

# A bayesian method of cosine-based word2vec bias estimation

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Represent uncertainty by a single probabilistic measure (point estimates).

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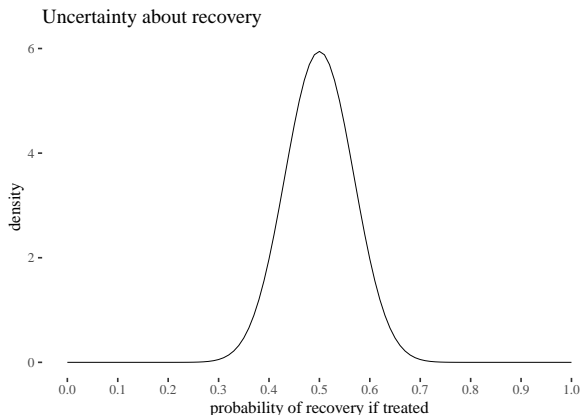
## Epistemological difficulties of imprecise probabilism

- hard to model the weight of and combine evidence.
- inertia of agnosticism.
- growth of uncertainty in light of new evidence.

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## My novel contribution

- Meet the desiderata motivating imprecision.
- Avoid these difficulties.
- Do so by using higher-order probabilities.



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- You've seen 4/10 patients recover.  $P(\text{recovery}) = .4?$
- You've seen 40/100 patients recover.  $P(\text{recovery}) = .4?$

(You could try to use classical statistical tools, such as confidence intervals, but they have their own problems.)

# Motivations for imprecision (uniformity and agnosticism)

## Agnosticism preservation requirement

Agnosticism about uniquely connected variables should be the same.

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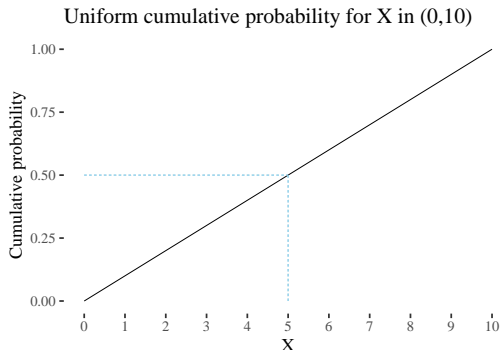
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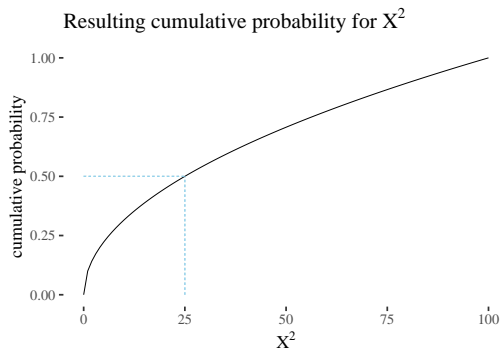
## The trouble



$$P(X < 5) = .5 = P(X^2 < 25)$$

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# Motivations for imprecision (expert evidence)

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Al:  $P(\text{It will rain tomorrow}) = .6$

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## Mathematical impossibilities

- Experts's agreement on independence is not preserved by linear pooling. (???)
- You can't at the same time hold the following: (???)

$$P(A = B) < 1 \tag{1}$$

$$P(r|A = a) = a \tag{2}$$

$$P(r|B = b) = b \tag{3}$$

$$\forall a, b \, P(r|A = a, B = b) = \alpha a + \beta b \tag{4}$$



# Imprecise probabilism and the motivations

## The main claim

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## Multiple sources?

Just put all those probabilities in one set.

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## Radical uncertainty and weight

Say one experiment observed 40/100 and another 8/15 recoveries.  
Simply putting the frequencies in a set won't do justice to the evidence.  
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## Dilation

Whenever your roommate goes to the track he flips a fair coin. If it comes up heads he bets on Speedy. Otherwise he places no bets. He comes home from the track grinning. You're now more uncertain about  $P(\text{heads})$  than before.

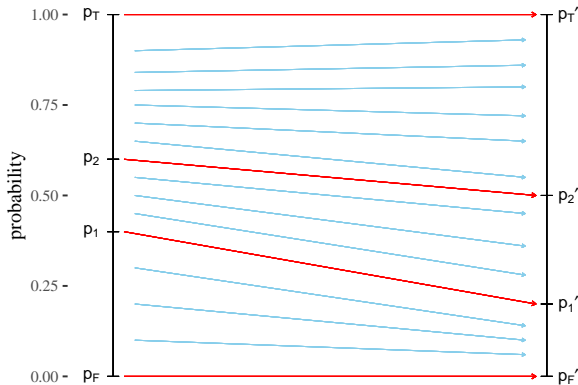
(???, ???, ???, ???, ???, ???; Walley, 1991)



# Epistemological challenges

## Belief inertia

Say you're fully agnostic about  $H$ :  $P(H) = [0, 1]$ .



