## Combining Intuition and Data

John D. Cook Singular Value Consulting

September 11, 2014

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### Why combine intuition and data?

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• Either one alone can lead to nonsense.

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- It's not really possible to separate them.
- Better to be explicit about it.

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- You must use all evidence.
- If evidence increases your belief that A is true, the same evidence must decrease your belief that A is false.
- If new evidence increases your belief in *A*, but does not change your belief in *B*, it must increase your belief in *AB*.

# It's all probability

P

- Plausibility of A
- Degree of belief in A
- Probability of A
- *P*(*A*)

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# Learning

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$$P(AB \mid C) = P(A \mid C) \cdot P(B \mid AC) = P(B \mid C) \cdot P(A \mid BC)$$

- Learning in any order leads to the same place.
- $P(AB \mid C) = P(A \mid C) \cdot P(B \mid AC) = P(B \mid C) \cdot P(A \mid BC)$
- This is essentially Bayes theorem.

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### From Bayes theorem to Bayesian Statistics

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### From Bayes theorem to Bayesian Statistics

• Create a model of what you want to study.

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- Let  $\theta$  be the parameter(s).

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- Let  $\theta$  be the parameter(s).
- $P(\theta \mid \mathsf{data}) \propto P(\mathsf{data} \mid \theta) \cdot P(\theta)$

# Priors

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• You must start with a prior.

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- You must start with a prior.
- You always know something before you collect data.

- You must start with a prior.
- You always know *something* before you collect data.
- It's not necessary, or possible, to be completely precise.

## The magic of Bayes

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• Use *everything* you know, intuition and data.

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- Use everything you know, intuition and data.
- The impact of the prior automatically fades as data arrive.
- All inference follows the same framework. No adhockery.
- Results are easy to interpret.

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• Is an undefeated player better than one who has been defeated?

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- Two wins and no losses vs. 90 wins and 10 losses

- Is an undefeated player better than one who has been defeated?
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- Probability Theory: The Logic of Science by E. T. Jaynes
- http://tinyurl.com/bayes-backgammon
- Contact info: http://JohnDCook.com