

Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

Two Approaches to Belief Revision

Ted Shear¹ Branden Fitelson² Jonathan Weisberg³

¹UC-Davis & University of Queensland
ebshear@ucdavis.edu

²Northeastern University
branden@fitelson.org

³University of Toronto
jonathan.weisberg@utoronto.ca

February 6, 2017

Shear, Fitelson & Weisberg Two Approaches to Belief Revision 1

Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

Today, we will be comparing two approaches to belief revision: a *Bayesian/Lockean* approach, and a *Logical/AGM* approach.

Here is an outline of the talk:

- Explain how the Bayesian/Lockean approach works,
- Explain how the Logical/AGM approach works,
- Compare and contrast these two approaches.
 - Along the way, we'll report some new results regarding the relationship between these approaches, and point out some connections to recent work (esp. Hannes's work on *stability*).
- First, some formal background and setup.

Shear, Fitelson & Weisberg Two Approaches to Belief Revision 2

Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

Our agents will possess *both* qualitative (yes/no) belief sets, \mathbf{B} , *and* numerical confidence/credence functions, $b(\cdot)$.

On the *belief* side, our agents entertain (classical, possible worlds) propositions on some finite *agenda* \mathcal{A} .

(1) \mathbf{B} is the set of propositions in \mathcal{A} believed by our agent.

- *Note*: when $p \in \mathbf{B}$, we write $\mathbf{B}(p)$.

(2) Given a *prior* belief set \mathbf{B} , the *posterior* belief set \mathbf{B}' is generated by revising the prior by E — i.e., $\mathbf{B}' = \mathbf{B} \star E$.

On the *credence* side:

(3) $b(\cdot)$ is a classical (Kolmogorov) probability function.

(4) Given a *prior* $b(\cdot)$, the *posterior* $b'(\cdot)$ is generated *via conditionalization* by E — i.e., $b'(\cdot) = b(\cdot | E)$.¹

¹Our results generalize to “minimum distance” [4] Bayesian updates satisfying (i) $b'(E) > b(E)$, (ii) $b'(E) \geq t$, and (iii) $b'(X) \geq t \Rightarrow b(E \supset X) \geq t$.

Shear, Fitelson & Weisberg Two Approaches to Belief Revision 3

Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

The Bayesian approach implies a (*weak*) *Lockean thesis* [5, 7].

$$\mathbf{B}(p) \text{ only if } b(p) \geq t.$$

It is well known [2, 9] that *strong (iff)* Bayesianism/Lockeanism leads to belief sets (e.g., lotteries) that are *not deductively cogent*.

Cogency. An agent's belief set \mathbf{B} is *cogent* iff it is (a) deductively *consistent* and (b) *closed* under logic.

We will not dwell on these well-known differences. Mainly, our talk will be about *cogent* Bayesian agents (similar to Leitgeb's).

But, first, we will briefly discuss “pure” Bayesianism. The *pure* Bayesian is a *strong Lockean (synchronically & diachronically)*.

Pure Bayesian Revision. When one revises one's beliefs, one's posterior belief set \mathbf{B}' is *Lockean* (in a *pure* sense), i.e.,

$$\mathbf{B}' = \mathbf{B} \star E := \{p \mid b(p | E) \geq t\}$$

Shear, Fitelson & Weisberg Two Approaches to Belief Revision 4

Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

To get a feel for how (pure) Bayesian revision works, it is helpful to note that it satisfies the following principle.

Very Weak Preservation. If $E, X \in \mathbf{B}$, then $\neg X \notin \mathbf{B} \star E$.

