

# Logical and Probabilistic Models of Belief Change

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# Plan

- ✓ Introduction to belief revision, AGM
- ✓ Possible worlds models, Bayesian models
- ✓ Updating probabilities, The value of learning, Lottery Paradox, Preface Paradox, Review Paradox
- ✓ Lottery Paradox, Preface Paradox, Review Paradox, Iterated belief revision, Context shifts, Becoming aware

Day 5 Interactive epistemology (Agreement Theorems), Convergence Theorems

## Leitgeb's Solution to the Lottery Paradox

In a context in which the agent is interested in *whether ticket  $i$  will be drawn*; for example, for  $i = 1$ : Let  $\Pi$  be the corresponding partition:

$$\{\{w_1\}, \{w_2, \dots, w_{1,000,000}\}\}$$

The resulting probability measure  $P_\Pi$  is given so that  $P$  is given by  $P$  so that:

$$P_\Pi(\{\{w_1\}\}) = \frac{1}{1,000,000} \quad P_\Pi(\{\{w_2, \dots, w_{1,000,000}\}\}) = \frac{999,999}{1,000,000}$$

There are two  $P_{\square}$ -stable sets, and one of the two possible choices for the strongest believed proposition  $B_W^{\square} = \{\{w_2, \dots, w_{1,000,000}\}\}$ .

If  $B_W^{\square}$  is chosen as such, our perfectly rational agent believes of ticket  $i = 1$  that it will not be drawn, (and of course P1 -P3 are satisfied).

For example, this might be a context in which a single ticket holder—the person holding ticket 1—would be inclined to say of his or her ticket: “I believe it won’t win.”

In a context in which the agent is interested in *which ticket will be drawn*: Let  $\Pi'$  be the corresponding partition that consists of all singleton subsets of  $W$ . The probability measure  $P^{\Pi'}$  is the uniform probability on  $W$ .

The only  $P$ -stable set—and hence the only choice for the strongest believed proposition  $B_W^{\Pi'}$ —is  $W$  itself: our perfectly rational agent believes that some ticket will be drawn, but he or she does not believe of any ticket that it will not win

For example, this might be a context in which a salesperson of tickets in a lottery would be inclined to say of each ticket: “It might win” (that is, it is not the case that I believe that it won’t win).

In either of the two contexts from before, the theory avoids the absurd conclusion of the Lottery Paradox; in each context, it preserves the closure of belief under conjunction; and in each context, it preserves the Lockean thesis for some threshold ( $r = \frac{999,999}{1,000,000}$  in the first case,  $r = 1$  in the second case)-all of this follows from  $P$ -stability and the theorem.

In the first  $\Pi$ -context, the intuition is preserved that, in some sense, one believes of ticket  $i$  that it will lose since it is so likely to lose.

In the second  $\Pi'$ -context, the intuition is preserved that, in a different sense, one should not believe of any ticket that it will lose since the situation is symmetric with respect to tickets, as expressed by the uniform probability measure, and of course some ticket must win.



Finally, by disregarding or mixing the contexts, it becomes apparent why one might have regarded all of the premises of the Lottery Paradox as true.

But according to the present theory, contexts should not be disregarded or mixed: partitions  $\Pi$  and  $\Pi'$  differ from each other, and different partitions may lead to different beliefs, as observed in the last section and as exemplified in the Lottery Paradox.

Accordingly, the thresholds in the Lockean thesis may have to be chosen differently in different contexts, and once again, this is what happens in the Lottery Paradox—which makes good sense: in the second  $\Pi'$ -context, by uniformity, the agent's degrees of belief do not give him or her much of a hint of what to believe. That is why the agent ought to be supercautious about her beliefs in that

## Iterated Belief Change

## Two Postulates of Iterated Revision

I1 If  $B \in Cn(\{A\})$  then  $(K * B) * A = K * A$ .

I2 If  $\neg B \in Cn(\{A\})$  then  $(K * A) * B = K * B$

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- ▶ Postulate I2 demands that first learning  $A$  followed by learning a piece of information  $B$  incompatible with  $A$  is the same as simply learning  $B$  outright. So, for example, first learning  $A$  and then  $\neg A$  should result in the same belief state as directly learning  $\neg A$ .

I3 If  $B \in K * A$  then  $B \in (K * B) * A$ .

I4 If  $\neg B \notin K * A$  then  $\neg B \notin (K * B) * A$ .

Robert Stalnaker. *Iterated Belief Revision*. Erkenntnis 70, pp. 189 - 209, 2009.



## Stalnaker's Counterexample to I1

<i>UUU</i>	<i>DDD</i>
<i>UUD</i>	<i>DDU</i>
<i>UDU</i>	<i>DUD</i>
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- ▶ Three independent (reliable) observers report on the switches: Alice says switch 1 is *U*, Bob says switch 2 is *D* and Carla says switch 3 is *U*.

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- ▶ I receive the information that the light is on. What should I believe?

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- ▶ I receive the information that the light is on. What should I believe?
- ▶ Cautious: *UUU*, *DDD*; Bold: *UUU*

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- ▶ Suppose there are two switches:  $L_1$  is the main switch and  $L_2$  is a secondary switch controlled by the first two lights. (So  $L_1 \rightarrow L_2$ , but not the converse)

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- ▶ So,  $L_2 \in Cn(\{L_1\})$  but (potentially)  
 $(K * L_2) * L_1 \neq K * L_1$ .

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- ▶ Alice reports that the coin in box 1 is lying heads up, Bert reports that the coin in box 2 is lying heads up.
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- ▶ Two new independent witnesses, whose reliability trumps that of Alice's and Bert's, provide additional reports on the status of the coins. Carla reports that the coin in box 1 is lying tails up, and Dora reports that the coin in box 2 is lying tails up.
- ▶ Finally, Elmer, a third witness considered the most reliable overall, reports that the coin in box 1 is lying heads up.

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Yet, since  $H_1 \wedge H_2 \in K'$  and  $H_1$  is consistent with  $H_2$ , we must have  $H_1 \wedge H_2 \in K' * H_1$ , which yields a conflict with the assumption that  $H_1 \wedge T_2 \in K' * (T_1 \wedge T_2) * H_1$ .

*...[Postulate I2] directs us to take back the totality of any information that is overturned. Specifically, if we first receive information  $\alpha$ , and then receive information that conflicts with  $\alpha$ , we should return to the belief state we were previously in, before learning  $\alpha$ . But this directive is too strong. Even if the new information conflicts with the information just received, it need not necessarily cast doubt on all of that information.*

*(Stalnaker, pg. 207–208)*

EP, P. Pedersen and J.-W. Romeijn. *When is an example and counterexample?*. Proceedings of TARK, 2013.

## What Do the Examples Demonstrate?

1. There is no suitable way to formalize the scenario in such a way that the AGM postulates (possibly including postulates of iterated belief revision) can be saved;
2. The AGM framework can be made to agree with the scenario but does not furnish a systematic way to formalize the relevant meta-information; or
3. There is a suitable and systematic way to make the meta-information explicit, but this is something that the AGM framework cannot properly accommodate.



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Our interest in this paper is the third response, which is concerned with the absence of guidelines for applying the theory of belief revision.

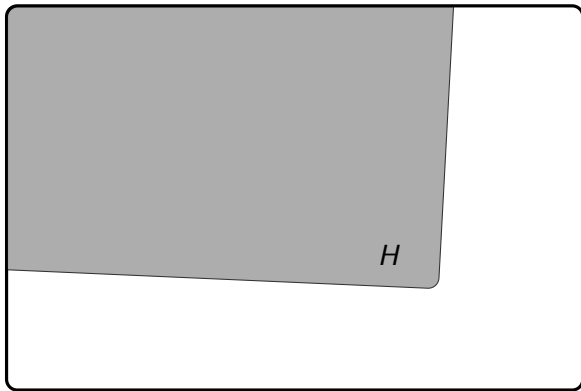
EP. *Dynamics for Probabilistic Common Belief*. Studies in Logic, 2015.

Starting with the same premises, using (for example) first-order logic, two agents cannot disagree about a conclusion.

Starting with the same probability, using (for example) strict conditionalization, two agents cannot disagree about their posterior probability given the same evidence.

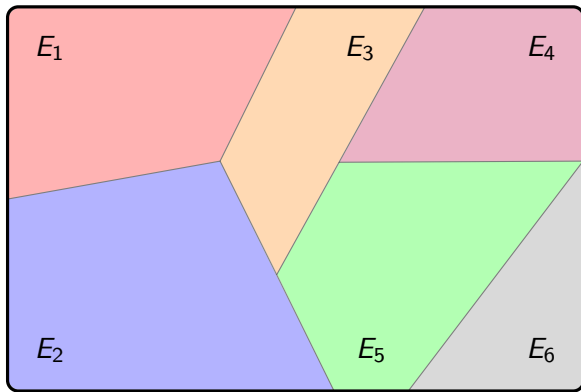
**Aumann's Agreeing to Disagree Theorem.** Suppose that  $n$  agents share a common prior and have different private information. If there is common knowledge of the posteriors of a fixed event, then the posteriors must be equal.

