows:

Restricted Reverse Bayesianism (RBB). For any $A, B \in \mathbf{X}$ (where $P(A) > 0$ and $P(B) > 0$), if

- the awareness growth from \mathbf{X} to \mathbf{X}^+ is evidentially irrelevant for A vs. B
- and Awareness Rigidity holds "locally" with respect to A and B ⁴

⁴That is, the probability of both A and B conditional on "how things were before" (formalised in the previous section) remains constant under the awareness growth from \mathbf{X} to \mathbf{X}^+ .

then according to any rational agent:

$$\frac{P(A)}{P(B)} = \frac{P^+(A)}{P^+(B)}$$

Restricted Reverse Bayesianism is not intended to be a substantive normative principle, since the antecedent conditions—stipulating “local” Awareness Rigidity and evidential irrelevance—secure the conclusion as a matter of logic (we illustrate this below after we formally define evidential irrelevance). RRB is nonetheless supposed to be illuminating in that it reveals sufficient conditions for Reverse Bayesianism to hold for some pair of propositions under awareness growth.

Of course, whether or not it really is the case that Reverse Bayesianism is entailed by the conditions stated in RRB depends on what it means for awareness growth to be “evidentially irrelevant for A vs. B ”. Intuitively, an awareness growth experience is evidentially irrelevant for A vs. B if the awareness growth neither favours A over B nor *vice versa*. More formally, we propose the following account of evidential irrelevance, *at least for special cases of awareness growth* (more on this restriction below):

Definition (Evidential irrelevance). *For any $A, B \in \mathcal{F}_{\mathbf{X}}$, we say that an agent’s awareness growth, from awareness context \mathbf{X} to \mathbf{X}^+ , where \mathbf{X}_j is the set of all basic propositions $X_j \in \mathbf{X}^+$ such that $X_j \notin \mathbf{X}$, is evidentially irrelevant for A vs. B whenever:*

$$\begin{aligned} \text{either (i) } & P^+(A \mid \bigvee_{X_i \in \mathbf{X}_j} X_i) = 0 = P^+(B \mid \bigvee_{X_i \in \mathbf{X}_j} X_i) \\ \text{or (ii) } & \frac{P^+(A \mid \bigvee_{X_i \in \mathbf{X}_j} X_i)}{P^+(B \mid \bigvee_{X_i \in \mathbf{X}_j} X_i)} = \frac{P(A)}{P(B)} \end{aligned}$$

We said that this definition only works for special cases of awareness growth. The special cases are those of pure expansion or pure refinement. In these cases, the “new” propositions of which the agent is aware (those in \mathbf{X}_j) are mutually inconsistent, and we suggest our definition of evidential irrelevance only for such cases. For mixed cases of awareness growth, RRB simply will not be illuminating in the intended way, since evidential irrelevance must itself be defined in terms of whether or not Reverse Bayesianism

holds for the propositions in question.⁵

Let us now briefly illustrate how, given the above definition of evidential irrelevance (and assuming “pure” awareness growth as per above), Reverse Bayesianism logically follows from the two conditions in RRB (i.e., from local Awareness Rigidity plus evidential irrelevance). Note first the new (more aware) space of possibilities, $\mathbf{W}_{\mathbf{X}^+}$, can be partitioned into two sets, on the one hand the possibilities associated with “how things were before”, T^* , and on the other hand the possibilities associated with (the disjunction of propositions in) the set of newly discovered basic propositions, \mathbf{X}_j .

Suppose now that in the shift from $\mathbf{W}_{\mathbf{X}}$ to $\mathbf{W}_{\mathbf{X}^+}$, Awareness Rigidity is satisfied for both A and B , that is, $P^+(A | T^*) = P(A)$ and $P^+(B | T^*) = P(B)$; hence,

$$\frac{P^+(A | T^*)}{P^+(B | T^*)} = \frac{P(A)}{P(B)}$$

For evidential irrelevance, as defined above, there are two cases to consider, (i) and (ii). Let’s first suppose that in addition to the assumed local Awareness Rigidity, (i) holds in the awareness growth from $\mathbf{W}_{\mathbf{X}}$ to $\mathbf{W}_{\mathbf{X}^+}$. That means that neither A nor B is consistent with any of the new possibilities, \mathbf{X}_j . So, from local Awareness Rigidity it then follows that:

$$\frac{P^+(A)}{P^+(B)} = \frac{P(A)}{P(B)}$$

Now suppose instead that in addition to the assumed local Awareness Rigidity, (ii) holds in the awareness growth from $\mathbf{W}_{\mathbf{X}}$ to $\mathbf{W}_{\mathbf{X}^+}$. That means that the ratio of the probability of A to the probability of B conditional on the possibilities associated with \mathbf{X}_j (i.e. all those possibilities not associated with T^*) is the same as the ratio of A and B in the old awareness context. So, again, it then follows that:

$$\frac{P^+(A)}{P^+(B)} = \frac{P(A)}{P(B)}$$

In sum, the two conditions in the statement of Restricted Reverse Bayesianism—

⁵ We are indebted to the work of Anna Mahtani (2020) for inspiring this presentation of RRB as a statement of sufficient conditions for Reverse Bayesianism, rather than a substantive norm. Mahtani’s counterexamples to a version of RRB we proposed in previous draft work prompted us to refine the principle. The previous version did not make all the requisite antecedent conditions explicit in the statement of RRB.

that is, local Awareness Rigidity and evidential irrelevance—logically imply the conclusion in the principle—that is, Reverse Bayesianism. Hence, Restricted Reverse Bayesianism cannot, as we previously noted, be a *substantive normative* principle. However, it is, we think, illuminating in that it states sufficient conditions for Reverse Bayesianism (for “pure” expansion/refinement cases).

Note that evidential irrelevance does not hold in the counterexamples to Reverse Bayesianism discussed in chapter 4.4 (these being cases of pure expansion/refinement as required). For instance, the probability ratio (in terms of probability function P^+) of your partner having an affair with your friend to their being warm friends, conditional on your partner and your friend meeting to organise a surprise party, is lower than the “old” probability ratio (in terms of probability function P) of your partner having an affair with your friend to their being warm friends. Similarly, in the example involving refinement, the probability ratio (in terms of P^+) of the movie being a thriller to its being a comedy, conditional on the disjunction of low- and high-level language, differs from the “old” probability ratio (in terms of P) of the movie being a thriller to its being a comedy. In sum, both in the counterexamples to Reverse Bayesianism involving expansion and in the counterexamples involving refinement, evidential irrelevance is violated.

On the other hand, the more standard examples of awareness growth, such as the rent-or-buy examples, *are* plausibly cases of evidential irrelevance (at least with respect to some pairs of “old” basic propositions), by the above definition. In the expansion case, you become aware that the apartment could be sold. This is irrelevant, for instance, to your relative confidence in the neighbours being noisy vs. quiet, since your new probability ratio (in terms of P^+) of the neighbours being noisy to their being quiet, conditional on the rental apartment being sold, is intuitively the same as your old probability ratio (in terms of P) of the neighbours being noisy to their being quiet. Likewise for the refinement case: the realisation that interest rates could go up or else go down or stay the same is irrelevant, for instance, to your relative confidence in the neighbours being noisy vs. quiet, since your new probability ratio (in terms of P^+) of the neighbours being noisy to their being quiet, both conditional on interest rates either going up or else going down or staying the same, matches your old probability ratio

(in terms of P) of the neighbours being noisy to their being quiet.

5.5 Concluding remarks on chapter 5

Our conclusion then is that there is no general norm of belief revision for awareness growth. (Alternatively—recall our discussion in chapter 4.2.2—one can translate our conclusion as follows: There is no general account of what it means for an agent to be *stable* before and after awareness growth.) Reverse Bayesianism simply characterises those cases in which belief revision under awareness growth happens to be conservative in the familiar way, since the relevant beliefs are not affected by the awareness growth. Restricted Reverse Bayesian describes sufficient conditions for when that is the case.

The spirit of our discussion has been simply to assess whether a rational agent would endorse specific principles for belief revision in response to learning. Diaconis and Zabell (1982: 827) refer to this as a “subjective” mode of justification. They have this mode of justification in mind when stating (e.g., *ibid.*: 822) that standard conditionalisation and its Jeffrey variant are applicable *just in case one judges that Rigidity would hold in the given circumstances*. We have effectively claimed that a rational agent may reasonably judge that neither Reverse Bayesianism nor Awareness Rigidity would hold in various cases of awareness growth.

Of course, there are also “stricter” justifications of conditionalisation, at least when it is interpreted as a *planning* norm,⁶ including the so-called “diachronic Dutch book argument”. On this argument, *planning* to revise one’s beliefs contrary to conditionalisation in the appropriate learning circumstances has bad pragmatic consequences that should be avoided; specifically, it makes one vulnerable to sure loss. One might wonder whether, notwithstanding our counterexamples, there is an argument of this sort for Reverse Bayesianism, or perhaps for Awareness Rigidity.

The answer is clearly “no”, we say, if the aim is to justify a very general and widely applicable norm for belief revision under growing awareness. As discussed in chapter 4.2.2, in the context of growing awareness, in particular, *planning* norms have limited viability. An agent cannot always

⁶Again, recall our discussion in chapter 4.2.2 of the different interpretations of norms of belief revision.

anticipate that she might experience awareness growth, let alone all the details of this awareness growth. So it would seem that she cannot, at least not always, specify a *plan* for belief revision under growing awareness.

We are yet to fully investigate circumstances in which awareness growth *is* anticipated, however. In these special circumstances, there may be a sense in which an agent can specify a plan for belief revision. One might wonder whether some sort of Dutch book argument could be made in favour of some norm for belief revision when it comes to these special circumstances. That is the question to which we turn in chapter 7, after having introduced anticipated awareness growth in the next chapter.

6

Anticipating Awareness Growth

6.1 Introduction

In previous chapters, we introduced the general idea of limited awareness and changes in awareness, and explained why these are both common and practically important. We have moreover proposed a general model of limited awareness and awareness growth. Finally, we have argued, against most of the previous literature on limited awareness, that there are no general norms of belief revision for growing awareness; indeed, the notion of conservative belief change is not well defined when it comes to awareness growth.

For the purposes of *decision-making*, however, what is of interest is not awareness growth in general, but rather, the *anticipation* of awareness growth (whether or not it actually comes to pass). If you are completely “unaware of your unawareness” in some choice situation, that is, if you do not even realise that you *could be* unaware of something that might turn out to be important for the outcome of your decision, then this lack of awareness will not play any role in your deliberation. In contrast, if you anticipate awareness growth—or, as it is often put, if you find yourself in a situation of “conscious unawareness” (Walker and Dietz 2011)—then your lack of awareness might, and arguably should, play a role in your deliberation.