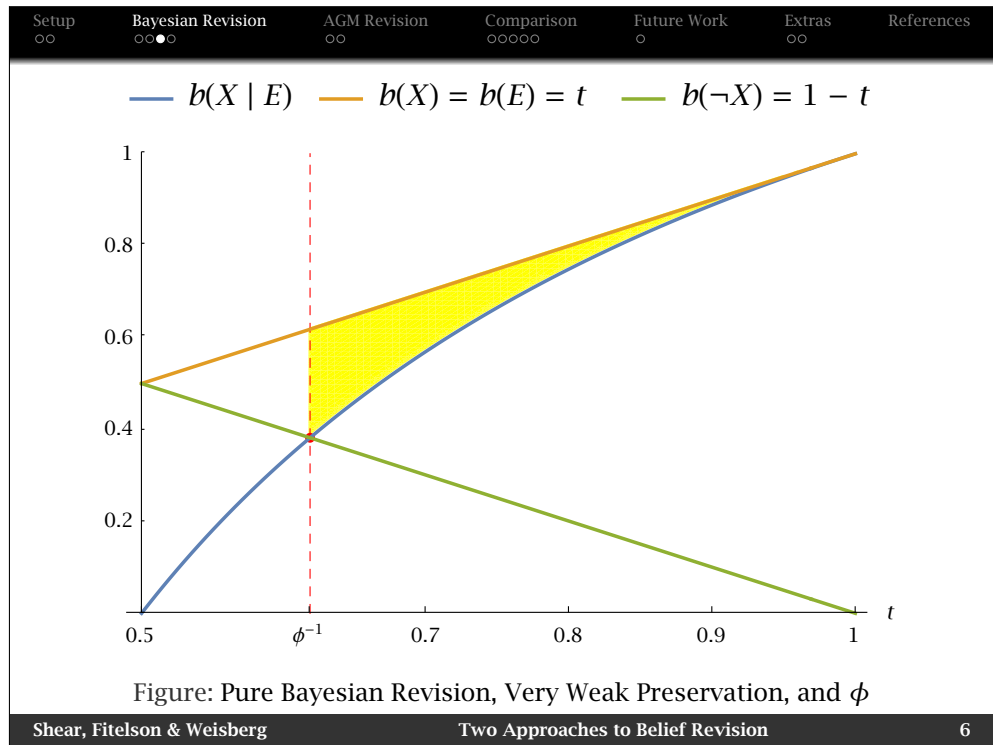
In words: *learning something you already believe should never cause you to **disbelieve** something you already believed.*

Strictly speaking, PBR does *not universally* satisfy VWP. But, it will — *so long as the Lockean threshold t is sufficiently large.*

Specifically, Pure Bayesian Revision is guaranteed to satisfy Very Weak Preservation, *provided that the Lockean threshold t is at least $\phi^{-1} \approx 0.618$, where ϕ is the Golden Ratio.*

Figure is helpful for visualizing why Pure Bayesian Revision, Very Weak Preservation, and The Golden Ratio are related in this way.

Shear, Fitelson & Weisberg Two Approaches to Belief Revision 5



Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

A stronger (but still pretty weak) version of Preservation has been discussed (more extensively) in the literature (*e.g.*, [13]).

Weak Preservation. If $E, X \in \mathbf{B}$, then $X \in \mathbf{B} \star E$.

In words: *learning something you already believe should never cause you to **not believe** something you already believed.*

Interestingly, Pure Bayesian Revision can *violate* Weak Preservation — *even for high Lockean thresholds $t \geq \phi^{-1}$.*

These high-credence (pure Bayesian) violations of Weak Preservation can be visualized *via* the *yellow region* in Figure 1.

However, if a Pure Bayesian *violates* Weak Preservation, *then* their prior belief set \mathbf{B} *must not have been closed under logic.*

Proof.
 Suppose (a) $E, X \in \mathbf{B}$, but (b) $X \notin \mathbf{B} \star E$. (b) implies $\Pr(X | E) < t$. Therefore, $\Pr(X \& E) < t \cdot \Pr(E) < t$. So, $X \& E \notin \mathbf{B}$. \square

Shear, Fitelson & Weisberg Two Approaches to Belief Revision 7

Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

AGM theory is the orthodox, *logical* account of belief revision and its constraints on revisions may *seem* quite modest.

AGM can be understood as embodying a *principle of conservativity* (*aka.*, *informational economy/minimal mutilation*).

Conservativity. When an agent with a prior belief set \mathbf{B} learns (exactly) E , she should revise to a posterior belief set \mathbf{B}' that:

- (1) *includes* E ,
- (2) is *cogent*, and
- (3) constitutes a *minimal change*² to \mathbf{B} .

AGM revision has a number of nice logico-mathematical properties that have made for fruitful investigation.

Specifically, it has a simple and straightforward *axiomatization*.

²“Minimal change” can be explicated (in *geodesic* terms) *via* a wide variety of measures of distance between prior and posterior belief sets [12, 3, 6].