Robert Aumann. *Agreeing to Disagree*. Annals of Statistics **4(6)**, pgs. 1236-1239 (1976).

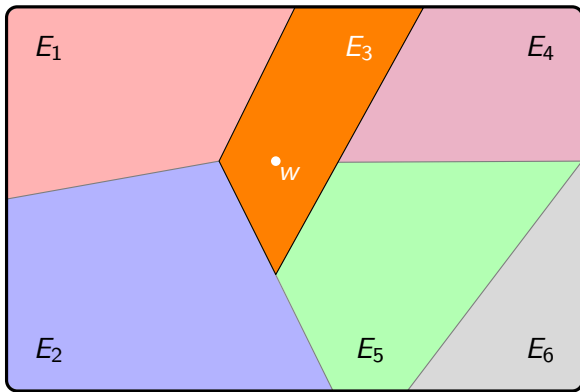


An **event/proposition** is a (definable) subset  $H \subseteq W$ .

A  **$\sigma$ -algebra** is the collection of events/propositions  
(closed under countable unions and complementation)



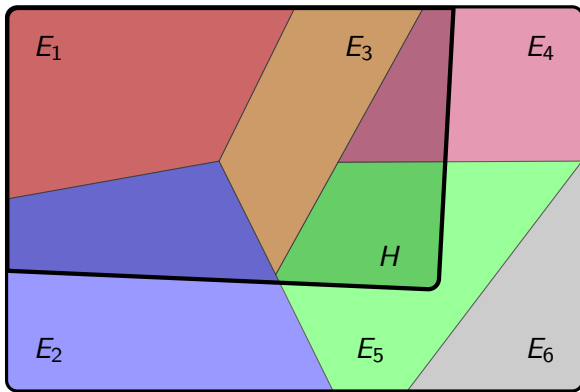
An **experiment/question/set of signals** is a partition  $\mathcal{E}$  on  $W$ .



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If  $w \in W$ , let  $\mathcal{E}[w] = E$  where  $w \in E \in \mathcal{E}$ .

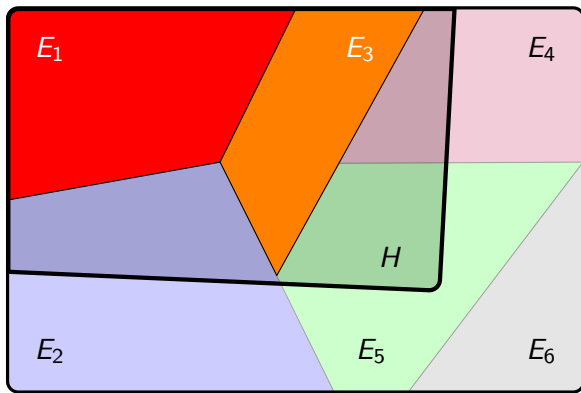
E.g, if  $\mathcal{E} = \{E_1, E_2, E_3, E_4, E_5, E_6\}$ , then  $\mathcal{E}[w] = E_3$



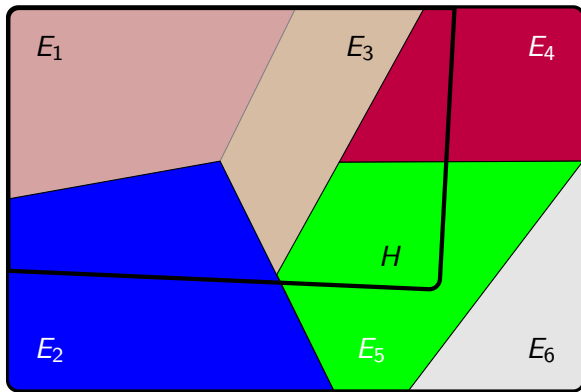
$K_{\mathcal{E}} : \wp(W) \rightarrow \wp(W)$ , where for  $H \subseteq W$ ,

$$K_{\mathcal{E}}(H) = \{w \mid \mathcal{E}[w] \subseteq H\}$$

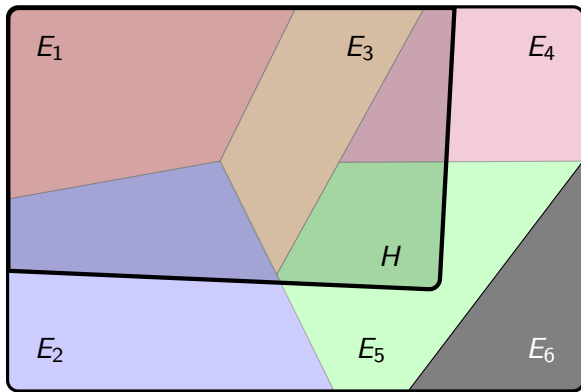




$$K_{\mathcal{E}}(H) = E_1 \cup E_3$$



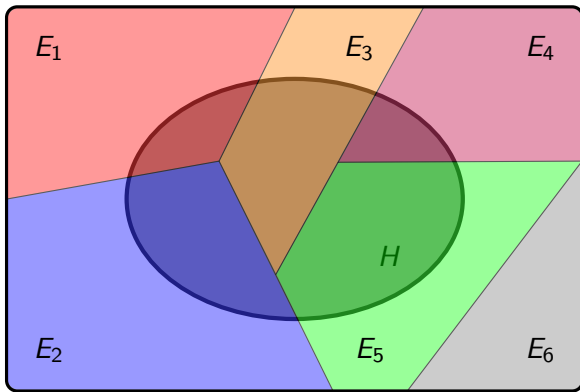
$$K_{\mathcal{E}}(H) = E_1 \cup E_3$$
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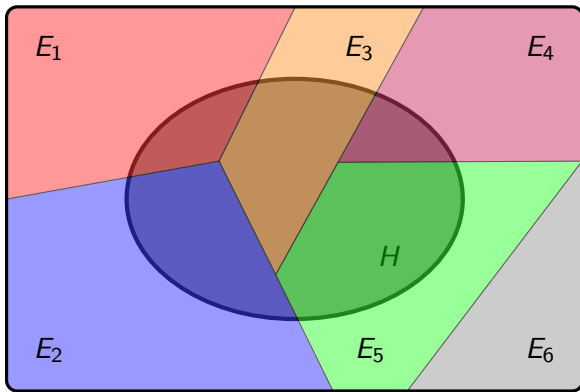
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$$K_{\mathcal{E}}(-H) = E_6$$

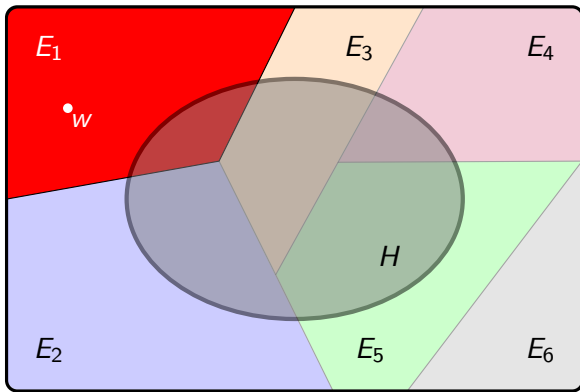


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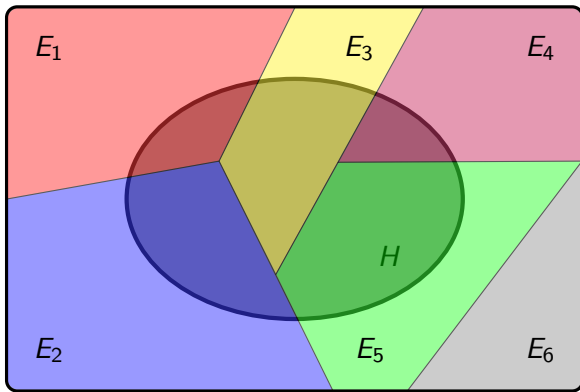
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$$\text{E.g., } p_{\mathcal{E},w}(H) = p(H \mid E_1)$$

A basic result about probabilities.