Then  $P(H|E) = [0, 1]$ , no matter what the evidence says.

(???,(???), ???, ???, ???, ???)

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## Question

As far as accuracy is concerned, why prefer the imprecise way?

Say the actual bias is .4. A first stab:

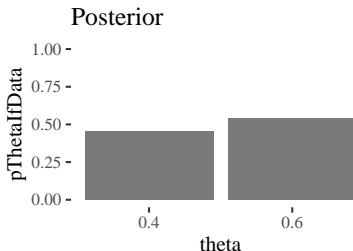
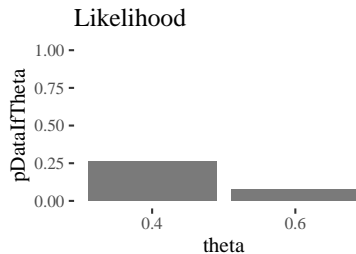
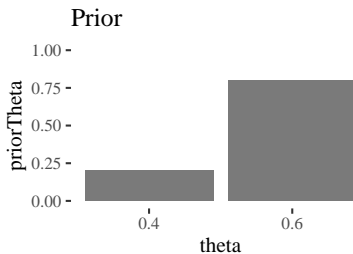
$$\text{inaccuracy}(\text{imprecise}) = \frac{0 + .2}{2} = .1$$

$$\text{inaccuracy}(\text{precise}) = .1$$

(???, ???; Schoenfield, 2017)

# Higher-order approach

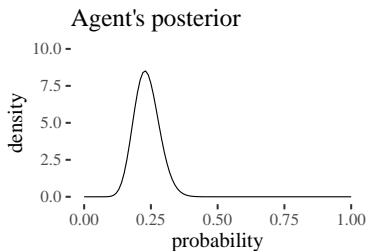
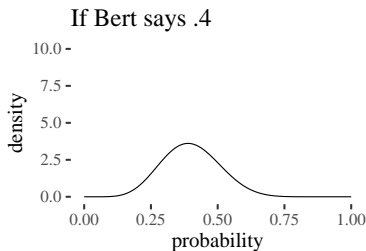
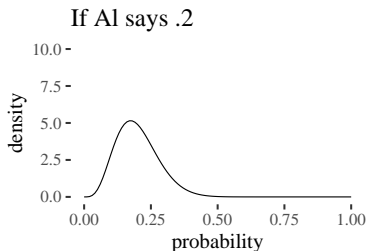
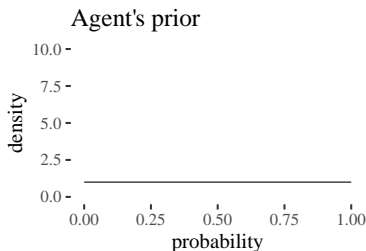
The mystery coin with (.2, .8) and 1 head in five tosses



theta	0.4	0.6
priorTheta	0.2	0.8
pData	0.113	0.113
pDataIfTheta	0.259	0.077
pThetaIfData	0.458	0.542

# Higher-order approach

## Two experts, uniform prior



# The key conjecture

With higher-order methods, it will be possible to:

- match the representation of uncertainty with the evidence,
- combine various sources of uncertainty without running into the known problems,
- use existing mathematics to make sense of accuracy in such contexts,
- capture reasoning with radical uncertainty without dilation and belief inertia.



# Tasks (discussion)

- Relate this representation to weight of evidence.
- Go beyond grid approximation and beta representation.
- Represent agnosticism so that:
  - agnosticism preservation holds, and
  - inertia fails.
- Represent opinion aggregation more generally, compare to known limitations and features of already studied opinion pooling methods.
- Investigate dilation-like phenomena in this context.
- Study known distance functions between probabilistic measures as candidates for key elements of accuracy measures.
- Explain away the apparent contrast between evidence-grounding and accuracy considerations (Easwaran & Fitelson, 2015).

Thank you for your attention!

# References

Easwaran, K., & Fitelson, B. (2015). Accuracy, coherence, and evidence. In T. S. Gendler & J. Hawthorne (Eds.), *Oxford studies in epistemology* (Vol. 5). Oxford University Press.

Schoenfield, M. (2017). The Accuracy and Rationality of Imprecise Credences: The Accuracy and Rationality of Imprecise Credences. *Noûs*, 51(4), 667–685. <https://doi.org/10.1111/nous.12105>

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