Coming back to table 2.2 in chapter 2.3, that is, our characterisation of different types of learning or feedback from the world, anticipated aware-

ness growth can be identified with the “planned-unfamiliar” category (i.e., the upper right cell). Recall that what is learnt in these cases is unfamiliar, in the sense that it does not correspond to a property that one had been aware of before the learning event. But the learning is “planned”, in the sense that one had anticipated it in advance.¹

Situations where one anticipates awareness growth that may be relevant to one’s choices are not uncommon. Intuitively, it would be reasonable to anticipate or at least be open to the possibility of awareness growth in *novel* situations—those in which one’s own action is unprecedented, or else one may otherwise experience the world in a new way. Moreover, one’s past experiences can serve as a guide to novelty. One may come to recognise situations that are in some important respects similar² to situations in which one has previously experienced awareness growth.

It would also seem reasonable to be open to the possibility of awareness growth in decision-situations that exceed some specified level of *complexity*. Again, one’s past experience can serve as a guide: If you have experienced awareness growth in the past in choice situations that exceed some level of complexity, then it would seem reasonable to at least be open to the possibility of awareness growth whenever you find yourself in a choice situation that exceeds that level of complexity.³

Radical technological change would seem to be a prime example of a novel (and complex) situation where we should anticipate awareness growth. We see now, with the benefit of hindsight, that early industrialists were, for instance, unaware of the possibility that their employment of new manufacturing and transportation techniques would eventually change the climate. Other similar scenarios lead us to think that deployment of a new technology is generally prone to causing awareness growth that turns out

¹Any anticipated awareness growth may turn out not to occur (as noted above). Nonetheless, we will often talk as if anticipated awareness does occur as anticipated. That is also a supposition of table 2.2. While one may plan for awareness growth that turns out not to occur, just as one may plan for an ordinary learning experience that turns out not to occur, the realisation of failed plans is not represented in table 2.2. This omission is purely for reasons of convenience and simplicity in presentation.

²Although the notion of similarity to past decision situations might seem elusive, and perhaps ill-defined, a formal and sophisticated decision theory based on this notion has been developed: Gilboa and Schmeidler (1995, 2001).

³This point is made by Grant and Quiggin (2013b), who consider more precise measures of complexity and develop a game theoretic model with unawareness and (what we call) the anticipation of awareness growth.

to be relevant to the choice at hand. But there will be more ordinary sorts of situations, too, where one may reasonably anticipate awareness growth. For instance, one might anticipate becoming aware of new artists or genres when one visits an arts festival—an example to which we shall later return.

In this chapter we however foreground the more dramatic or extreme cases of anticipated awareness growth, such as the prospect of geoengineering that we introduce in section 6.2. We use primarily this example to further illustrate the anticipation of awareness growth and its role in decision-making. We go on to propose, in section 6.3, how to model an agent who anticipates awareness growth, such that this anticipation may play a role in her decision making. Finally, in section 6.4, we reflect on how our model of anticipated awareness growth reveals this phenomenon to be in some ways very ordinary, more or less akin to ordinary reasoning about less-than-fully-specified contingencies as described in any standard decision model.

6.2 Example: Solar Radiation Management

Solar radiation management (SRM), also known as solar geoengineering, has been discussed as a most likely very efficient and effective way of combating (and potentially even reversing) climate change. One such technique would consist in injecting reflective aerosol particles into the stratosphere. These particles would then reflect a small amount of inbound sunlight back out into space, thus making the planet cooler than it would otherwise be.

Although the idea behind SRM is partly inspired by the eruption of large volcanoes that naturally blast reflective sulphate particles into the stratosphere, nothing like the scale of SRM that would be needed to combat climate change has ever been tried. Thus, as for instance the Solar Radiation Management Governance Initiative frankly admits, the potential “side effects” of SRM are not well understood and are in fact mostly unknown.⁴

SRM would thus seem to be an example where we do, and should, anticipate awareness growth, if (or when) it is seriously tried. There may be various possible “side effects” which we cannot yet articulate, but we think one of these side effects is bound to occur if we do try SRM on the scale needed to combat climate change. So none of the possible outcomes

⁴See <http://www.srmgi.org/what-is-srm/>.

of SRM can currently be described in perfect detail. That is to say, we may anticipate awareness growth by *refinement*.

In addition (or alternatively), we may suspect there is some unknown contingency that is inconsistent with any contingency of which we are aware, and that could make solar radiation management either much more positive, or much more negative, than the possible outcomes of which we are currently aware. That is to say, we may anticipate awareness growth by *expansion*.

Despite SRM being an example where we presumably anticipate awareness growth, we may nonetheless have some rough estimate of the desirability of SRM. For instance, although we recognise that things could go wrong in ways we have not yet considered, we might think that the expected value or utility is positive (when SRM is compared to not employing SRM). More generally, the fact that we anticipate awareness growth if we perform some action need not entail that we cannot evaluate the action—and compare it to its alternatives—in terms of (at least rough, or imprecise) expected utility.

It might seem puzzling that we can evaluate an option, by estimating its expected utility, even when we take the option to be associated with potential awareness growth. We have more to say about this in the following sections, but let us here offer some preliminary remarks. To begin with, it helps to see the more dramatic cases of awareness growth at one end of a spectrum that also includes, at the other end, the less dramatic, ordinary cases of awareness growth.

Part of the appeal of going to a music or film festival, for instance, is the prospect of finding out about new artists or genres, that is, the possibility of *becoming aware* of artists or genres the existence of which one had previously been unaware. So, we may hope for some abstract outcome, or experience, the content of which we cannot quite articulate yet. Still, it seems evident that we can and do compare the option of going to a film festival, say, with the option of staying home, despite there being important aspects of the film festival of which we realise we are currently unaware.⁵ The same arguably

⁵Since discovering a new film genre might be “transformative”, for instance in the (epistemic) sense that one cannot know what experiencing films of that genre is like before having had the experience, our claim that one can compare the option of going to a film festival with the option of staying home is in contrast with Paul’s (2014) influential work on “transformative experience”. This is not the place to discuss Paul’s argument in detail,

applies to the evaluation of SRM, despite it being a more extreme case of anticipated awareness growth.

Nevertheless, it might seem reckless, especially in extreme cases of anticipated awareness growth, to base one's choices entirely on considerations of expected utility. For instance, even if, conditional on the contingencies of which we are currently aware, SRM has positive expected utility, one might wonder whether that suffices to justify a choice to try SRM on, say, a global scale. After all, since we anticipate awareness growth, we are open to the possibility that SRM will go wrong in ways that we haven't yet considered. In particular, it might seem that risk averse agents should take such possibilities to be reasons against basing decisions purely on considerations of expected utility maximization.⁶

However, it is unclear whether we should take such caution to be inconsistent with expected utility maximisation. In fact, the model we shall soon introduce makes such caution consistent with expected utility maximisation. In particular, the model allows for maximising expected utility *all-things-considered* while turning down an option that has positive expected utility *given the contingencies of which one is aware*.

6.3 Modelling anticipated awareness growth

Previously we argued that when modelling an agent's limited state of awareness, from her own perspective, it is not always apt to use a "catch-all" to represent that which she is unaware of. In particular, if the agent is "unaware of her unawareness", and thus does not anticipate any growth in awareness, then in so far as we are trying to model the agent's deliberation, we should resist that modelling choice. The upshot is that the experience of awareness growth cannot always be modelled as the refinement of a catch-all or some other proposition(s).

but in a nutshell, our response is that even if one cannot fully know what it is like to, say, experience some art genre before having experienced it, that does not rule out the possibility of evaluating the choice-worthiness of options involving that experience by, for instance, reasoning in accordance with a decision-model like that discussed below. (For a similar response to Paul's argument, see Bykvist and Stefánsson 2017).

⁶Similarly, one might speculate that risk-averse agents—who are, to put it roughly, particularly worried about the unknown—might place a higher value on increasing their awareness than those who are less risk averse. In fact, Quiggin (2016) proves that within his framework, the expected value (to an agent) of gaining awareness is greater the more risk averse is the agent.

But that is not to say that it never makes sense to appeal to a catch-all proposition to describe an agent’s reasoning. If we are trying to model an agent who anticipates awareness growth (by expansion), then it does seem natural to use some sort of catch-all to model that which the agent takes herself to be currently unaware of. However, as the reader may recall from chapter 3, by “catch-all” we (unlike some philosophers whose work we discussed) do not here mean strictly speaking *all* possibilities of which the agent is unaware. That is not something that can sensibly feature in an agent’s reasoning, or so we argued. Rather, we take the catch-all in question to be some abstract proposition standing in for a broad class of contingencies that the agent thinks she may later be in a position to concretize. For clarity, we will refer to the kind of proposition we have in mind as a “subjective catch-all”.

It may be useful to start with a standard decision-model (in which there is no anticipated awareness growth), against which we contrast a model of an agent who is aware of their potential lack of awareness and how it may grow. The matrix in table 6.1 is such a model. The agent assumes there are n states of the world (from s_1 to s_n), that determine the outcome of the available acts, i.e., the objects of choice, f and g . Finally, $f(s_i)$ is the outcome that obtains when f is chosen (or performed) and s_i is the actual (or true) state of the world.

	s_1	...	s_n
f	$f(s_1)$...	$f(s_n)$
g	$g(s_1)$...	$g(s_n)$

Table 6.1: Decision with no anticipated awareness growth

Since both the states of the world and the outcomes in this model are thought to be maximally specific in all ways that might be relevant to the decision—as is usually assumed—and since the states s_1 to s_n are thought to exhaust the set of possible contingencies—as is also usually assumed—there is no anticipated awareness growth. For instance, if each outcome $f(s_i)$ is thought to contain no uncertainty whatsoever—which in turn means that each state s_i is taken to be a maximally specific description of ways in which the world might be (prior to the choice between f and g)—and if the agent is in addition certain that one of outcomes $f(s_1)$ to $f(s_n)$ will be realised by the choice of f , then an agent who is appropriately modelled with a matrix like

the one in table 6.1 takes herself to be fully aware (in this decision situation).