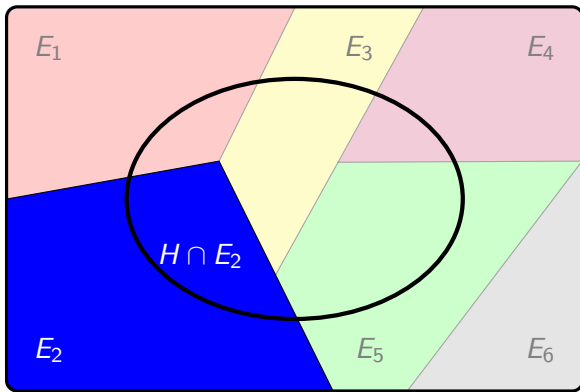
For any finite partition  $\mathcal{E} = \{E_i\}$  of  $W$  and an event  $H$ ,

$$p(H) = \sum_i p(E_i)p(H | E_i)$$

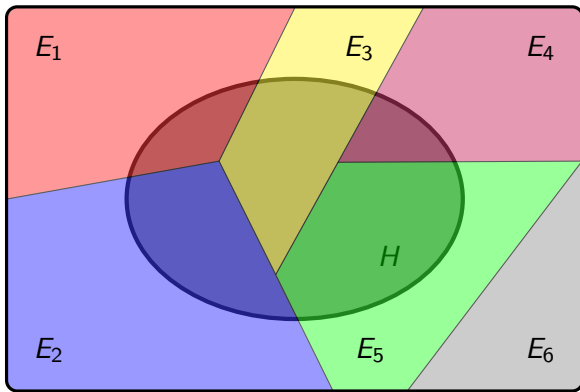


$$p(H) = p(H \cap E_1) + \cdots + p(H \cap E_6)$$

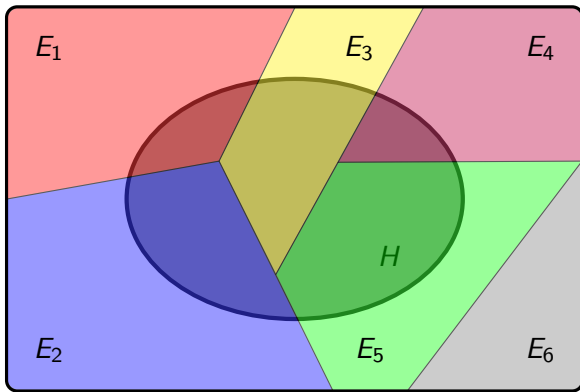




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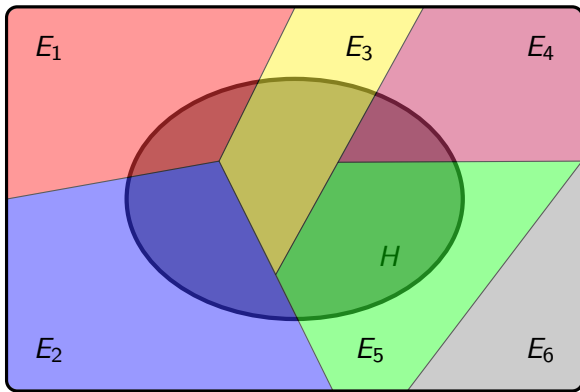


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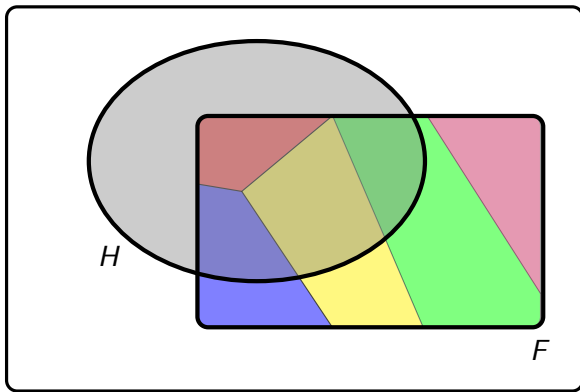
A basic result about probabilities.

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# Common Knowledge

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*It is not common knowledge who defined ‘Common Knowledge’...*

## The first formal definition of common knowledge?

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**Fixed-point definition:**  $\gamma := i$  and  $j$  know that ( $\varphi$  and  $\gamma$ )

G. Harman. *Review of Linguistic Behavior*. Language (1977).

J. Barwise. *Three views of Common Knowledge*. TARK (1987).

The first formal definition of common knowledge?

M. Friedell. *On the Structure of Shared Awareness*. Behavioral Science (1969).

R. Aumann. *Agreeing to Disagree*. Annals of Statistics (1976).

The first rigorous analysis of common knowledge

D. Lewis. *Convention, A Philosophical Study*. 1969.

**Fixed-point definition:**  $\gamma := i$  and  $j$  know that ( $\varphi$  and  $\gamma$ )

G. Harman. *Review of Linguistic Behavior*. Language (1977).

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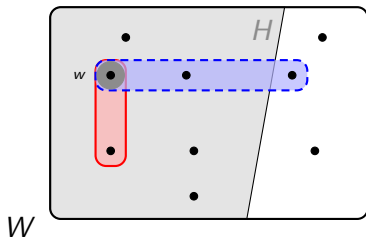
**Shared situation:** There is a *shared situation*  $s$  such that (1)  $s$  entails  $\varphi$ , (2)  $s$  entails everyone knows  $\varphi$ , plus other conditions

H. Clark and C. Marshall. *Definite Reference and Mutual Knowledge*. 1981.

M. Gilbert. *On Social Facts*. Princeton University Press (1989).

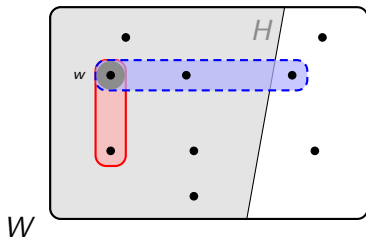
P. Vanderschraaf and G. Sillari. "*Common Knowledge*", *The Stanford Encyclopedia of Philosophy* (2009).

<http://plato.stanford.edu/entries/common-knowledge/>.



Each agent  $i$  is associated with a partition  $\mathcal{E}_i$ .

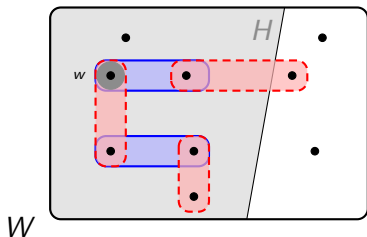
$$K_i : \wp(W) \rightarrow \wp(W) \text{ where } K_i(H) = \{w \mid \mathcal{E}_i[w] \subseteq H\}$$



Each agent  $i$  is associated with a partition  $\mathcal{E}_i$ .

$$w \notin K_1(H) \quad \text{and} \quad w \in K_2(H)$$

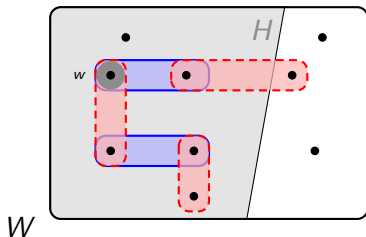




**Everyone Knows:**  $K(H) = \bigcap_{i \in \mathcal{A}} K_i(H)$

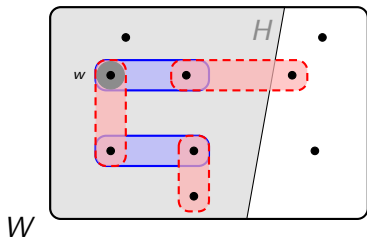
$K^m(H)$  for all  $m \geq 0$  is defined as:

$$K^0(H) = H \quad K^m(H) = K(K^{m-1}(H))$$

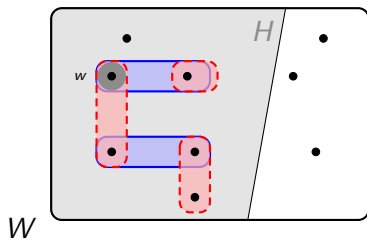


**Common Knowledge:**  $C : \wp(W) \rightarrow \wp(W)$  with

$$C(H) = \bigcap_{m \geq 0} K^m(H)$$



$$w \in K(H) \quad w \notin C(H)$$

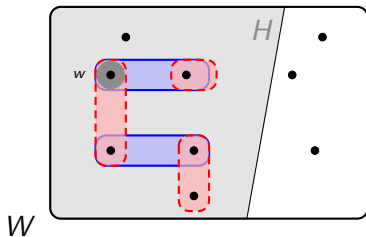


$$w \in C(H)$$

**Fact.**  $w \in C(H)$  if every **finite path** starting at  $w$  ends in a state in  $H$

There is a finite path from  $w$  to  $v$  if there is  $v_1, \dots, v_m$  such that  $w = v_1$ ,  $v = v_m$  and there are  $E_1, \dots, E_{m-1} \in \cup \mathcal{E}_i$  such that  $\{v_1, v_2\} \subseteq E_1, \dots, \{v_{m-1}, v_m\} \subseteq E_{m-1}$ .

$I_C(w) = \{v \mid \text{there is a finite path from } w \text{ to } v\}$ , so  
 $C(H) = \{w \mid I_C(w) \subseteq H\}$ .



**Fact.**  $w \in C(H)$  if every **finite path** starting at  $w$  ends in a state in  $H$

**Theorem.** Suppose that  $n$  agents share a common prior and have different private information. If there is common knowledge in the group of the posterior probabilities, then the posteriors must be equal.

