

Reasoning for Humans: Clear Thinking in an Uncertain World

PHIL 171

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$$X \rightarrow Y, Z \rightarrow Y \models (X \vee Z) \rightarrow Y$$

$$X \rightarrow Y, \neg X \rightarrow Y \models Y$$

If $X \models Y$ and $\neg X \models Y$, then Y is a tautology

“ P evidentially supports Q given R ” means the same thing as
“ $P \wedge R$ evidentially supports Q ”

P evidentially supports Q given R $Pr(Q | P \wedge R) > 0.5$

$P \wedge R$ evidentially supports Q $Pr(Q | P \wedge R) > 0.5$

“ P is positively relevant to Q given R ” does not mean the same thing as
“ $P \wedge R$ is positively relevant to Q ”

P is positively relevant to Q given R $Pr(Q | P \wedge R) > Pr(Q | R)$

$P \wedge R$ is positively relevant to Q $Pr(Q | P \wedge R) > Pr(Q)$

Sure-thing Principle: If

1. P is “good evidence” for Q given R , and
2. P is “good evidence” for Q given $\neg R$

then P is “good evidence” for Q

Unconditional Sure-thing Principle: If

1. $P \wedge R$ is “good evidence” for Q and
2. $P \wedge \neg R$ is “good evidence” for Q

then P is “good evidence” for Q

Sure-thing Principle: Evidential Support

In any stochastic truth table, if

1. $Pr(Q | P \wedge R) > 0.5$; and
2. $Pr(Q | P \wedge \neg R) > 0.5$

then $Pr(Q | P) > 0.5$

(assuming that all the conditional probabilities are defined)

Unconditional Sure-thing Principle: Positive Support

In any stochastic truth table, if

1. $Pr(Q | P \wedge R) > Pr(Q)$; and
2. $Pr(Q | P \wedge \neg R) > Pr(Q)$

then $Pr(Q | P) > Pr(Q)$

(assuming that all the conditional probabilities are defined)

Sure-thing Principle: Positive Support

There is a stochastic truth table such that

1. $Pr(Q | P \wedge R) > Pr(Q | R)$; and
2. $Pr(Q | P \wedge \neg R) > Pr(Q | \neg R)$

but $Pr(Q | P) \leq Pr(Q)$

(assuming that all the conditional probabilities are defined)

| | <i>P</i> | <i>Q</i> | <i>R</i> | |
|-------|----------|----------|----------|-----------------------------------|
| 0.2 | T | T | T | |
| 0.125 | T | T | F | $Pr(Q P \wedge R) = 0.8$ |
| 0.05 | T | F | T | $Pr(Q R) \approx 0.789$ |
| 0.075 | T | F | F | $Pr(Q P \wedge \neg R) = 0.625$ |
| 0.4 | F | T | T | $Pr(Q \neg R) \approx 0.521$ |
| 0 | F | T | F | $Pr(Q P) \approx 0.722$ |
| 0.11 | F | F | T | $Pr(Q) = 0.725$ |
| 0.04 | F | F | F | |

Simpson's Paradox

Suppose that a University is hiring in Philosophy and Mathematics.
13 men and 13 women apply for jobs.

| | Men | Women | |
|-------------|-----|-------|--------------------------------------|
| Mathematics | 1/5 | 2/8 | <i>success rate better for women</i> |
| Philosophy | 6/8 | 4/5 | <i>success rate better for women</i> |

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| Mathematics | 1/5 | 2/8 | <i>success rate better for women</i> |
| Philosophy | 6/8 | 4/5 | <i>success rate better for women</i> |
| University | 7/13 | 6/13 | <i>success rate better for men</i> |

How can it be that each Department favors women applicants, and yet overall men fare better than women?

G. Malinas and J. Bigelow. *Simpson's Paradox*. The Stanford Encyclopedia of Philosophy, Fall 2012 Edition, Edward N. Zalta (ed.).