	E_1	...	E_n	??
f	$f(E_1)$...	$f(E_n)$	$f(??)$
g	$g(E_1)$...	$g(E_n)$	$g(??)$

Table 6.2: Decision with anticipated awareness growth

Consider now the matrix in table 6.2. We focus initially on the anticipated awareness growth by *refinement* that is represented here. Suppose that E_1 to E_n is the finest and most exhaustive “partition” of the space of contingencies that the agent of interest can come up with. (We see below that it is not really a partition, according to the agent.) Each E_i is not taken to be a (fully specified) state of the world, but rather an *event* that may not fully determine the outcome of the two available acts. So, for instance, $f(E_i)$ leaves open what precisely are the concrete outcomes that may arise. The agent recognises this, let us assume, but does not know how to further partition the E_i into elements that leave no room for unawareness. In other words, the agent realises that she is (or, at least, she takes herself to be) unaware of some contingencies that are consistent with each E_i but which, if they materialise, deliver different outcomes for the two acts.

Now we turn to the anticipated awareness growth by *expansion* that is represented in table 6.2. We see from the “??” column in the table that the agent does not regard E_1 to E_n to be an *exhaustive* set of mutually exclusive events. In other words, the agent is aware of the fact that there might be contingencies that are inconsistent with all of E_1 to E_n and which would, if they materialise, determine the outcome of the two available acts. These further contingencies, that the agent cannot properly articulate, are represented in the table by “??”, which stands for a subjective catch-all. Note that, while we call this anticipated awareness growth by “expansion”, formally speaking, the agent here anticipates *refining* her catch-all. There is a sense in which she cannot truly anticipate awareness growth by expansion.⁷

One reason it is important to model anticipated awareness growth in the way just discussed is psychological realism. That has been our focus so far in this discussion. If the agent anticipates (or at least is open to

⁷Recall that in chapter 3.4 we mentioned various models in the literature that treat all cases of awareness growth as awareness growth that was anticipated. As per our discussion here, these models appeal to either a “catch-all” or else to otherwise coarse events that the agent expects will be later refined or made more concrete.

the possibility of) awareness growth, then a decision model in which her “states” are somewhat crude, and/or in which there is a variable or catch-all representing contingencies she is not fully aware of, is consistent with how she herself sees her epistemic predicament. (The same is not true when an agent is completely unaware, that is, when she is not even aware of her own potential limitations as far as awareness is concerned.) But it is important then that any such catch-all is a subjective one. No psychological realism is gained by adding propositions to the model that go beyond the abstract contingencies that the agent herself has gotten the whiff of.

Another reason for pursuing our model of anticipated awareness growth is that it will often be important for an agent’s reasoning about what to do. Consider again the solar radiation management example and suppose that a policy-maker’s epistemic situation, including her anticipated awareness growth, when she deliberates about whether to try SRM, is represented by a table like 6.2. We can read the options f and g as doing and refraining from SRM respectively.

First, let us focus on anticipated awareness growth by expansion, represented by the subjective catch-all, “??”, which in some sense is easier for a modeller to get a handle on. Suppose that the policy-maker reasons that if any of the events E_1 to E_n obtain, then she would prefer implementing SRM rather than abstaining. Still, because the policy-maker anticipates, or is at least open to the possibility of, awareness growth by expansion if SRM is implemented (as represented by the subjective catch-all), she finds SRM to be so risky that she prefers to abstain all things considered. This is to say that the utility of SRM given the subjective catch-all is sufficiently negative and the catch-all is sufficiently probable (where both the utility and the probability are the policy-maker’s) for the overall expected utility of SRM to be lower than the alternative, according to her.

To make the above claim more precise, suppose that the policy-maker’s preference relation *conditional on* events E_1 to E_n ⁸ satisfies the appropriate coherence constraints, such that this conditional preference relation can be represented as maximising expected utility (recall our discussion in chapter 1.2.2 of representation theorems, and our later mention in chapter 1.3.2 of

⁸Piermont (2017) calls such conditional preferences “contingency plans”, and uses them to formally characterize anticipated awareness growth (or “introspective unawareness”, to use his term).

Karni and Vierø's (2017) representation theorem for anticipated awareness growth). Then we can, in this case, infer from her preferences that the expected utility of implementing SRM, given the catch-all, Ω , is sufficiently negative, compared to the expected utility of not implementing SRM given Ω , such that SRM is not worth the risk all-things-considered. In contrast, had the decision-maker preferred to implement SRM, then, given that she prefers SRM over its alternative conditional on events E_1 to E_n , we would instead have inferred that the expected utility of implementing SRM given Ω is *not* sufficiently negative compared to the expected utility of not implementing SRM given Ω . More generally, when we use a model that includes a subjective catch-all, the assumptions of standard decision theory for full awareness (supposing the contingencies of which she is aware) allow us to say at least something about the policy-maker's attitudes to that which she suspects she currently has limited awareness.

A method like this cannot, however, be used to estimate the policy-maker's attitudes to contingencies that she anticipates discovering due to what we called a *complete* (as opposed to partial) refinement. Recall that we assumed that the events E_1 to E_n are the most fine-grained contingencies that the policy-maker can come up with; however, she suspects that these contingencies can all be refined according to some property of which she is currently unaware. Hence, there is no possibility of considering the policy-maker's preferences conditional on what she takes to be fully specified contingencies to determine her attitude to that of which she is yet to become aware.

In contrast, if the policy-maker anticipates a *partial* refinement—which we have so far mostly set aside for reasons of simplicity—then a method very similar to that discussed above to elicit the attitude to anticipated expansion could be used for anticipated refinement. Let us focus on some given event E_i that the agent anticipates will be later refined. For our SRM example, this event may be “SRM causes severe smog”. Now it may be that the policy-maker is unsure of all of the maximally specific ways in which this event may be true. But she may comprehend some of the ways it may be true. For instance, she may regard one of the maximally specific states constituting E_i to be “SRM causes severe smog but otherwise status quo (with respect to natural systems functioning)” (call this E_{i1}). Assume that she thinks there is some other way that E_i might be true that she cannot quite

articulate; it may be described as “SRM causes severe smog and otherwise not the status quo” (call this E_{i2}). This latter instantiation of E_i is like a local catch-all; that is, it acts as a catch-all within the partition of outcomes that comprise E_i . Then the comparison of the policy-maker’s preference between f and g conditional on E_i with her preference between f and g conditional on E_{i1} —and similarly the comparison of the policy-maker’s preference between f and g conditional on E_i with her preference between f and g conditional on E_{i2} —will reveal how her optimism or pessimism about the specific outcomes associated with E_i , which she currently cannot articulate, affects her comparison of the two acts, f and g . Moreover, if the policy-maker’s preferences satisfy the appropriate constraints (recall again our discussion of representation theorems), then we can even find precisely how (un)favourable she expects to be the outcomes associated with the local catch-all event E_{i2} .

6.4 Not such extraordinary reasoning?

The reader might wonder whether our model of anticipated awareness growth reveals that it is not in fact a hitherto neglected aspect of reasoning. The thought might be that our model described above is no different from a standard model of reasoning. So either anticipated awareness growth is unremarkable or else it is something that has been accommodated in decision models all along. While we are, to a great extent, sympathetic to this line, we see the similarities between anticipated awareness growth and “ordinary” reasoning—to be discussed in what follows—as helpful for better understanding the former, rather than reason to dismiss it.

First, let us consider the extent of similarity. Given our favoured (subjective) interpretation of catch-alls, they are, in many ways, just like any ordinary proposition that we use to model agents’ practical reasoning. In general, the propositions that are thought to feature in an agent’s reasoning are quite abstract and lacking in details. In deciding whether to ride my bike or else drive to work, for instance, I may not anticipate any awareness growth and yet nonetheless appeal to rather abstract propositions, like “some unusual road incident occurs”. That is, I do not dwell on the specific ways in which an unusual road incident may occur (whether it be, for instance, a collision between cars or a flock of swooping magpie birds

harassing cyclists). I can form a preference for driving rather than riding given merely the abstractly-described circumstance, and I can also assign some probability to said circumstance occurring. Indeed, upon reflecting on any ordinary case of reasoning, one might surmise that, when it comes to abstraction, the difference between a “catch-all” and other propositions that ordinarily feature in decision models is one of degree rather than kind.

Similarly, there seems to be much in common between, on the one hand, the coarse-grained events we introduced to model agents who anticipate awareness growth by refinement, and, on the other hand, the maximally specific states of the world in traditional decision models. After all, the states in traditional decision models are themselves abstract to varying degrees; they are only specific enough to account for everything that matters to the agent in comparing the options available to her. Coming back to my decision about whether to drive or ride to work: I do not consider, for instance, whether or not my neighbour has her breakfast before 9am, or whether or not the school children will be in the mood to wave to me. That is because these finer details about how the world might be do not matter to my decision, given my goals and values.

That said, there is a difference worth acknowledging: the *reasons for abstraction* differ when it comes to modelling anticipated awareness growth compared to modelling “ordinary” reasoning, i.e., reasoning that does not involve anticipated awareness growth. In the former case, the agent suspects that she omits contingencies in the course of her reasoning that she will later recognise to be relevant to her decision but which, regrettably, she is currently unaware of. In the latter case, while the agent may recognise that she omits contingencies in the course of her reasoning, she does not consider this regrettable since she does not regard these contingencies to be relevant to her decision (and if she later comes to think that these contingencies are relevant, then she can simply re-evaluate her decision in light of them). Nor does she expect to become aware of any other contingencies that turn out to be relevant to her decision.

The question is what hangs on this interpretative difference. Apparently not so much, since it seems that any given decision model, like that in table 6.2, could be read either way. While we have described table 6.2 as a model of anticipated awareness growth, it may otherwise serve as a model of an agent’s reasoning where there is no anticipated awareness growth. It

often makes sense for a model to include a highly abstract “catch-all” or a coarse-grained partition of events, even though the agent neither anticipates awareness growth by refinement nor by expansion. And this is true even when modelling the reasoning, or “internal perspective”, of an agent, which is the perspective on which we have focused.

One can see this by reflecting on our earlier example of deciding whether to drive or ride to work. Other examples are in this regard even more vivid. For instance, when deciding how to invest one’s pension, one need not take detailed account of extreme events—such as a meteorite striking the earth and killing all of humanity—that would affect each investment in the same way. One way to model this reasoning would be to include a “catch-all”, which in this case would be interpreted as the set of contingencies that yield the same outcome for all the pension alternatives. Continuing with this example, one also need not, when making this investment choice, consider all possible percentage points by which interest rates might change; instead, one would presumably just consider coarse-grained events such as “low”, “moderate”, and “high” interest rates.