Robert Aumann. *Agreeing to Disagree*. *Annals of Statistics* **4** (1976).

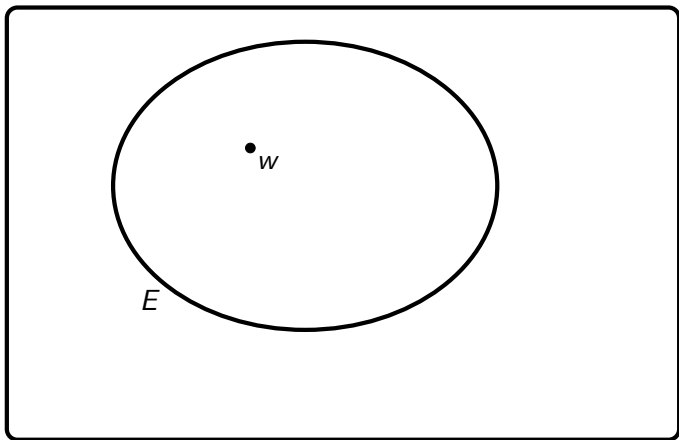
Suppose that  $W$  is,  $E \subseteq W$  is an event, and two (or more) agents with partitions  $\mathcal{E}_i$ . Let  $p$  be the **common prior**.

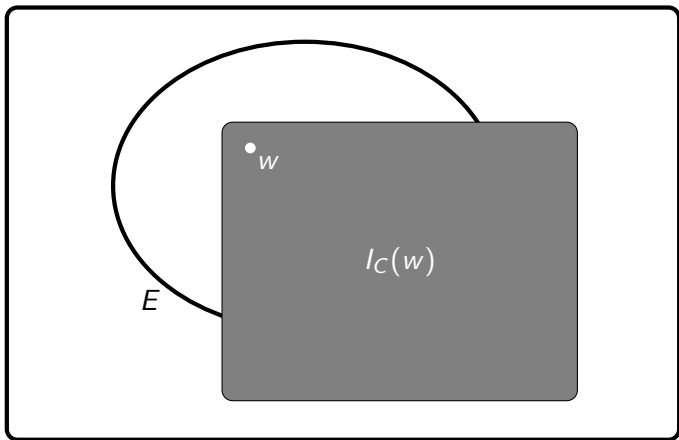
The agent's posterior probabilities of the event  $E$  are *random variables*:  $P_i^E : W \rightarrow [0, 1]$ ,  $P_i^E(w) = p(E \mid \mathcal{E}_i[w])$ .

So,  $\llbracket P_i^E = r \rrbracket = \{w \mid P_i^E(w) = r\}$

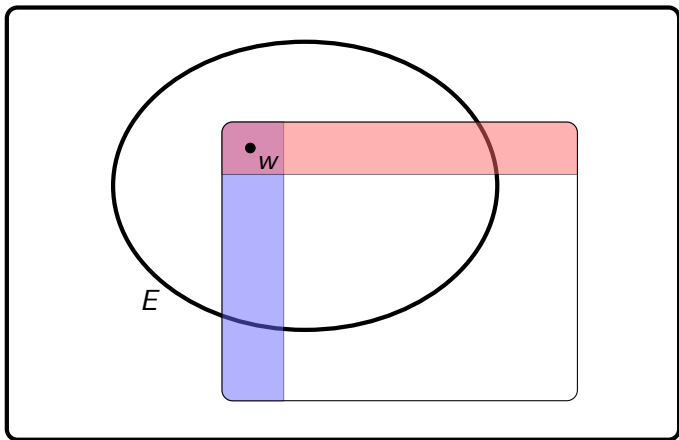
Assume that  $w \in C(\llbracket P_1^E = r \wedge P_2^E = q \rrbracket)$ .



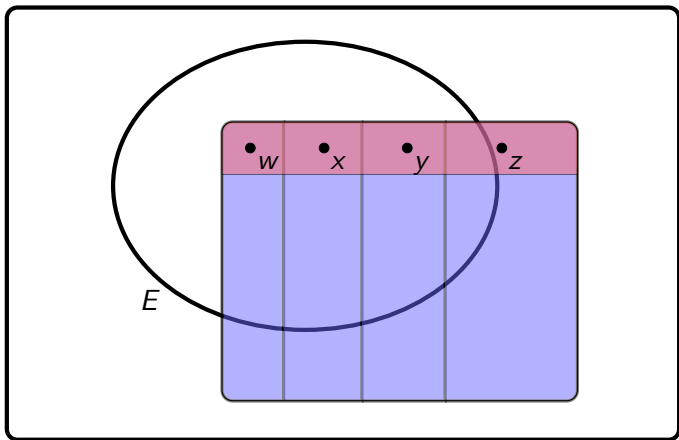




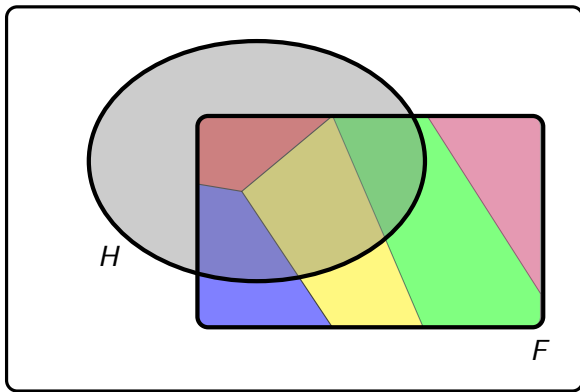
$$I_C(w) \subseteq \llbracket P_1^E = r \wedge P_2^E = q \rrbracket$$



$$p(E \mid \mathcal{E}_1[w]) = q, p(E \mid \mathcal{E}_2[w]) = r$$



$$p(E \mid \mathcal{E}_1[w]) = p(E \mid \mathcal{E}_1[x]) = p(E \mid \mathcal{E}_1[y]) = p(E \mid \mathcal{E}_1[z]) = q$$



$$p(H | F) = \sum_i p(E_i | F)p(H | E_i)$$

**Fact.** If  $p(H | E_i) = q$  for all  $i$ , then  $p(H | F) = q$ .

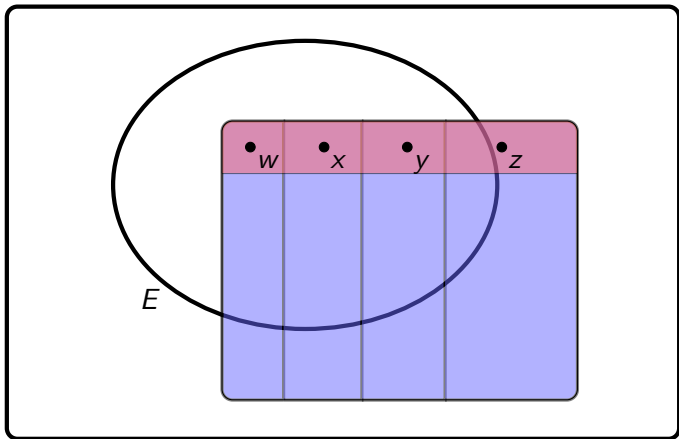
**Fact.** Suppose that  $\mathcal{E} = \{E_1, \dots, E_m\}$  partitions  $F$ . If  $p(H | E_i) = q$  for all  $i$ , then  $p(H | F) = q$ .

$$\begin{aligned} p(H | F) &= \sum_i p(E_i | F)p(H | E_i) \\ &= \sum_i p(E_i | F)q \\ &= q \sum_i p(E_i | F) \\ &= q \end{aligned}$$

**Fact.** Suppose that  $\{F_i\}$  is a partition of  $F$  (so  $F = \bigcup_i F_i$  and  $F_i \cap F_j \neq \emptyset$  for  $i \neq j$ ). If  $p(E | F_i) = q$  for all  $i$ , then  $p(E | F) = q$ .

If  $p(E | F_i) = q$ , then  $p(E \cap F_i) = qp(F_i)$ .