Shear, Fitelson & Weisberg Two Approaches to Belief Revision 8

Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

Basic Gärdenfors postulates:

- (*1) $\mathbf{B} * E = \text{Cn}(\mathbf{B} * E)$ Closure
- (*2) $E \in \mathbf{B} * E$ Success
- (*3) $\mathbf{B} * E \subseteq \text{Cn}(\mathbf{B} \cup \{E\})$ Inclusion
- (*4) If E is consistent with \mathbf{B} , then $\mathbf{B} * E \supseteq \mathbf{B}$ Preservation
- (*5) If E is non-contradictory, then $\mathbf{B} * E$ is consistent Consistency
- (*6) If $X \Leftrightarrow Y$, then $\mathbf{B} * X = \mathbf{B} * Y$ Extensionality

[Note: Strictly speaking, Gärdenfors's (*4) was **Vacuity** and not **Preservation**.
 (*4) If E is consistent with \mathbf{B} , then $\mathbf{B} * E \supseteq \text{Cn}(\mathbf{B} \cup \{E\})$ Vacuity

But, given the other postulates, these two axioms are equivalent. And, it makes our presentation more elegant/continuous to use **Preservation** here, because it implies both of the weaker forms of Preservation above.]

Shear, Fitelson & Weisberg Two Approaches to Belief Revision 9

Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

(*3) $\mathbf{B} * E \subseteq \text{Cn}(\mathbf{B} \cup \{E\})$ Inclusion

Proposition. $*$ satisfies Inclusion.

Proof: Suppose $X \in \mathbf{B} * E$. By the definition of $*$,

$$b(X | E) \geq t.$$

It is a theorem of the probability calculus that

$$\text{Pr}(E \supset X) \geq \text{Pr}(X | E).$$

Therefore,

$$b(E \supset X) \geq t.$$

Hence, $E \supset X \in \mathbf{B}$. So, by *modus ponens* (for \supset), we arrive at

$$X \in \text{Cn}(\mathbf{B} \cup \{E\}).$$

Therefore, $\mathbf{B} * E \subseteq \text{Cn}(\mathbf{B} \cup \{E\})$. □

Shear, Fitelson & Weisberg Two Approaches to Belief Revision 10

Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

(*4) If E is consistent with \mathbf{B} , then $\mathbf{B} * E \supseteq \mathbf{B}$ Preservation

Proposition. *Even Cogent Bayesians can violate Preservation.*

Proof: Our agent will sample a single object at random from the urn on the right.

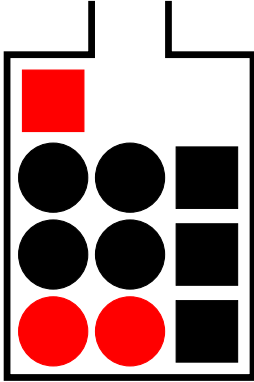
E = 'The object sampled will be red'
 X = 'The object sampled will be a circle'

Assume our (Cogent) Bayesian agent has a Lockean threshold of $t = 0.85$.

Then, $E \supset X$ is *the only proposition* initially believed by our agent (see slide 16). *I.e.*,

$$\mathbf{B} = \{E \supset X\}.$$

Now, *what happens if she learns E?*



Shear, Fitelson & Weisberg Two Approaches to Belief Revision 11

Setup Bayesian Revision AGM Revision Comparison Future Work Extras References

(*4) If E is consistent with \mathbf{B} , then $\mathbf{B} * E \supseteq \mathbf{B}$ Preservation

Proposition. *Even Cogent Bayesians can violate Preservation.*

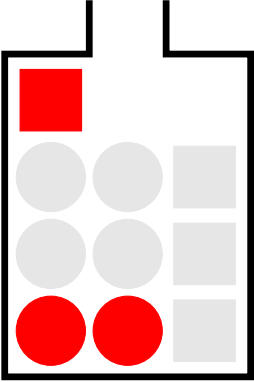
Proof: (continued)