Intuitively, one does not want to say in this case that the decision-maker realises she is currently *unaware* of relevant contingencies that she may later come to be aware of. Quite the contrary. It is her considered judgement that the further details of the contingencies in question are not worth dwelling on since they will not affect her choice. However, if she were to come to believe that, say, a meteorite strike could affect the various investments differently, then she could (and should) include this in her reasoning, unlike someone who is truly unaware of this possibility.⁹

We can also use table 6.2 to model an agent who suspects that *something* that she deliberately excluded from her deliberation should in fact be included, while not being able to specify further what she wrongly excluded. In that case, the agent’s predicament will be much like someone’s who anticipates awareness growth. In fact, there is an intuitive sense in which this agent does anticipate awareness growth: she anticipates becoming aware of the fact that something that she took to be irrelevant to her deliberation in fact is relevant.¹⁰

⁹Similarly, if the agent were to come to think that, say, the difference between a 0.02 and 0.021 interest rate is of importance to her decision, then she could factor this into her reasoning, unlike someone who is truly unaware of this possible refinement.

¹⁰Karni and Vierø (2017: 317) make a similar observation about their model of anticipated

Finally, we said above that the similarities between anticipated awareness growth and ordinary reasoning may be instructive. For starters, the similarities provide further grounds for the proposal of the last section that agents who anticipate awareness growth can maximise expected utility just like ordinary reasoners. Furthermore, we have grounds to think that the credences and utilities underlying these expected utility evaluations are largely based on the same sorts of considerations, whether the case involves anticipated awareness growth or not. With respect to the SRM example, for instance, that the policy maker anticipates later becoming more aware of the contingencies represented by the catch-all does not mean that she cannot evaluate the probability and utility of the catch-all based on what she is currently aware of, sketchy as that may be. For instance, the policy maker may judge the utility of the catch-all to be negative, based on the fact that in the past the unforeseen outcomes of major technological change have been bad sorts of disruptions to existing natural and social systems. This was the sort of reasoning we described in the introduction to this chapter.

By way of contrast, note that some take the evaluation of utilities and probabilities in cases of anticipated awareness growth to be somewhat problematic or else highly idiosyncratic. Laurie Paul (2014), for instance, is sceptical about an agent's ability to evaluate contingencies that she herself takes herself to be unacquainted with (recall footnote 5), and says that to the extent the agent can do so, all she can go on is the value of some general experience of discovery (the "revelatory value", as she calls it). While not similarly sceptical about the evaluation of yet-to-be-articulated contingencies, Karni and Vierø (2017) seem to agree on the last point, suggesting that the agent's evaluation of a catch-all is to be interpreted as how much she generally likes being surprised.¹¹

We resist such exceptional treatment of anticipated awareness growth. To be sure, when an agent anticipates awareness growth, this has some bearing on the bases for her belief and desire judgments. For instance, her general attitude towards surprise plausibly plays a greater role than usual (as we suggested with our earlier arts festival example). Moreover, the

awareness growth.

¹¹Moreover, Grant and Quiggin (2013a, 2013b) argue that, when it comes to anticipated awareness growth, probability estimates are necessarily based on *induction* and options are evaluated by *heuristics*, which differ from the standard (deductive and calculative) methods assumed for fully aware agents.

agent will have different sorts of projections about her attitudes in the future. Precisely how these projections of her more aware future self can and should bear on her present attitudes is the topic of the next chapter. For now we emphasise that the reasoning of an agent who anticipates awareness growth has much in common with what we have called “ordinary” reasoning, i.e., the reasoning of an agent who does not anticipate awareness growth.

6.5 Concluding remarks on chapter 6

We have now explained in informal terms what it means to anticipate awareness growth, and illustrated why such anticipation is both common and important. We have moreover suggested a formal way of accounting for such anticipation within a decision model. We appealed to abstract propositions, whether a “subjective catch-all” in the case of anticipated expansion, or else a coarse-grained “partition” of events in the case of anticipated refinement. Moreover, we considered why anticipated awareness growth, thus understood, is in some ways unremarkable. We have however not yet discussed in detail how such anticipation, in particular, an agent’s projections of her more aware future self, should (or should not) affect her current credences and preferences. That is the topic to which we turn next.

7

Awareness Reflection

7.1 Introduction

In chapters 4 and 5 we considered the dynamics of awareness growth. In particular, we considered what changes in credence are rational when an agent with limited awareness experiences awareness growth (or more minimally, what changes in credence are consistent with the agent being, epistemically speaking, “stable” or the same agent). We argued that more or less anything goes. There is no general requirement of rationality dictating how you should change your beliefs upon becoming aware of new contingencies. (Likewise, there is no straightforward way to detect whether your belief change makes you a different person, epistemically speaking, upon becoming aware of new contingencies.)

In the last chapter we however turned our attention to a special case of limited awareness, namely, when one anticipates growth in awareness or more generally suspects that there is something of which one is unaware. As we saw, this special case of limited awareness is both common and tremendously important from a practical point of view. While not itself a dynamic phenomenon, anticipated awareness growth is intimately related to the dynamics of belief. These are cases where an agent predicts (rightly or wrongly) that she is about to experience awareness growth.

In this chapter we continue examining anticipated awareness growth, and ask whether there is a norm of rationality that constrains an agent’s credences in these special circumstances. To be clear, this would not be a *diachronic* norm, governing *change* in credence, but rather a *synchronic*

norm, governing an agent's credences *at a given time*. We consider both an informal and a formal argument for such a norm, which seems to be very constraining indeed: When you anticipate awareness growth, your current credences should match your expected¹ future credences in the event that you experience awareness growth. Moreover, there is arguably an equally constraining norm on preference: When you anticipate awareness growth, your current preferences should match your predicted future preferences in the event that you experience awareness growth.²

7.2 Informal argument

We will focus on a very simple decision problem to try to get clear on what (if anything) rationality requires of an agent who anticipates awareness growth. Suppose, for instance, that you are trying to decide whether you should go to the beach rather than stay at home and finish the latest Netflix series that you have been binging on. The pleasantness of going to the beach is highly sensitive to what the weather will be like, which you are uncertain of. (Let us imagine that you do not have access to a trustworthy weather forecast.) Staying at home is the risk-free option, since you take yourself to know that you will have a relaxing day on the couch no matter what the weather will be like.

Now, suppose that the only weather conditions that you consider are *sunny* and *clouded*. You suspect, however, that there is some weather condition that you have left out, which may affect the outcome of your decision to go to the beach. (We, the modellers, may see that this weather condition is *misty*.) Your decision problem is represented by table 7.1.³

	sunny	clouded	??
Beach	Beach & sunny	Beach & clouded	Beach & ??
Home	Home & sunny	Home & clouded	Home & ??

Table 7.1: Beach or Home

¹As will become apparent, we are using “expected” here in the technical (mathematical) sense. More precisely, the norm requires that the mathematical expectation of your more aware credences, calculated relative to your own prediction about your future credences, should be the same as your less aware credences.

²We thank Michael Nielsen for encouraging us to consider “reflection principles” in relation to growing awareness, and thereby inspiring our investigations in this chapter.

³The example is inspired by Bradley (2017: 254).

To make the example a little more concrete, suppose that your credence that it will be sunny is 0.4, which is also your credence that it will be clouded. Your credence in the (subjective) catch-all weather contingency, denoted ‘??’ in table 7.1, is 0.2. Now we can ask: Is this probability, 0.2, indicative of (or constrained by) your predicted future credences? Conversely: Once you manage to fill in the catch-all, how do you expect that this will change your credences in all three weather contingencies? In other words, now when you are unaware of the content of the catch-all, what probability do you expect that you will assign it when you become aware of its content, and what probabilities do you expect you will then assign sunny and clouded?

One natural answer would be that you should not expect the above growth in awareness to yield any change, positive or negative, in your credences. You may of course believe that this awareness growth *could* affect your credences in some direction. For instance, you may think that you *could* become more confident of the catch-all, but you should then also believe that this awareness growth *could* affect your credences in the “other direction”, such that you are less confident in the catch-all. More precisely, the claim is that there should be no *expected* change of credence after awareness growth, where the expectation is based entirely on your own prediction about your future epistemic state.

What considerations support the above claim? Informally, one could argue for the above claim by noting that if you *did* expect your credence in, say, the catch-all, to change one way or another once you become aware of its content, then you should now change your degrees of belief in that same direction—assuming (as seems reasonable in this case) that you do not predict that you would be any less rational, or less informed, if your awareness were to grow. For instance, suppose that you expect that if you were to become aware of what the catch-all consists in, you would become more confident than you currently are that it will neither be sunny nor clouded. Then it would seem that you should revise upward your current degree of belief that it will neither be sunny nor clouded. You should, as it were, defer to your more aware self. For why would you not trust someone who is exactly like you except more aware than you are?

Now, it could of course be that when it comes to some cases of anticipated awareness growth, one predicts that the person one will become if one’s awareness grows is sufficiently dissimilar to one’s current self that one is

not willing to treat that person as someone exactly like oneself except more aware. In that case, one predicts that one will not be the same epistemic agent before and after the awareness growth. (As noted above, one way to read our conclusion in chapters 4 and 5 is that there is no general way to discern, just by examining her credence change, whether an agent fails to be the same epistemic agent after a growth in awareness. Any such prediction must therefore draw on further insights, beyond the credence change itself.) Imagine for instance that one predicts the experience of going to an arts festival to be so “epistemically transformative” (Paul 2014) that the belief system that one will have after the experience differs from one’s current system in some fundamental ways. In that case, it seems that rationality does not require that one defer to one’s future self, so to speak, and thereby match one’s credences to one’s expected future credences.

Although we do not want to rule out the possibility of awareness growth being transformative, in the sense of turning you into a different agent, we shall set such experiences aside for now, and instead focus on what arguably are more typical examples of anticipated awareness growth. In fact, the principles we discuss below explicitly stipulate the awareness growth to which they apply to not change who you are, since they only apply to events where the *only* thing that happens is that you gain more awareness. Hence, these principles do not apply to events where you gain more awareness *and* as a result become a different agent.

We can call the principle that the above considerations about the beach-or-home example seem to support *Awareness Reflection*, after the traditional Reflection principle (van Fraassen 1984).⁴

Awareness Reflection (Informal version). *For any awareness context and any proposition A (in that context), if you predict that between now and time t, the only thing that happens is that you gain more awareness, then your current credence in A should equal your currently expected credence in A for time t (if the latter is well-defined). Conversely, your currently expected credence in A for time t should (if well-defined) equal your current credence in A.*

The qualification that the relevant values be well-defined is due to the

⁴Note that even if Awareness Reflection is true, it need not, in any sense, be a fundamental norm of epistemic rationality. Rather, it may simply be an instance of a more general norm of deference to those who are experts relative to your current self. In fact, Brian Hedden (2015) argues that this is true of epistemic reflection principles in general.

fact that the informal arguments above only seem to establish that there is something strange with a person who currently has a credence for A that differs from her currently well-defined expectation of credence at time t . We get back to this qualification in section 7.3.2, when discussing a formal argument for Awareness Reflection.