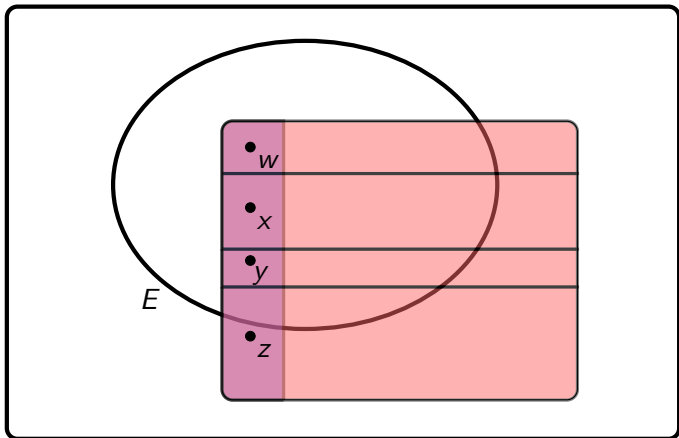
$$\begin{aligned} p(E | F) &= \frac{p(E \cap F)}{p(F)} = \frac{p((E \cap F_1) \cup \dots \cup (E \cap F_n))}{p(F)} \\ &= \frac{p(E \cap F_1) + \dots + p(E \cap F_n)}{p(F)} = \frac{qp(F_1) + \dots + qp(F_n)}{p(F)} \\ &= \frac{q(p(F_1) + \dots + p(F_n))}{p(F)} = \frac{qp(F)}{p(F)} = q \end{aligned}$$



$$p(E \mid \mathcal{E}_1[w]) = p(E \mid \mathcal{E}_1[x]) = p(E \mid \mathcal{E}_1[y]) = p(E \mid \mathcal{E}_1[z]) = q$$

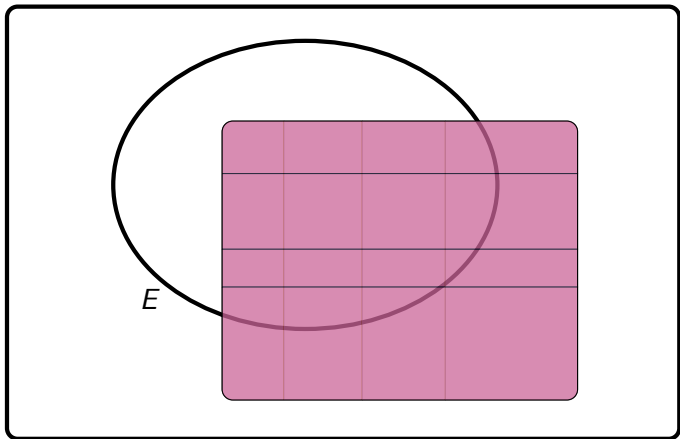
$$\text{So, } p(E \mid I_C(w)) = q.$$





$$p(E \mid \mathcal{E}_2[w]) = p(E \mid \mathcal{E}_2[x]) = p(E \mid \mathcal{E}_2[y]) = p(E \mid \mathcal{E}_2[z]) = r$$

So,  $p(E \mid I_C(w)) = r$ .



Thus,  $q = p(E \mid I_C(w)) = r$ .

## Qualitative versions

*like-minded individuals cannot agree to make different decisions.*

M. Bacharach. *Some Extensions of a Claim of Aumann in an Axiomatic Model of Knowledge*. Journal of Economic Theory, 37(1), pgs. 167-190, 1985.

J.A.K. Cave. *Learning to Agree*. Economic Letters, 12(2), pgs. 147 - 152, 1983.

D. Samet. *Agreeing to disagree: The non-probabilistic case*. Games and Economic Behavior, 69, pgs. 169-174, 2010.

# Rational Disagreement

M. Caie. *Agreement Theorems for Self-Locating Belief*. Review of Symbolic Logic, 2016.

J. Halpern and W. Kets. *Ambiguous Language and Common Priors*. Games and Economic Behavior, 2014.

H. Lederman. *People with common priors can agree to disagree*. Review of Symbolic Logic, 8(1), pp. 1145, 2015.

A. Rubinstein and A. Wolinsky. *On the logic of 'agreeing to disagree' type results*. Journal of Economic Theory, 1, 184193, 1990.

# Dynamic characterization of Aumann's Theorem

- ▶ How do the posteriors *become* common knowledge?

J. Geanakoplos and H. Polemarchakis. *We Can't Disagree Forever*. Journal of Economic Theory (1982).

# Dynamic characterization of Aumann's Theorem

- ▶ How do the posteriors *become* common knowledge?

J. Geanakoplos and H. Polemarchakis. *We Can't Disagree Forever*. Journal of Economic Theory (1982).

- ▶ What happens when communication is not the the whole group, but pairwise?

R. Parikh and P. Krasucki. *Communication, Consensus and Knowledge*. Journal of Economic Theory (1990).

$$t = 0 \quad \langle W, \mathcal{E}_{0,a}, \mathcal{E}_{0,b}, p \rangle$$

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$$P_{0,a}^E(w) = r_0 \quad P_{0,b}^E(w) = q_0$$



$$t = 0 \quad \langle W, \mathcal{E}_{0,a}, \mathcal{E}_{0,b}, p \rangle$$

$$P_{0,a}^E(w) = r_0 \quad P_{0,b}^E(w) = q_0$$

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$$P_{0,a}^E(w) = r_0 \quad P_{0,b}^E(w) = q_0$$

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$$P_{1,a}^E(w) = r_1 \quad P_{1,b}^E(w) = q_1$$

$$t = 0 \quad \langle W, \mathcal{E}_{0,a}, \mathcal{E}_{0,b}, p \rangle$$

$$P_{0,a}^E(w) = r_0 \quad P_{0,b}^E(w) = q_0$$

$$t = 1 \quad \langle W, \mathcal{E}_{1,a}, \mathcal{E}_{1,b}, p \rangle$$

$$P_{1,a}^E(w) = r_1 \quad P_{1,b}^E(w) = q_1$$

$$t = 2 \quad \langle W, \mathcal{E}_{2,a}, \mathcal{E}_{2,b}, p \rangle$$

$$P_{2,a}^E(w) = r_2 \quad P_{2,b}^E(w) = q_2$$

$$t = 3 \quad \langle W, \mathcal{E}_{3,a}, \mathcal{E}_{3,b}, p \rangle$$

$$\vdots \quad \quad \quad \vdots$$

## Geanakoplos and Polemarchakis

- ▶ Assuming that the information partitions are finite, given an event  $A$ , the revision process converges in finitely many steps.

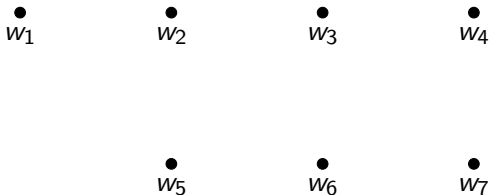
# Geanakoplos and Polemarchakis

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- ▶ For each  $n$ , there are examples where the process takes  $n$  steps.

## Geanakoplos and Polemarchakis

- ▶ Assuming that the information partitions are finite, given an event  $A$ , the revision process converges in finitely many steps.
- ▶ For each  $n$ , there are examples where the process takes  $n$  steps.
- ▶ An *indirect communication* equilibrium is not necessarily a *direct communication* equilibrium.

## 2 Scientists Perform an Experiment



They agree the true state is one of seven different states.

## 2 Scientists Perform an Experiment

$$\frac{2}{32} \bullet w_1$$

$$\frac{4}{32} \bullet w_2$$

$$\frac{8}{32} \bullet w_3$$

$$\frac{4}{32} \bullet w_4$$

$$\frac{5}{32} \bullet w_5$$

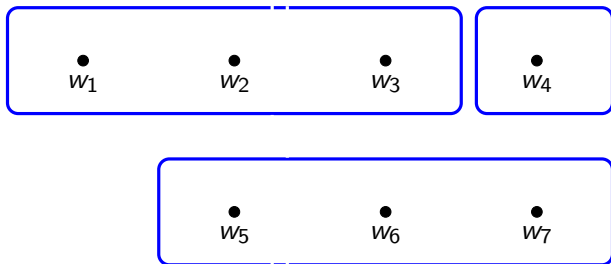
$$\frac{7}{32} \bullet w_6$$

$$\frac{2}{32} \bullet w_7$$

They agree on a common prior.

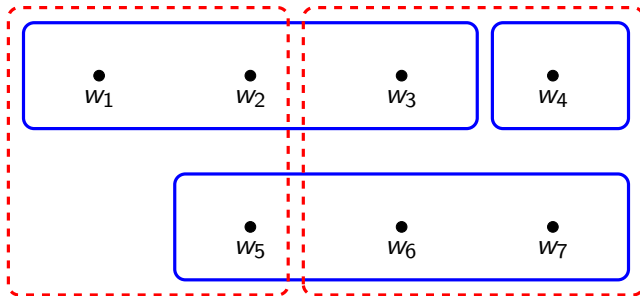


## 2 Scientists Perform an Experiment



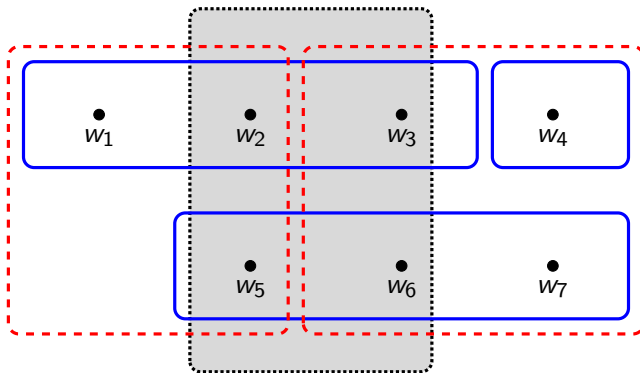
They agree that Experiment 1 would produce the blue partition.