If she learns E (which is consistent with \mathbf{B}), this *rules out the black objects*. Hence,

$$b(E \supset X | E) = 2/3 < 0.85.$$

Thus, $E \supset X \notin \mathbf{B} * E$, but $E \supset X \in \mathbf{B}$. □

Note: our Bayesian agent is *cogent at both the prior and posterior times*. To see this, note that her posterior belief set will be:

$$\mathbf{B}' = \mathbf{B} * E = \{E, E \vee X, E \vee \neg X\}$$


Shear, Fitelson & Weisberg Two Approaches to Belief Revision 12

We conclude with two final theorems concerning Bayesian failures of **Preservation**. The first result suggests Bayesian revision can be seen as being *more epistemically risk-averse* (in one sense) than AGM (which can be seen as *more risk-seeking*).³

Theorem. \ast violates **Preservation** (wrt E, \mathbf{B}) iff $\mathbf{B} \ast E \subset \mathbf{B} \ast E$.

Proof: See Extras Slide 17

So, whenever Bayesian and AGM revision diverge, AGM is more “epistemically demanding” on the agent’s beliefs — in the sense that AGM requires the agent to have *strictly more beliefs* than Bayesianism requires. In this sense, AGM may be seen to be more *epistemically risk-seeking* than Bayesianism.

Our final result brings us back to *The Golden Ratio*...

³Pettigrew [16] has independently argued (via the use of an epistemic *Hurwicz Criterion*) that **Cogency** implies its own variety of *risk-seeking*.

Our final theorem reveals an interesting relationship between *Cogent* Bayesian Revision, **Preservation**, and *The Golden Ratio*.

Theorem. If a *cogent* Bayesian violates **Preservation** (wrt E, \mathbf{B}), then their Lockean threshold t must lie on the half-open interval $[\phi^{-1}, 1)$.

Proof: See the Appendix of our paper.

Corollary. If a *cogent* Bayesian’s Lockean threshold t lies *outside* the half-open interval $[\phi^{-1}, 1)$, then *they must obey* AGM.

To sum up: we have the following three central results.

(Pure Bayesian Revision & $t \in [\phi^{-1}, 1]$) \Rightarrow V.W. Preservation

Cogent Bayesian Revision \Rightarrow W. Preservation

(*Cogent* Bayesian Revision & $t \notin [\phi^{-1}, 1)$) \Rightarrow Preservation/AGM

● *The Holy Grail* would be to find a *purely qualitative axiomatization* of (pure or cogent) Bayesian revision.

- van Eijck & Renne [8] recently provided an axiomatization of (strong) *Lockeanism* with $t = 1/2$, which can be used to axiomatize *pure* Bayesianism (with a $t = 1/2$ threshold).
- And, the $t = 1$ (extremal) case *just is* AGM revision [10, 11]. The problem of axiomatizing Bayesian revision for *intermediate* thresholds $t \in (1/2, 1)$ remains open.
- Since pure Bayesian revision may be described in “minimal distance” terms [4], this suggests a general “geodesic update” framework in which we plan to investigate *contraction* (\div) and other kinds revision. For instance:

1. Let b^* be the **closest probability function** to b s.t. $b^*(p) < t$.
2. $\mathbf{B} \div p := \{p \mid b^*(p) \geq t\}$

p	$b(p)$	$b(p \mid E)$	$p \in \mathbf{B}$?	$p \in \mathbf{B} \ast E$?	$p \in \mathbf{B} \ast E$?	$p \in \text{Cn}(\mathbf{B} \cup \{E\})$?
$E \wedge X$	2/10	2/3	No	No	Yes	Yes
$E \wedge \neg X$	1/10	1/3	No	No	No	No
$\neg E \wedge X$	4/10	0	No	No	No	No
$\neg E \wedge \neg X$	3/10	0	No	No	No	No
E	3/10	1	No	Yes	Yes	Yes
X	6/10	2/3	No	No	Yes	Yes
$E \equiv X$	5/10	2/3	No	No	Yes	Yes
$E \equiv \neg X$	5/10	1/3	No	No	No	No
$\neg E$	7/10	0	No	No	No	No
$\neg X$	4/10	1/3	No	No	No	No
$E \vee X$	7/10	1	No	Yes	Yes	Yes
$E \vee \neg X$	6/10	1	No	Yes	Yes	Yes
$\neg E \vee X$	9/10	2/3	Yes	No	No	Yes
$\neg E \vee \neg X$	8/10	1/3	No	No	No	No

Table: Cogent Bayesian counterexample to **Preservation**