In addition to it seeming odd, from a purely epistemic perspective, to violate Awareness Reflection, one might suspect that violating the principle can be undesirable from a purely practical point of view. After all, if you violate Awareness Reflection, then you expect that what you are now willing to pay for some bet is more than what you will find the bet to be worth if you gain more awareness. For instance, if you expect that, once you become aware of the content of the catch-all weather contingency, you will be less confident in it being sunny than you currently are, then what you are now willing to bet on it being sunny is more than what you expect you will be willing to bet when your awareness grows. One might suspect that a clever bookie could exploit this discrepancy. That is indeed the case: A so-called *Dutch book strategy* can be employed against you if you violate Awareness Reflection, as we shall soon see.

7.3 Formal argument

We now consider a formal argument that supports the informal considerations and conclusions of the last section. We start by formalising the norm, Awareness Reflection, that we introduced above, and we then propose a so-called “Dutch book argument” for this norm.

7.3.1 Awareness Reflection formalised

As usual, we will use P to denote your less aware credences. What matters in the case of anticipated awareness growth is what you predict your credences to be in the future, having experienced awareness growth. Let \mathbf{P}^+ denote the *proposition* that your credences in your more aware state can be represented by P^+ .⁵ Assume for now that you do indeed make such a prediction—that

⁵ The supposed domain of this probability function, P^+ , is simply your current algebra of propositions, which we earlier denoted \mathcal{F} . Your anticipated awareness growth (from \mathcal{F} to \mathcal{F}^+) amounts, formally, to a *refinement* of your current set of possibilities including the subjective catch-all. As such, your credences over the propositions in the “coarse” algebra,

you entertain the proposition \mathbf{P}^+ in your current awareness context. In fact, in line with our comments above, it is important for what follows that the proposition in question, \mathbf{P}^+ , is slightly more complex: it denotes that your credences in your more aware state can be represented by P^+ and you are otherwise just as informed (and just as rational) as you are now when your credences are represented by P . More generally, you are the same epistemic agent, just more aware.

As a first attempt, we might try to formally present the principle we have been discussing and will now investigate further as:

Awareness Reflection (Formal version). *For any awareness context X and any proposition $A \in X$, and for any rational P and P^+ :*

$$P(A \mid \mathbf{P}^+) = P^+(A)$$

Informally, the principle says that the degree to which you should now believe A , given that you will in your more aware state believe A to some particular degree, say r , is that degree, r . (In the next section we consider a weakening of this principle.)

Note that the above (conditional) formulation of Awareness Reflection is, under special conditions, logically equivalent to an expectational version that might on the face of it seem to better capture our informal discussion of responding to your expectation about your more aware degrees of belief. Another virtue of the expectation version is that, unlike Awareness Reflection, it does not give the impression of lacking actual-credence guidance. Faced with Awareness Reflection, one might wonder how one can use that norm to guide one's credence, given that one will never become certain that one will have any particular credence function in a more aware state; hence, one will never be in a position to conditionalise on a proposition like \mathbf{P}^+ .

The special conditions mentioned above that are crucial for the logical equivalence claim can be spelled out as follows: you are *certain* you will be the same agent after awareness growth,⁶ and are uncertain only about

\mathcal{F} , even once you become more aware, should satisfy the probability calculus.

⁶This is in line with Briggs (2009), who argues for a qualification of the traditional Reflection principle which can be put as follows: for reflection principles to be generally plausible, we need to assume that the agent, when their credence is represented by P , is *certain* that they are the same agent as the one that will be represented by P^+ . (Cf. one of the interpretations of Conditionalization that we suggested in chapter 4.2.2—that this pattern of belief revision is constitutive of being the same agent upon revising one's credences).

which of a number of candidate credence functions will represent your more-aware credences. As such, there is some set of credence functions, call them P_1^+ to P_n^+ , that are the candidates for representing your more-aware credences. As before, \mathbf{P}_i^+ denotes the *proposition* that (you are the same agent and) your more-aware credences can be represented by P_i^+ . Since $\{\mathbf{P}_1^+, \dots, \mathbf{P}_n^+\}$ is a partition of the set of all possibilities⁷ (and assuming that each has positive probability), the *Law of total probability* entails that:

$$P(A) = \sum_{i=1}^n P(\mathbf{P}_i^+) \cdot P(A | \mathbf{P}_i^+)$$

But then Awareness Reflection implies that:

$$P(A) = \sum_{i=1}^n P(\mathbf{P}_i^+) \cdot P_i^+(A) \quad (7.1)$$

Less formally, Awareness Reflection implies that your current credence in A should equal your expected credence in A in your more aware state. Moreover, the expectational formula implies Awareness Reflection. Updating 7.1 on \mathbf{P}_j^+ gives us:

$$P(A | \mathbf{P}_j^+) = \sum_{i=1}^n P(\mathbf{P}_i^+ | \mathbf{P}_j^+) \cdot P_i^+(A | \mathbf{P}_j^+) \quad (7.2)$$

But if we then assume that one always is certain of one's own credence, that is, $P_j^+(\mathbf{P}_j^+) = 1$, which implies that $P_j^+(A | \mathbf{P}_j^+) = P_j^+(A)$, we get:

$$\sum_{i=1}^n P(\mathbf{P}_i^+ | \mathbf{P}_j^+) \cdot P_i^+(A | \mathbf{P}_j^+) = P_j^+(A) \quad (7.3)$$

Combining 7.2 and 7.3, we get Awareness Reflection:

$$P(A | \mathbf{P}_j^+) = P_j^+(A)$$

So, the conditional version of Awareness Reflection is logically equivalent (under the conditions discussed) to an expectational version, where the

⁷If, contrary to our stipulated assumption, you think it possible that you will *not* be the same agent after the growth in awareness, then $\{\mathbf{P}_1^+, \dots, \mathbf{P}_n^+\}$ is *not* a partition of the possibility space, as these propositions are not exhaustive.

latter is more pertinent to one's actual credences at a time, and is thus the focus of our informal considerations in section 7.2.

7.3.2 A Dutch book argument for Awareness Reflection

Return now to the simple (conditional) Awareness Reflection principle. As those familiar with the Dutch book argument for the traditional Reflection principle (due to van Fraassen 1984) might immediately recognise, a Dutch book strategy can be employed against you if you violate Awareness Reflection. A Dutch book strategy is a betting strategy that consists of bets that you consider individually fair, or acceptable, but which nevertheless together ensure that you lose. In other words, no bet in the strategy is unfavourable, as judged by your own degrees of belief, but together the bets ensure your loss. An important premise in so-called Dutch book *arguments* is that being vulnerable to a sure loss is a sign of irrationality. Hence, if Dutch book arguments are generally valid, Awareness Reflection may be a requirement of rationality.⁸

Below we describe a Dutch book argument for Awareness Reflection.⁹ The bets have prices and prizes between 0 and 1. These numbers can, for instance, be interpreted as dollars, or units of well-being, or anything else that we can assume to be valued linearly (at least in the zero to one interval).

Suppose that you violate Awareness Reflection by being more confident of A than you expect you will be when your awareness grows; in particular, your current conditional degree of belief in A , given that you believe A to degree r in your more aware state, is some degree greater than r . More formally, $P(A \mid \mathbf{P}_j^+) > r$ even though $P_j^+(A) = r$; in violation of Awareness Reflection. First the bookie offers you the following three bets, each of which you accept, assuming that you use your degrees of belief or credences to evaluate bets by their expected value.

- Bet 1 costs you $P(A \& \mathbf{P}_j^+)$ and pays you 1 if $A \& \mathbf{P}_j^+$ is true but pays 0 otherwise.
- Bet 2 costs you $P(A \mid \mathbf{P}_j^+) \cdot P(\neg \mathbf{P}_j^+)$ and pays you $P(A \mid \mathbf{P}_j^+)$ if $\neg \mathbf{P}_j^+$ but pays 0 otherwise.

⁸For a more detailed discussion of Dutch book arguments, see Richard Pettigrew's forthcoming element in this series (2020).

⁹Our formulation is similar to Vineberg's (2011).

- Bet 3 costs you $(P(A | \mathbf{P}_j^+) - r) \cdot P(\mathbf{P}_j^+)$ and pays you $P(A | \mathbf{P}_j^+) - r$ if \mathbf{P}_j^+ but pays 0 otherwise.

Now, if $\neg\mathbf{P}_j^+$, then you win $P(A | \mathbf{P}_j^+)$ from Bet 2, which is the sum of what you paid for Bet 1 (which you have lost) and for Bet 2, and you have also lost Bet 3, for which you paid $(P(A | \mathbf{P}_j^+) - r) \cdot P(\mathbf{P}_j^+) > 0$; so, you are at a net loss.

However, if \mathbf{P}_j^+ , then you have won Bet 3, thus gained $P(A | \mathbf{P}_j^+) - r$, for which you paid $(P(A | \mathbf{P}_j^+) - r) \cdot P(\mathbf{P}_j^+)$; but you have lost Bet 2, for which you paid $P(A | \mathbf{P}_j^+) \cdot P(\neg\mathbf{P}_j^+)$. So, from bet 2 and 3 you are at a loss: the combined net outcome from these bets is $rP(\mathbf{P}_j^+) - r$ which, since both r and $P(\mathbf{P}_j^+)$ are between 0 and 1, is less than 0. But Bet 1 is not settled until the truth of A is known. What the bookie now does, is to buy from you a Bet 4 that pays him 1 if A but 0 otherwise, exploiting your new degrees of belief; that is, he offers a price of r for Bet 4, which you accept (again assuming that you use your degrees of belief to evaluate bets by their expected value). Then, if A is true, you win Bet 1 but lose this final bet, and the reverse is true if A is false; so, in either case, you end up with $P(\mathbf{P}_j^+)(r - P(A | \mathbf{P}_j^+))$. Thus, by the assumption that $P(A | \mathbf{P}_j^+) > r$, you are again at a net loss.

An analogous strategy could be used to exploit out if you had instead violated Awareness Reflection by $P(A | \mathbf{P}_j^+) < r$ even though $P_j^+(A) = r$.

In other words, whatever happens, you are sure to lose, and the bookie is sure to win, if you use your awareness-reflection-violating degrees of belief to decide which bets to accept. So, in so far as being vulnerable to sure loss, due to your degrees of belief, is a sign that our degrees of belief are irrational, we can conclude that it is irrational to violate Awareness Reflection.