## 2 Scientists Perform an Experiment



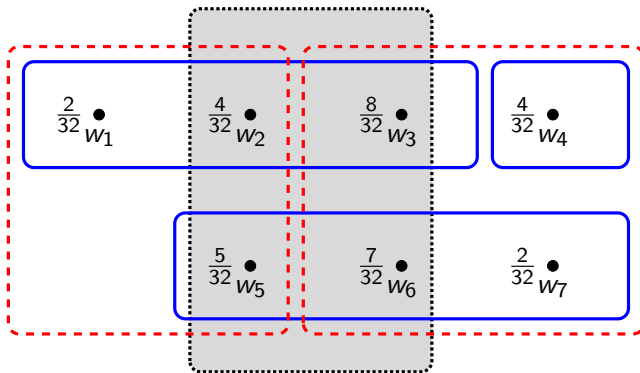
They agree that Experiment 1 would produce the blue partition and Experiment 2 the red partition.

## 2 Scientists Perform an Experiment



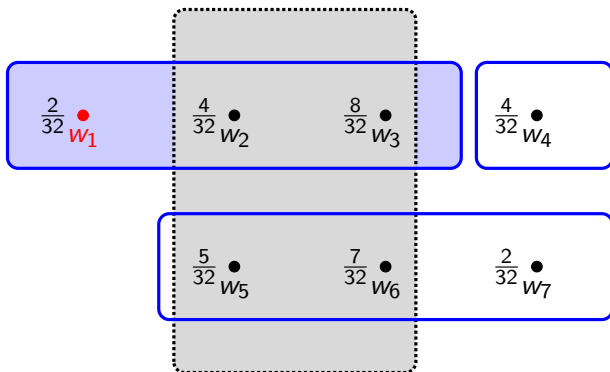
They are interested in the truth of  $E = \{w_2, w_3, w_5, w_6\}$ .

## 2 Scientists Perform an Experiment



So, they agree that  $P(E) = \frac{24}{32}$ .

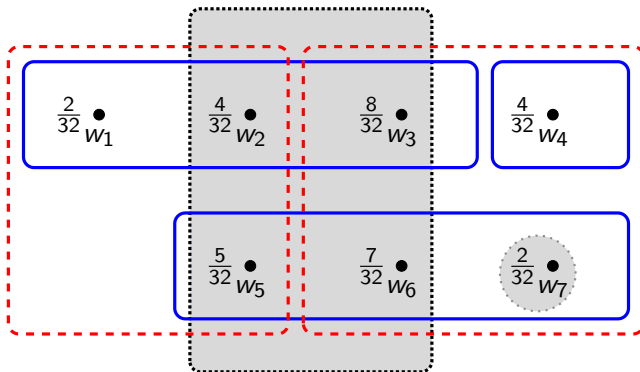
## 2 Scientists Perform an Experiment



Also, that if the true state is  $w_1$ , then Experiment 1 will yield

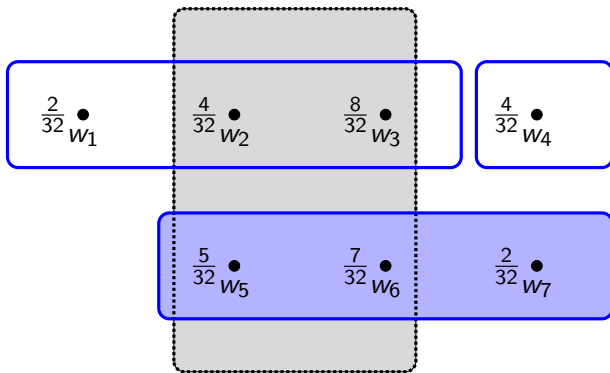
$$P(E|I) = \frac{P(E \cap I)}{P(I)} = \frac{12}{14}$$

## 2 Scientists Perform an Experiment



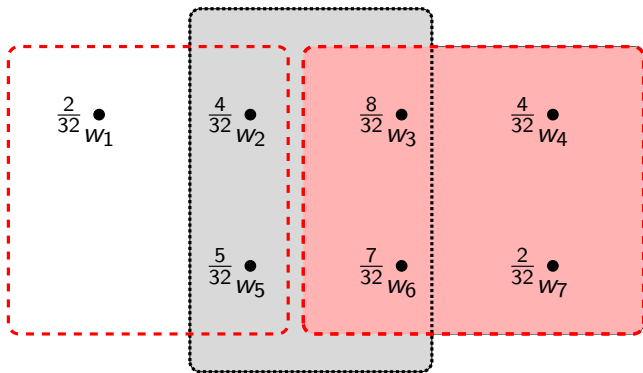
Suppose the true state is  $w_7$  and the agents perform the experiments.

## 2 Scientists Perform an Experiment



Suppose the true state is  $w_7$ , then  $Pr_1(E) = \frac{12}{14}$

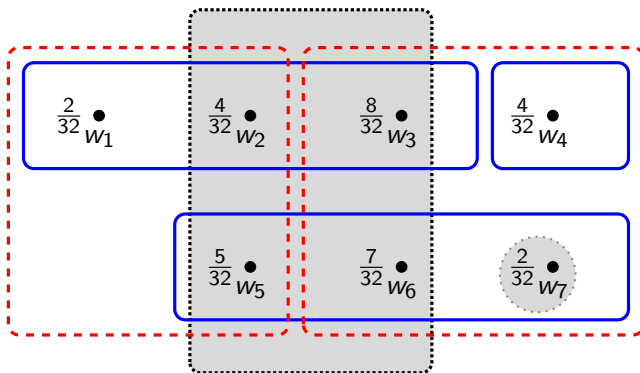
## 2 Scientists Perform an Experiment



Then  $Pr_1(E) = \frac{12}{14}$  and  $Pr_2(E) = \frac{15}{21}$

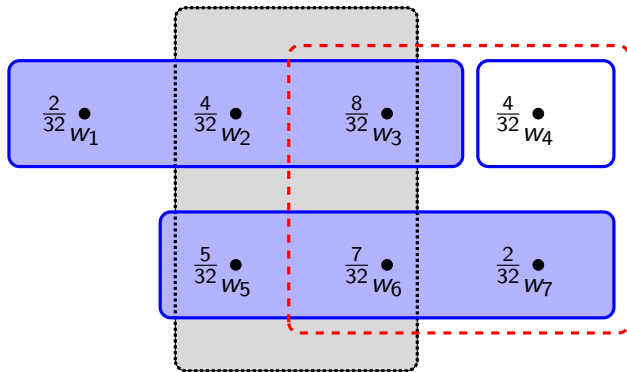


## 2 Scientists Perform an Experiment



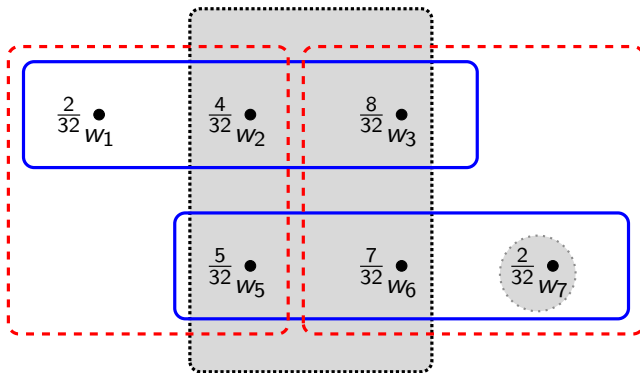
Suppose they exchange emails with the new subjective probabilities:  $Pr_1(E) = \frac{12}{14}$  and  $Pr_2(E) = \frac{15}{21}$

## 2 Scientists Perform an Experiment



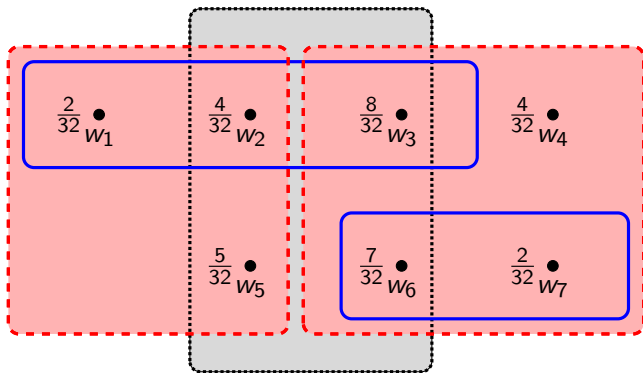
Agent 2 learns that  $w_4$  is **NOT** the true state (same for Agent 1).