We should note that the above Dutch book argument depends on the assumption that $P(A | \mathbf{P}_j^+)$ is defined, that is, that it takes *some* value. (Others have noted the corresponding assumption in the Dutch book argument for the traditional Reflection principle, e.g. Briggs 2009.) It will however be undefined if \mathbf{P}_j^+ has zero probability, according to the agent, and similarly if at least one of \mathbf{P}_j^+ and $A \& \mathbf{P}_j^+$ has *no* probability, according to the agent. Throughout the book we have been assuming, for reasons of simplicity, that an agent has precise probabilities in those propositions of which she is aware; hence, if \mathbf{P}_j^+ or $A \& \mathbf{P}_j^+$ has no probability, according to the agent, then that means that she is not aware of the proposition(s) in question.

The assumption that $P(A | \mathbf{P}_j^+)$ is well-defined is however far from being self-evident. Hence, we can either take the assumption as being part of the Dutch book argument, or we can weaken Awareness Reflection to:

Awareness Reflection (Formal, weaker version). *For any awareness context and any proposition A , and for any rational P and P^+ :*

$$P(A | \mathbf{P}^+) \not\geq P^+(A)$$

So we see that an agent is only subject to the threat of a Dutch book and thus the norm of Awareness Reflection (now assumed to refer to the stronger version) in those circumstances where she both entertains the various candidates for her future more-aware credences and is also certain she will be the same agent upon becoming more aware. But would these conditions ever plausibly hold? For one thing, in the context of awareness change, anticipated as it may be, the agent cannot even articulate the events that she may or may not come to learn, which differs from the traditional cases to which Dutch book arguments have been applied. Perhaps this fact could be used to argue that even if Dutch book arguments are valid when it comes to the ordinary Reflection principle and Conditionalization, the argument is not valid when it comes to Awareness Reflection. By way of response, however, one might point out that the agent need not be able to predict *what* she will become aware of in order to predict *how* this awareness growth may affect her credences in propositions of which she is already aware. The Dutch book argument above, and ultimately the norm of Awareness Reflection, is concerned just with the latter kind of prediction.

7.4 Awareness Reflection vs. Reverse Bayesianism

Although Awareness Reflection, when it applies, is in many ways a very strong and very conservative requirement, it turns out that an agent can satisfy the principle without satisfying, or predicting that she will satisfy, the conservative rule we examined in chapter 4: Reverse Bayesianism. Below we illustrate this fact by returning to the decision between staying at home and going to the beach, represented in table 7.1. We go on to discuss why this is further reason to doubt that Reverse Bayesianism is a general norm for belief revision under growing awareness.

Recall that we assumed that in your less aware state, you believe to degree 0.4 that it will be clouded and you also believe to degree 0.4 that it will be sunny. Now suppose that, in line with Awareness Reflection, you expect that, after awareness grows, you will believe each of these possibilities to degree 0.4. However, you believe this to be the case because you now (in your less aware state) believe that you will (in your more aware state) *either* believe to degree 0.2 that it will be sunny and to degree 0.6 that it will be clouded, *or* believe to degree 0.6 that it will be sunny and to degree 0.2 that it will be clouded, and you now (in your less aware state) find each of these possible future epistemic states to be equally likely and together exhaustive. More formally:

$$P(\text{sunny}) = P_1^+(\text{sunny}) \cdot P(\mathbf{P}_1^+) + P_2^+(\text{sunny}) \cdot P(\mathbf{P}_2^+) = (0.6)0.5 + (0.2)0.5 = 0.4$$

$$P(\text{clouded}) = P_1^+(\text{clouded}) \cdot P(\mathbf{P}_1^+) + P_2^+(\text{clouded}) \cdot P(\mathbf{P}_2^+) = (0.2)0.5 + (0.6)0.5 = 0.4$$

Note that in this case, while you satisfy Awareness Reflection, you believe that when awareness grows, you will either be three times more confident that it will be sunny than clouded, or three times more confident that it will be clouded than sunny. And since we are assuming that these are the only two future epistemic states that you consider possible, you are in fact *certain* that you will be three times more confident in one of these weather contingencies than the other. Before awareness grows, however, you are equally confident in these two weather contingencies. So, you are certain that you will violate Reverse Bayesianism, understood as a diachronic norm. In fact, you *do* violate Reverse Bayesianism, if it is rather understood as a “planning” norm (recall our discussion in chapter 4.2.2), since you predict or plan that your relative credence in clouded versus sunny weather will change one way or the other upon awareness growth, despite you being the same epistemic agent.

Note that Awareness Reflection is strictly logically weaker than (the planning version of) Reverse Bayesianism. If one plans or predicts that one’s credence change upon awareness growth will conform with Reverse Bayesianism, then one will also satisfy Awareness Reflection. That is because awareness growth that is anticipated is, formally speaking at least, awareness growth by refinement (we mention this in footnote 5 and also in

chapter 6.3). In cases of refinement, Reverse Bayesianism requires simply that all credences in propositions of which you were already aware (including here the subjective catch-all) stay constant. So if you satisfy (the planning version of) Reverse Bayesianism, you trivially satisfy Awareness Reflection.

While it is a logically weaker norm, in many cases the person who satisfies Awareness Reflection while predicting a credence change in violation of Reverse Bayesianism behaves like a person who predicts that they will not change their beliefs at all when awareness grows, in accordance with Reverse Bayesianism. For instance, suppose again that you satisfy Awareness Reflection by splitting your confidence between two different future epistemic states, as in the above example. Moreover, suppose that you want to base your choice, partly at least, on what you expect yourself to believe when you gain more awareness. Then if you are what we might call *uncertainty neutral with respect to your future beliefs*—in the sense that, in so far as you take your predictions about your future beliefs into account, you only consider your expected future degrees of belief—then you will act just like someone who predicts that their beliefs will not change at all when awareness grows, in accordance with Reverse Bayesianism.

However, a person who is sensitive to their own uncertainty about their future beliefs, will, even though they satisfy Awareness Reflection, often act quite differently from someone whose predictions accord with Reverse Bayesianism. For instance, if you are averse to uncertainty of this kind, then you might then not be willing to risk finding yourself in a situation where awareness grows just as you are arriving at the beach in a way that results in you becoming three times more confident that it will be clouded than that it will be sunny; and hence, you will now essentially act as though you were more confident that it will be clouded than that it will be sunny, even though you are actually now equally confident that it will be clouded as that it will be sunny. In contrast, someone whose predictions about their future beliefs accord with Reverse Bayesianism will not behave that way, since they are not uncertain about their future beliefs in clouded versus sunny.

So, Awareness Reflection is strictly logically weaker than (the planning version of) Reverse Bayesianism, and the two norms also have different behavioural implications. What should we conclude from this? We suggest

that this observation casts further doubt on Reverse Bayesianism being a general norm of rational belief change for growing awareness (however such a norm is interpreted). Awareness Reflection is highly constraining with respect to how one's predicted changes in credence upon awareness growth should relate to one's current credences. And yet this norm does not require that one's predicted changes in credence accord with Reverse Bayesianism. So the latter norm apparently goes out on a limb. Moreover, the traditional arguments do not seem to offer any support for this limb. For instance, the Dutch book argument proposed in the last section only secures (as far as its assumptions hold) the weaker norm, Awareness Reflection.

7.5 Preference Awareness Reflection

The intuition behind—and the informal argument in favour of—Awareness Reflection can be applied more generally to your preferences: If you predict that as your awareness grows (but everything else remains fixed) you will reverse a current preference, then that is arguably a reason to reverse it now. For in this case too one could ask why you would not defer to someone who is exactly like you in every respect except that they are more aware than you are.

Consider again for instance the beach-or-home example. Suppose that you predict¹⁰ now that when you are able to specify the catch-all weather contingency, you will prefer not to go (or not to have gone) to the beach. (And as before, let us suppose that you predict that you will be no less rational, and no less informed, when your awareness has grown.) Then it would seem that, intuitively, you should not prefer to go to the beach now, in your less aware state. In other words, it would seem we should accept:

Preference Awareness Reflection. *For any awareness context and any pair of actions a and b , if you predict that between now and time t , the only thing that happens is that you gain more awareness, then you should not now predict that*

¹⁰We can think of this prediction of yours as corresponding to your expectation of expected utility. For instance, that you now predict that you will later prefer not to go to the beach, means that your current credences over your future credence and utility functions is such that your current expectation for your future expected utility assignment to the option of going to the beach is lower than your current expectation for your future expected utility assignment to the option of not going to the beach.

your preference ranking of a vs. b at time t differs from your current ranking of a vs. b.

It might be illuminating to compare the above principle to a stronger and more general version. Let's use the term "Preference Reflection" for the principle that you should not now predict that your preference ranking of *a* vs. *b* at some later time *t* differs from your current ranking of *a* vs. *b*.¹¹ Note that Preference Awareness Reflection is a special case of Preference Reflection in that the former says that the latter holds in the special case where the only thing that happens until time *t* is that the agent gains more awareness. We do not claim that Preference Reflection is a general requirement of rationality. However, it would seem irrational to violate the general principle just because you predict that you will gain more awareness.¹²

Since Preference Awareness Reflection only holds in cases where you predict that "between now and time *t*, the only thing that happens is that you gain more awareness", the most obvious complaints one would have about a more general principle like Preference Reflection seem not to apply to this special case.¹³ Now, similarly to what we acknowledged in the belief case, it could be that you predict an experience by which your awareness grows to be so "personally transformative" (Paul 2014) that your fundamental values will change in such a way that you will not be the same person. Experimenting with planetary-scale geoengineering or with hallucinogenic drugs might be transformative in this way for the global community and for a single person respectively. Again, we do not want to rule out the possibility that, upon awareness growth, your values change in such a way that you will not be the same person. But we nevertheless set such

¹¹Preference Reflection is similar to what Arntzenius (2008) calls "Desire Reflection", which is stated in terms of a numerical representation of desire (i.e., utility or "desirability") rather than in terms of binary preference.

¹²Similarly, Arntzenius claims that Desire Reflection should not be violated merely because one conditionalises upon new evidence (2008: 279). He weakens Desire Reflection by adding a condition stipulating that the only thing that happens in the relevant time interval is that the agent conditionalises on new evidence, and calls the resulting principle "Weak Desire Reflection". It is this weaker principle that he ends up defending.

¹³For instance, Hedden (2015) argues that what we called Preference Reflection is undermined by the fact that what you prefer often depends on what you have chosen; hence, Preference Reflection will often imply that what you now should prefer depends on what you believe you will choose. However, if between now and time *t* you have made some choice, then Preference Awareness Reflection will not hold, since in that case it is not true that "between now and time *t*, the only thing that happens is that you gain more awareness". Similarly with personal identity: if your identity changes between now and *t*, then it is not true that you have only gained awareness between now and *t*. And so on.

possibilities aside for now. And note that the principle does not apply to such cases, since in them it is not true that the *only* thing that happens is that you gain more awareness.

So, let's return our focus to the presumably more typical, non-transformative cases of anticipated awareness growth. Can we make any positive argument in favour of Preference Awareness Reflection? Unlike the Awareness Reflection principle for belief, a traditional (diachronic) Dutch book argument, like that discussed in the last section, cannot be made against you if you violate Preference Awareness Reflection. However, you are vulnerable to a more general "dynamic consistency" argument. In particular, you will be willing to pay a price to limit your future options and to bind yourself if your preferences violate Preference Awareness Reflection; so, having preferences that violate this principle may be costly.

For instance, suppose that while you now strictly prefer going to the beach to staying home, you predict that you will prefer to stay home if you become aware of the content of the catch-all weather contingency. Then there is some price that you should be willing to pay to remove the option of staying home if and when you have become aware of the catch-all. Similarly, you should be willing to pay some cost to make a binding decision now to go to the beach.¹⁴ And one might think that there is something irrational about an attitudinal state that makes one vulnerable to such unfortunate, and seemingly unnecessary, expenses.

However, one might wonder whether the above dynamic consistency argument doesn't prove much. One can often predict, at some stage of one's life, that one will have different preferences at later stages in one's lives. For instance, when the sleepless nights that are associated with raising toddlers are fresh in one's memory, one might undergo a vasectomy to prevent oneself from acting on a future temptation to have another child. But it would seem that paying to limit one's options can in that case be perfectly rational. Is the predicted preference change in the beach example, and the associated willingness to bind oneself in that case, any different?

One potential difference, that might suggest that the dynamic consistency argument does at least indicate some problem in the beach-or-home

¹⁴The same goes, of course, for outcomes: If you expect that your preference between two outcomes will change when your awareness grows, then you might now be willing to pay to bind yourself from swapping outcomes in the event that your awareness grows.

example, is that in that example one expects to undergo a preference change as a result of *gaining more information or worldly wisdom*. So, the person who pays to bind herself from staying at home if she becomes more aware, is binding herself from acting on more information than what she now has. In other words, she is accepting a price for being able to act on less rather than more information. And that does seem irrational—at least if one assumes that the gain in information, that is, the awareness growth, is not associated with some change in fundamental values nor does it change the agent in some other fundamental way (which seems plausible to assume in the beach-or-home example).¹⁵ In contrast, in the vasectomy case, binding may seem rational because we assume that the person has more information, or a more vivid memory of what it is like to have small children, when he makes the decision compared to some later time. Alternatively, the person may predict that his fundamental values change—in a way that essentially makes him a different person—as he gets older. In any case, paying a price to prevent oneself from acting on a more aware epistemic state could be irrational in a way that many instances of binding are not.

7.6 Concluding remarks on chapter 7

Let's take stock of the main message of this chapter. We have considered very strict requirements on agents who anticipate awareness growth, namely, requirements that such agents neither expect that their credences be affected on balance by awareness growth, nor predict (in the sense of fn. 10) that awareness growth will reverse any of their preferences.

This finding provides an additional reason for including a “catch-all” when modelling an agent who anticipates awareness growth. Suppose that an agent is choosing between options f and g , abstractly represented in table 6.2 above, and let's assume that the agent reasons that conditional on all the events that she is aware of, that is, events E_1 to E_n , she would prefer f to g ; nevertheless, she unconditionally prefers g to f . Given the above argument, we cannot rationalise this preference pattern by stipulating that the agent's expectation of expected utility for f and/or g upon awareness

¹⁵Note that in this respect the dynamic argument for Preference Awareness Reflection is like the much discussed dynamic arguments in favour of the Independence axiom and the Sure Thing principle. For a discussion, see e.g. McClennen (1990). See too Steele (2010, 2018) for a general discussion of dynamic arguments for principles of rationality.

growth diverges from her current expected utility for f and/or g . Such divergence would be irrational. Instead, we can rationalise it by stipulating that the probability weighted utility of the catch-all outcome for f is sufficiently negative, compared to the catch-all outcome for g , that the overall expected utility of g is greater than that of f . But then we need to include a catch-all when modelling this agent's epistemic state. And this catch-all should of course be what we in the last chapter called "subjective", that is, standing in for possibilities that the agent herself thinks she is missing and will later become aware of, rather than, say, standing in for the set of all possibilities that she has actually left out.

8

Conclusion

8.1 ‘Whereof one cannot speak, thereof one must be silent’

We noted at the outset of this book that there are surely limits to what can be said about one’s own limited awareness at a time. In reasoning, one tries to account for all the possible ways that the world might be that are relevant to one’s practical purposes. But this reasoning is limited by one’s vantage point. One may not be able to discern all the possible contingencies that an onlooker or even one’s later self is able to discern. Those things of which one is unaware are forcibly absent from one’s reasoning.¹

The question is whether limited awareness, duly recognised as an absence in an agent’s reasoning, may nonetheless play a role in her reasoning. This book has built on previous work by others in reckoning with this key question. By way of analogy: even if a driver is not able to see potholes in the road, she may react in better and worse ways when she encounters a pothole, and she may practice better and worse defensive driving to avoid or guard against any fallout from such encounters. Similarly, even if one is unaware of all relevant contingencies, perhaps one may adjust one’s reasoning in better and worse ways when one encounters a contingency of which one was previously unaware, and perhaps there are better and worse ways to reason defensively in order to guard against any fallout from such

¹Hence use of a well-known quote from Wittgenstein (1922, proposition 7) as the title of this section. While apt (roughly speaking) for our purposes, the quote conveys, in Wittgenstein’s discussion, a rather different point.

encounters.

Standard decision theory does not deal in such reasoning “potholes” and the normative issues they raise. It simply ignores them, assuming that the contingencies of which an agent is aware at a time are just those of which she is ever aware, at any time. The problem is that this does not do justice to the experiences of many in reasoning. We do apparently encounter our own limited awareness as we undergo awareness growth. And this leads us to expect and be wary of encountering awareness growth in similar kinds of scenarios. Standard decision theory thus seems to let us down. We would ideally like to use decision theory to help us reason about what to do about the world’s most pressing problems, such as say climate change and species extinction. However, as the examples throughout this book suggest, these are the types of decisions where we have particularly strong reasons to think that our choices may result in outcomes, or depend on contingencies, of which we are currently unaware. Since the standard decision-models, such as those of Savage (1954) and Jeffrey (1965), were developed for decision-making with full awareness, they are apparently not well-suited to help us solve these pressing problems.

Still, wish as we might for a way to deal with reasoning potholes, especially when the stakes are high, they may evade norms of rationality. Trying to reason *beyond uncertainty* may be akin to searching for the holy grail. While this book joins the search for norms of rationality for responding to as well as anticipating growing awareness, we do not presume that there are any such norms to be found. Indeed, the conclusions that we arrive at over the course of the book are somewhat ambiguous: we suggest norms for responding to and anticipating growing awareness but we also cast doubt on whether these norms are truly substantive or distinct. Even when read in the most deflationary way, however, our analysis shows that reckoning with limited awareness is important for finessing one’s reasoning roadmap at any given time—for determining what, all things considered, is one’s best assessment of the possible contingencies, given the various learning experiences and further choices one may encounter later.

8.2 Norms for limited awareness

Let us then recap the normative conclusions of the book, however substantive or distinct one may regard them.

Note that an initial task, in order to even begin contemplating norms, was to characterise limited awareness and subsequent awareness growth. Hence in chapter 2 we described decision models that represent an agent's own fallible perspective. Such models may fail to account for all possible contingencies, and when/if an agent recognises this, she undergoes awareness growth. In chapter 3 we considered more carefully how to model a transition from one state of awareness—what we dub an *awareness context*—to another. We suggested that this modelling exercise is in itself enlightening with respect to better understanding an agent's reasoning at a time. Indeed, we hope that our discussion in these early chapters contributes not only to the awareness literature, but more generally highlights the limits of the standard possible-world models of cognitive states.

With a general model of awareness growth in hand, we were in a position to investigate potential norms of rationality for, respectively, responding to and anticipating awareness growth. We approached these topics in turn, the former in chapters 4 and 5 and the latter in chapters 6 and 7.

One might say that the most striking new norm(s) we identified regarding limited awareness were those canvassed at the end of the book: awareness version(s) of the so-called *Reflection Principle*. The belief norm, which we dubbed *Awareness Reflection*, applies in cases where an agent anticipates her awareness growth in a rather precise way: she can specify all her possible future credences (over those propositions in her current awareness context) after her awareness has grown, whatever it is that she comes to be aware of. If, in addition, the agent takes herself to be the same agent after the awareness growth in question, her current credences should be the expectation of her future credences.

We take Awareness Reflection to be a substantive and distinct belief norm. There are reasons why it might be downplayed. In particular, while on first glance the norm appears to strongly constrain how an agent's credences at one time relate to her credences at another time, on closer inspection one sees that it is not actual future credences but rather predicted future credences that play a role. Once this is appreciated, the norm can

be understood as simply describing how a rational agent arrives at her current credences: by thinking through what are her possible better-informed credences and taking the expectation of these future credences. This is very much in line with the familiar Reflection Principle, and indeed in chapters 6 and 7 we emphasised the continuity between “ordinary” reasoning and reasoning that involves anticipated awareness growth. But that continuity is the very reason we regard Awareness Reflection to be a surprising norm. The fact that anticipated awareness growth is akin to anticipated learning of a more ordinary kind is itself an important finding.

Our analysis of anticipated awareness growth brings further perspective to the earlier more negative findings of the book. Initially we set out to explore simply whether there are better and worse ways to respond to some particular awareness growth (or alternatively whether there are ways to discern whether one is the same agent having experienced awareness growth). In other words, does rationality (or agential stability) impose any general constraints on the relationship between one’s credences prior to and post some growth in awareness? Against the popular position in the philosophy and economics literature, we argued that there are no such general constraints. This position can be seen to resonate with Awareness Reflection in that this norm does not constrain the predictions that an agent makes about her future credences after awareness growth. An agent may predict any kind of credence change, including rather radical changes, so long as her current credences are the expectation of her predicted future credences post awareness growth.