## 2 Scientists Perform an Experiment



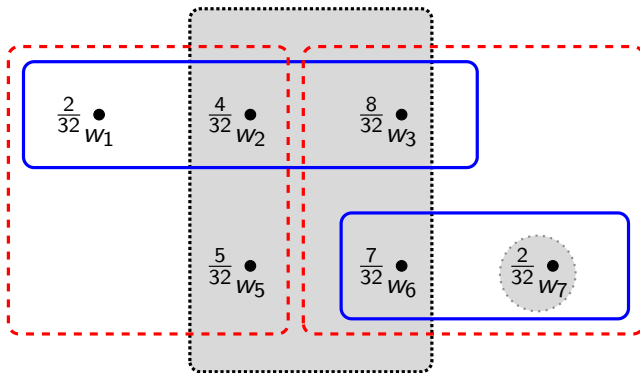
Agent 2 learns that  $w_4$  is **NOT** the true state (same for Agent 1).

## 2 Scientists Perform an Experiment



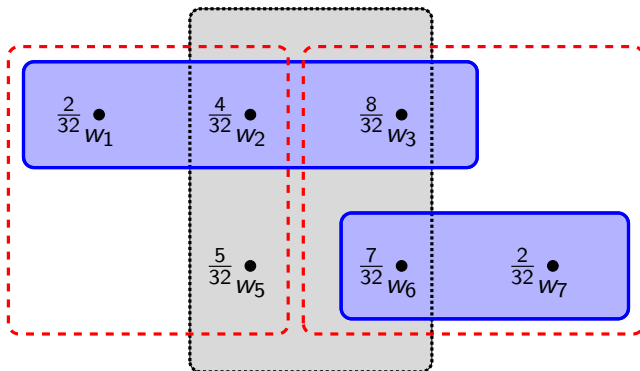
Agent 1 learns that  $w_5$  is **NOT** the true state (same for Agent 1).

## 2 Scientists Perform an Experiment



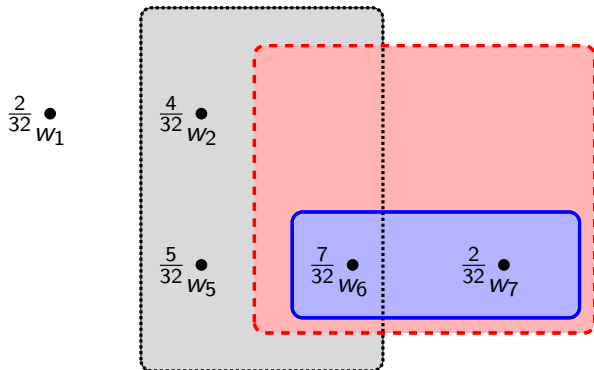
The new probabilities are  $Pr_1(E|I') = \frac{7}{9}$  and  $Pr_2(E|I') = \frac{15}{17}$

## 2 Scientists Perform an Experiment



After exchanging this information ( $Pr_1(E|I') = \frac{7}{9}$  and  $Pr_2(E|I') = \frac{15}{17}$ ), Agent 2 learns that  $w_3$  is **NOT** the true state.

## 2 Scientists Perform an Experiment



No more revisions are possible and the agents agree on the posterior probabilities.

# Rational Disagreement

M. Caie. *Agreement Theorems for Self-Locating Belief*. Review of Symbolic Logic, 2016.

J. Halpern and W. Kets. *Ambiguous Language and Common Priors*. Games and Economic Behavior, 2014.

H. Lederman. *People with common priors can agree to disagree*. Review of Symbolic Logic, 8(1), pp. 1145, 2015.

A. Rubinstein and A. Wolinsky. *On the logic of 'agreeing to disagree' type results*. Journal of Economic Theory, 1, 184193, 1990.



# Rational disagreement over time

D. Blackwell and L. Dubins. *Merging of opinions with increasing information*. The Annals of Mathematical Statistics, 33, pp. 882 - 886.

D. Samet and D. Monderer. *Stochastic Common Learning*. Games and Economic Behavior, 9, pgs. 161 - 171, 1995.

S. Huttegger. *Merging of Opinions and Probability Kinematics*. Review of Symbolic Logic, 2015.

D. Blackwell and L. Dubins. *Merging of opinions with increasing information*.  
The Annals of Mathematical Statistics, 33, pp. 882 - 886.

Suppose that  $W$  is a set of atomic events and  $\mathcal{F}$  is a  $\sigma$ -field on  $W$ .

The elements of  $W$  can be thought of as possible worlds and the members of  $\mathcal{F}$  as propositions.

E.g.,  $W$  can be the set of all infinite sequences of coin tosses and  $\mathcal{F}$  contains all propositions about coin tossing events of interest. (It may include limiting events such as  $\lim_{n \rightarrow \infty} S_n = 1/2$  where  $S_n$  is the total number of heads in the first  $n$  flips of the coin.)

Let  $\mathcal{E}_1, \mathcal{E}_2, \dots, \mathcal{E}_n, \dots$  be an infinite sequence of partitions on  $W$  such that  $\mathcal{E}_{n+1}$  *refines*  $\mathcal{E}_n$ .

$\mathcal{E}_n$  is the information that the agent receives at time  $n$ .

$\mathcal{E}_n[w]$  is the element of  $\mathcal{E}_n$  containing  $w$ .

For each  $n$ , let  $\mathcal{F}_n$  be the  $\sigma$ -algebra generated by  $\mathcal{E}_n$ . We assume that  $\mathcal{F} = \cup_n \mathcal{F}_n$ .

$$\text{For } A \in \mathcal{F}, P(A \mid \mathcal{E}_n[w]) = \frac{P(A \cap \mathcal{E}_n[w])}{P(\mathcal{E}_n[w])}$$

$$P(A \mid \mathcal{E}_n[w]) \rightarrow \begin{cases} 1 & w \in A \\ 0 & w \notin A \end{cases}$$

for **almost every**  $w$  with regard to the prior probability  $P$ .

The set of  $w$  for which the above does not hold has measure 0.

Convergence to certainty yields a first pass on merging of opinions.

Suppose that Eve's degrees of beliefs are represented by  $P$  and Adam's by  $Q$ , and let  $P_n[A](w) = P(A \mid \mathcal{E}_n[w])$  and  $Q_n[A](w) = Q(A \mid \mathcal{E}_n[w])$ .

Then  $P_n[A]$  and  $Q_n[A]$  both converge to zero or to one almost surely with respect to the priors  $P$  and  $Q$ , respectively.

Now, Eve believes with certainty that  $P_n[A]$  and  $Q_n[A]$  will agree in the limit *whenever she assigns probability one to any set to which Adam assigns probability one.*

For then, since  $Q_n[A]$  goes to certainty a.e. ( $Q$ ), Eve also believes with probability one that her's and Adam's conditional probabilities for any event  $A$  are the same in the limit.



This result applies to particular events  $A$ . But it does not say anything about Eve's and Adam's overall conditional probabilities.

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The Blackwell-Dubins theorem fills this gap.

As Adam and Eve observe more coin tosses and update by Bayesian conditioning, their degrees of belief will become close *uniformly* in all events.

Moreover, the Blackwell-Dubins theorem does not require that conditional probabilities converge (as in convergence to certainty). Eve's and Adam's conditional probabilities may get closer even if they don't converge.

## Variational Distance

Suppose that  $\mu$  and  $\nu$  are two measures over all events in  $\mathcal{F}$ .

$$d(\mu, \nu) = \sup_{A \in \mathcal{F}} |\mu(A) - \nu(A)|$$

## Variational Distance

Suppose that  $\mu$  and  $\nu$  are two measures over all events in  $\mathcal{F}$ .

$$d(\mu, \nu) = \sup_{A \in \mathcal{F}} |\mu(A) - \nu(A)|$$

$P$  is said to **merge to**  $Q$  if for  $Q$  almost every  $w$ ,  
 $d(P_n(w), P_n(w)) \rightarrow 0$  as  $n \rightarrow \infty$ .

Almost everywhere (a.e.  $Q$ ), given any  $\epsilon > 0$ , there is an  $n_0$  such that

$$|P(A \mid \mathcal{E}_n[w]) - Q(A \mid \mathcal{E}_n[w])| < \epsilon$$

for all  $n > n_0$  and for all  $A \in \mathcal{F}$ . The number  $n_0$  may depend on  $\epsilon$  and on  $w$  but not on  $A$ .

If  $P$  merges to  $Q$ , then Adam (with probability  $Q$ ) believes with probability one that his conditional degrees of belief for all propositions  $A$  will get arbitrarily close.

# Absolute Continuity

$Q$  is absolutely continuous relative to  $P$  ( $Q \ll P$ ) if for all  $A \in \mathcal{F}$ ,

$$Q(A) > 0 \implies P(A) > 0$$

**Blackwell & Dubbins Theorem.** If  $Q \ll P$ , then  $P$  merges to  $Q$ .

S. Huttegger. *Merging of Opinions and Probability Kinematics*. Review of Symbolic Logic, 2015.



In general, no set of initial beliefs is more, or less, justified than another. This implies that conditioning on the same information can lead two agents to have different posterior beliefs.