Nevertheless some of our examples in chapters 4 and 5 suggest there will be many occasions in which an agent’s credence change upon awareness growth will be more minimal. These are interesting cases to characterise. We do so in the form of our *Restricted Reverse Bayesianism* rule. We do not take the rule to be a substantive norm; it simply describes cases in which what the agent becomes aware of is *evidentially irrelevant* to the pair of “old” propositions that is of interest. In such circumstances, it follows that the relevant aspect of the agent’s belief change will be conservative in the familiar way: the relative probabilities of the pair of propositions in question will remain constant. We suggest that others who have proposed more general norms for conservative belief change under awareness growth either explicitly or implicitly focus only on cases in which the awareness

growth is evidentially irrelevant to the pair of propositions at issue.

8.3 Two challenges revisited

All this speaks to at least one of the challenges that we raised in the introductory chapter with respect to whether studying limited awareness and awareness growth is a worthwhile project. One of the worries was whether there could be any normative upshots from better understanding limited awareness, since unlike, say, failures of transitivity, this is not something an agent can do anything about. We suggested that the proof would be in the pudding, and we hope that our summary of the book's findings in the last section makes for a convincing case that our project is indeed interesting from a normative perspective. At the very least, the process of modelling limited awareness and awareness growth is important for a more sophisticated view of reasoning at a time.

The other challenge raised in the introductory chapter concerns whether a study of limited awareness could be scientifically respectable. That is, the worry was that by introducing lack of awareness into a model of an agent, we would inevitably have to start making assumptions about the agent that cannot even in principle be empirically verified.

One partial response to this challenge, which we briefly discussed in chapters 6 and 7, was that as long as an agent has a consistent *conditional* preference relation, given all the events of which she is aware, we can infer a great deal about her attitudes to that which she takes herself to be unaware of from her all-things-considered preference (and, ideally, choice) between options for which she anticipates awareness growth. Moreover, as we mentioned in the introduction, decision theories have already been developed that allow for a representation of agents' attitudes to that of which they are unaware.

However, for the readers not convinced by the above responses, we can offer a partner-in-crime response. As we discussed in the introduction, we see the extension to limited awareness as a natural next step in decision theory's historical trajectory, from the *objective* expected utility theory of von Neumann and Morgenstern (1947), where the only subjective element is the extent to which agents desire outcomes, to the *subjective* expected utility theories of Savage (1954) and Jeffrey (1965), where the extent to which agents

believe that various events will occur is also subjective. As we pointed out, the latter two theories however do not allow for any subjectivity when it comes to what is possible or available—that is, they do not allow for limited awareness.

Now, a well-known problem—or feature, depending on one’s philosophical views—of introducing the additional subjective variable to represent agent’s beliefs, is that it invariably introduces some (additional) assumptions about the subject that cannot even in principle be empirically verified. For instance, Savage assumes that agents’ have preferences between any functions from his set of states of the world to his set of consequences. But, as has been much discussed, some of these functions will correspond to options that are not only physically (perhaps even *metaphysically*) impossible, but also impossible according to the agent whose attitudes are being represented (see e.g. Joyce 1999: chapter 3). But evidently we cannot devise a choice scenario that reveals an agent’s preference between options that she thinks are impossible.

Jeffrey on the other hand does not make as strong non-empirical assumptions as Savage. But an implication of this is that Jeffrey’s framework results in inconsistent representations of agents’ attitudes. In particular, for any agent who satisfies all assumptions of Jeffrey’s framework, and for most² contingent but logically independent propositions *A* and *B*, the agent will both be representable as believing *A* more strongly than *B* and *B* more strongly than *A*.³ However, since people who satisfy all of Jeffrey’s assumptions have to be very rational indeed, it is assumed that this seeming inconsistency is a problem with the representation, not with the agents themselves. But this assumption cannot, within Jeffrey’s system, be empirically justified—what can at best be observed, namely, an agent’s preferences as revealed by her choices, is consistent with her both believing *A* more strongly than *B* and *B* more strongly than *A*.

In sum, the move from objective to subjective expected utility theory, which was celebrated as a great achievement, brought with it assumptions about agents that are even in principle empirically unverifiable. So, while the hard-lined empiricist may still not be comforted by this, at least we

²In particular, for all propositions that are not of “neutral” desirability, that is, no more desirable than the tautology.

³This assumes that the agent’s preferences are not unbounded. For a discussion of this assumption, see Joyce (1999: chapter 4.)

take comfort in the fact that we are in good company in accepting the introduction of empirically unverifiable assumptions about agents as a cost of making decision theory more subjective.

8.4 Connection to applied work

Alongside theoretical developments in modelling and understanding growing awareness, there has and continues to be progress of a more practical kind. This is in the form of decision support tools for relatively novel and/or complex decisions that assist in identifying what are the relevant possible contingencies and one's attitudes towards them. We regard this work as complementary to the more general and abstract treatment of growing awareness that is the focus of this book. Ideally the two would mutually inform each other.

For instance, one popular family of decision-support approaches is known as *scenario-based planning* (Schwartz 1996). The aim is to assist in mapping out the range and boundaries of the possibility space for a given decision—to determine just how disparate are the scenarios or fully detailed ways that the world might be that are pertinent to the choice at hand. A simple qualitative approach to this effect involves identifying the key factors or “axes” that discriminate the outcomes of different options. For instance, two such key factors with respect to future global emissions scenarios (that are pertinent to mitigation decisions) are purportedly demographic (population) change and the rate and direction of technological change (Nakicenovic et al. 2000). A telling range of emissions scenarios can thus be constructed from combinations of extreme values on these and the other key axes. In other cases, the key axes and associated spread of scenarios may be less easy to discern with the naked eye, so to speak. If it so happens that there is nonetheless a rich predictive model available, computer-assisted scenario discovery approaches (e.g., Groves and Lempert 2007, Bryant and Lempert 2010) may be useful. One such approach involves automated identification and clustering of what may be hundreds to millions of potentially important scenarios generated by a complex predictive model with large ranges for the parameter values. The idea is that once scenarios are clustered they are cognitively accessible and thus more meaningfully evaluated.

We suggest that scenario-based planning and other decision-support approaches associated with “horizon scanning” are, among other things, implicitly techniques for anticipating awareness growth, in that they aim for coverage of the possibility space whether or not all details can as yet be specified. As such, these methods may effectively flesh out what reckoning with limited and growing awareness looks like in the context of real and significant decision problems. This more practical task has not been our preoccupation in this book. But we hope that our work here may provide food for thought and helpful markers in the sand, so to speak, for those working at this important practical end of the spectrum.

8.5 Further research

As noted at the outset, our methodology in studying limited awareness was to consider this phenomenon in isolation. We thus set aside various other challenges to orthodox expected utility theory. For instance, we did not entertain generalisations of expected utility theory that accommodate alternative representations of an agent’s uncertainty such as imprecise probabilities (or sets of probability functions). Nor did we entertain generalisations of expected utility theory that accommodate, on the preference side, more complex kinds of risk aversion. For those sympathetic to theories of either kind, our analysis can be understood as making certain idealising assumptions, for instance, that in the cases we examine the agent’s credences are precise probabilities and she is moreover risk-neutral in whatever is the relevant sense of that term.

But idealising in this way may not be apt if in fact the phenomenon of awareness growth is not independent (in a normative sense) of the representation of uncertainty and/or risk aversion. Let us focus just on the former issue. One might think that an appropriate response to awareness growth requires “going imprecise”: roughly, the “new” propositions of which one has become aware are assigned maximally imprecise credence (perhaps conditional on some relevant “old” proposition). Moreover, one might think that when an agent anticipates awareness growth, her uncertainty is often so severe that it is best represented by imprecise probabilities. As such, one might argue that allowing credences to be imprecise is crucial for modelling awareness growth and the anticipation of it in a compelling

way, and for expressing the relevant norms.

While we have some sympathy for this line, we think it would be very surprising indeed if imprecise credence were rationally mandated for growing awareness and/or its anticipation. And, as noted above, we do not think the mere permissibility of imprecise credence in contexts of growing awareness undercuts our analysis. Those who subscribe to imprecise credence being rationally permissible can view our method as one of idealisation: we assume that in all cases under investigation the agent happens to have precise credences. Ditto for non-standard kinds of risk aversion. It would be extremely surprising if growing awareness mandated some particular risk-averse attitude. And if non-standard kinds of risk aversion are merely permissible, then this does not compromise our approach to studying growing awareness.

All that said, it remains an interesting project to relax any idealisations or scope restrictions that are inherent in our analysis. We welcome further investigation of how growing awareness interacts with non-standard representations of uncertainty and risk aversion. Note that some inroads have already been made on this project, at least with respect to imprecise attitudes. For instance, Bradley (2017: ch. 12) permits imprecise attitudes in his exposition of growing awareness. Economists Dominiak and Tserenjigmid (2018) explore in a working paper the transition from precise to imprecise probabilities upon growing awareness and suggest constraints on such a transition. Our findings in this book provide a base from which to examine these models/proposals that relate growing awareness to imprecise attitudes.

There is a further idealisation in our study of growing awareness that is tangential to the treatment of uncertainty and risk aversion. Our account makes the standard assumption that agents *progress* in their reasoning and view on the world. In particular, no forgetting or becoming less aware is allowed. The assumption of exclusive progress may be relatively harmless for most normative enquiries. However, when it comes to awareness growth, there may be progressive reasons to eradicate old concepts in one's proposition space, since it may not always be a matter of forgetting. Rather, it may be a matter of conceptual learning, where old and outdated concepts are replaced by new ones.

We make this suggestion with caution, because we suspect that this way

of thinking of an agent's concepts—they they may be replaced by more explanatory or in some other sense more apt concepts—would represent a huge expansion of the notion of reasoning and learning that decision theory has been designed to capture. It would involve seeing learning not as a linear accumulation of knowledge but rather a jerky process that admits of conceptual revolutions à la Kuhn (1962). So this may be very far from incremental future research. And it may ultimately be regarded beyond the scope of decision theory. But all the more reason, we say, to make some preliminary investigations to see how different reasoning would look were we to allow for progressive “loss of awareness”.

8.6 Closing remarks

As the above (incomplete) list of issues left out from this book illustrate, there is still a lot of work to be done on the connection between rationality and limited awareness. That is, there remain important theoretical questions quite apart from the many further practical ones associated with making good decisions under limited awareness. But we believe that we have nevertheless clarified and moved the frontier of research on limited awareness. Our main hope for this book, however, is simply that it will bring further attention to the importance of studying limited and growing awareness, this being a fundamental aspect of our predicament in the world as reasoning agents.

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