...disagreement often turns out to be transient and disappears as one gets more information. In other words, as more evidence becomes available a consensus may emerge...diverging opinions are just a sign that not enough evidence has accumulated.

If long run consensus fails we may be able to trace it back, not to the individual irrationality of an agent, but to the fact that the agents' initial beliefs did not observe the absolute continuity requirement.

If the absolute continuity requirement holds, another possible explanation is that merging takes place very slowly; beliefs would merge if agents were given more evidence, but this might not always be possible. The long run may be too long.

Agents assumes to undergo learning experiences that yield a sequence  $P_1, P_2, \dots$  of probability measures on  $(W, \mathcal{F}_1), (W, \mathcal{F}_2), \dots$

Each agent believes with probability one that she will revise her probabilities by performing *probability kinematics* with  $p_1, p_2, \dots$ . Each probability measure  $P_n$  is fully determined by attaching probability values to members of  $\mathcal{E}_n$ .

$$P_n(A) = \sum_{E \in \mathcal{E}_n} P(A | E) p_n(E)$$

(AC) If  $P_n \ll P|_{\mathcal{F}_n}$ , then for every  $\epsilon > 0$ , there is a  $\delta_n > 0$  such that

$$P(B) < \delta_n \implies P_n(B) < \epsilon$$

for all  $B \in \mathcal{F}_n$ .

## Uniformly Absolutely Continuous

1.  $P_n \ll P|_{\mathcal{F}_n}$  for each  $n$  and
2. for every  $\epsilon > 0$  there is a  $\delta > 0$  such that for all  $n$ ,

$$P(B) < \delta \implies P_n(B) < \epsilon$$

for all  $B \in \mathcal{F}_n$

- ▶ Condition 1. requires that events which are assigned probability zero now are expected to have probability zero in the future.
- ▶ Condition 2. says that this also holds in the limit. Otherwise, there may be events  $F_1, F_2, \dots$  such that  $P(F_n) \rightarrow 0$  as  $n \rightarrow \infty$ , while the sequence  $P_1(F_1), P_2(F_2), \dots$  is bounded away from zero.

Suppose now that there are two prior probability measures  $P$  and  $Q$  which are updated successively by probability kinematics on  $\mathcal{E}_1, \mathcal{E}_2, \dots$  using the distributions  $p_1, p_2, \dots$  and  $q_1, \dots$ , respectively.

Using the Jeffrey update rule this leads to the new probability measures  $P_n$  and  $Q_n$  on  $\mathcal{F}$  for  $n \geq 1$ .

It is quite obvious that arbitrary choices of sequences  $p_1, p_2, \dots$  and  $q_1, q_2, \dots$  need not lead to merging. But this is also true for conditioning.

Recall that one requirement of the Blackwell-Dubins theorem is that agents condition on the *same factual evidence*.

Thus, the important question is whether beliefs merge for probability kinematics whenever  $p_n$  and  $q_n$  represent the same uncertain information.

But what does it mean to get the same uncertain evidence?



## Hard Jeffrey Shift

A **hard Jeffrey shift** sets values for  $p_n$  regardless of the prior probability  $P_{n-1}$ , and so may destroy any information about the partition that was encoded in the prior.

As an example, consider a measurement instrument that makes noisy observations of a physical process, such as coin flips. Let's call this setup a 'mechanical observer'.

The probability space  $(W, \mathcal{F})$  represents the set of states and events of the process. At each stage  $n$ , the output of the mechanical observer is a probability distribution over the partition  $\mathcal{E}_n$ .

The probabilities for members of the partition are determined by repeated previous observations under symmetric conditions in order to specify measurement error.

More generally, a hard Jeffrey shift can be viewed as a noisy signal where the noise has the form of a probability distribution over a partition such that the distribution is known to every observer.

In terms of hard Jeffrey shifts, having the same uncertain evidence at stage  $n$  means that  $p_n = q_n \mathcal{F}_n$ .

Suppose, for example, that Adam and Eve are two scientists observing coin flips with the help of a mechanical measurement instrument. They might not feel comfortable approximating their learning process by conditionalization if their measurements are not precise enough. Instead, they plan to update by the same hard Jeffrey shifts at each stage of their experiment.

Can they be certain to have similar beliefs after having taken

$$(M) \quad P_n(F) = P_{n-1}(F) \text{ for all } F \in \mathcal{F}_{n-1}$$

**Theorem** (Hutteger). Suppose that  $q_n = p_n$ , that the sequence  $Q_n$ ,  $n = 1, 2, \dots$  is uniformly absolutely continuous relative to  $Q$ , and that  $Q \ll P$ . If condition (M) holds, then  $d(P_n, Q_n) \rightarrow 0$  and  $n \rightarrow \infty$ .

## Soft Jeffrey Shifts

$\mathcal{E}_1 = \{E_1, E_2, E_3, E_4\}$  and consider the learning experience given by:

$$\left(\frac{1}{5} : E_1, \frac{3}{10} : E_2, \frac{1}{2} : E_3, 0 : E_n\right)$$

## Soft Jeffrey Shifts

$\mathcal{E}_1 = \{E_1, E_2, E_3, E_4\}$  and consider the learning experience given by:

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Consider instead:

- ▶  $p_1(E_1) = 2 \cdot P(E_1)$ ;
- ▶  $p_1(E_2) = \frac{1}{2} \cdot P(E_2)$ ;
- ▶  $p_1(E_3) = 5 \cdot P(E_3)$ ; and
- ▶  $p_1(E_4) = 0 \cdot P(E_4)$

If  $P(E_1) = \frac{1}{10}$ ,  $P(E_2) = \frac{3}{5}$ ,  $P(E_3) = \frac{1}{10}$  and  $P(E_4) = \frac{1}{5}$ , then probability kinematics will lead to the same result whether or not it is a hard or “soft” Jeffrey shift.



Do beliefs merge when agents have the same soft uncertain evidence?

We are going to see that this need not be the case. If Adam and Eve start with different (but mutually absolutely continuous) prior probabilities for infinite sequences of coin flips, and if both observe principle (M) as well as undergo the same soft Jeffrey shifts, their posterior degrees of beliefs may not get close to each other in the long run. [Theorem 6.3 in Hutteger]

“Our results lead to the conclusion that, even under otherwise favorable circumstances, a soft kind of information allows individual rationality to be consistent with sustained disagreement. I don’t think that this is a weakness of the broadly Bayesian approach advocated in this essay. Merging of beliefs happens when it should, i.e., under conditions which may, for example, hold for certain carefully designed scientific investigations. But the claim of merging is not a no-brainer that can be used across the board.”  
(Hutteger)

R. G. Fryer, P. Harms and M. Jackson. *Updating Beliefs When Evidence is Open to Interpretation: Implications for Bias and Polarization*. manuscript, 2017.

- ▶ State:  $\{A, B\}$
- ▶ Signals:  $\{a, b, ab, \emptyset\}$ 
  - $a$  is a signal for state  $A$
  - $b$  is a signal for state  $B$
  - $ab$  is an ambiguous signal
  - $\emptyset$  represents the lack of a signal
- ▶  $\lambda_0$ : prior probability
- ▶  $q$ : probability of observing a signal
- ▶  $p$ : probability that an agent receives signal  $a$  if the state is  $A$  and  $b$  if the state is  $B$
- ▶  $\pi$ : probability that a signal "becomes" ambiguous.

**Theorem.** A Bayesian-updating agent who forms beliefs conditional upon the full sequence of signals has a posterior that converges to place probability 1 on the correct state, almost surely.

**Proposition.** Suppose that a nontrivial fraction of experiences are open to interpretation so that  $\pi > \frac{p-1/2}{p}$ . Consider two interpretative agents 1 and 2 who both use the **maximum likelihood rule** but have differing priors: agent 1's prior is that  $A$  is more likely (so 1 has a prior  $\lambda_0 > 1/2$ ) and agent 2's prior is that  $B$  is more likely (so 2 has a prior  $\lambda_0 < 1/2$ ). Let the two agents see exactly the same sequence of signals. With a positive probability that tends to 1 in  $\pi$  the two agents will end up polarized with 1's posterior tending to 1 and 2's posterior tending to 0. With a positive probability tending to 0 in  $\pi$  the two agents will end up with the same (possibly incorrect) posterior tending to either 0 or 1

# Plan

- ✓ Introduction to belief revision, AGM
- ✓ Possible worlds models, Bayesian models
- ✓ Updating probabilities, The value of learning, Lottery Paradox, Preface Paradox, Review Paradox
- ✓ Lottery Paradox, Preface Paradox, Review Paradox, Iterated belief revision, Context shifts, Becoming aware
- ✓ Iterated belief revision, Interactive epistemology (Agreement Theorems), Convergence Theorems

Thank You!

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[pacuit.org/nls2017/belchange/](http://pacuit.org/nls2017/belchange/)