Summary

- If $X \models Y$ and $\neg X \models Y$, then $\models Y$.
- Both the Sure-thing Principle and the Unconditional Sure-Thing Principle holds for evidential support:

If $Pr(Q | P \wedge R) > 0.5$ and $Pr(Q | P \wedge \neg R) > 0.5$, then
 $Pr(Q | P) > 0.5$

- The Sure-thing Principle does not hold for positive relevance:

There are stochastic truth tables where: $Pr(Q | P \wedge R) > Pr(Q | R)$
and $Pr(Q | P \wedge \neg R) > Pr(Q | \neg R)$, but $Pr(Q | P) \leq Pr(Q)$

- The Unconditional Sure-Thing Principle holds for positive relevance:

If $Pr(Q | P \wedge R) > Pr(Q)$ and $Pr(Q | P \wedge \neg R) > Pr(Q)$, then
 $Pr(Q | P) > Pr(Q)$

Bayes Theorem

Conditional Probability

The probability of H given E , $Pr(H | E)$, is defined to be

$$Pr(H | E) = \frac{Pr(H \wedge E)}{Pr(E)}.$$

provided $Pr(E) > 0$.

$$Pr(H | E) = Pr(E | H) \frac{Pr(H)}{Pr(E)}$$

$$Pr(H | E) = Pr(E | H) \frac{Pr(H)}{Pr(E)}$$

Bayes theorem is important because it expresses the quantity $Pr(H | E)$ (the probability of a hypothesis H given the evidence E)—which is something people often find hard to assess—in terms of quantities that can be drawn directly from experiential knowledge.

Example: Suppose you are in a casino and you hear a person at the next gambling table announce “Twelve”. We want to know whether he was rolling a pair of dice or a roulette wheel.

That is, compare $Pr(\text{Dice} \mid \text{Twelve})$ with $Pr(\text{Roulette} \mid \text{Twelve})$.

Based on our background knowledge of gambling we have $Pr(\text{Twelve} \mid \text{Dice}) = 1/36$ and $Pr(\text{Twelve} \mid \text{Roulette}) = 1/38$.

Based on our observations about the casino, we can judge the prior probabilities $Pr(\text{Dice})$ and $Pr(\text{Roulette})$.

But this is now enough to *calculate* the required probabilities.

Posterior

Prior of H

$$Pr(H|E) = Pr(E|H) \frac{Pr(H)}{Pr(E)}$$

Likelihood

Prior of observing E

A diagram illustrating Bayes' theorem. The equation $Pr(H|E) = Pr(E|H) \frac{Pr(H)}{Pr(E)}$ is centered. Four labels with arrows point to parts of the equation: 'Posterior' points to $Pr(H|E)$; 'Likelihood' points to $Pr(E|H)$; 'Prior of H ' points to the numerator $Pr(H)$; and 'Prior of observing E ' points to the denominator $Pr(E)$.

$$Pr(H | E) = Pr(E | H) \frac{Pr(H)}{Pr(E)}$$

$$Pr(H | E) = Pr(E | H) \frac{Pr(H)}{Pr(H)Pr(E | H) + Pr(\neg H)Pr(E | \neg H)}$$

Bayes Theorem

Diagram illustrating Bayes Theorem with labels and arrows:

- Posterior** points to $Pr(H|E)$
- Likelihood** points to $Pr(E|H)$
- Prior of H** points to $Pr(H)$
- Prior of observing E** points to $Pr(E)$

$$Pr(H|E) = Pr(E|H) \frac{Pr(H)}{Pr(E)}$$

$$Pr(H | E) = Pr(E | H) \frac{Pr(H)}{Pr(E)}$$

$$Pr(H | E) = Pr(E | H) \frac{Pr(H)}{Pr(H)Pr(E | H) + Pr(\neg H)Pr(E | \neg H)}$$

Suppose you select one card from a standard deck of cards.

Q is “the card is a queen card” and F is “the card is a face card”.

What is $Pr(Q | F)$?

| | Q | F | $Q \wedge F$ |
|----------------|---|---|--------------|
| $\frac{1}{12}$ | T | T | T |
| $\frac{1}{12}$ | T | F | F |
| $\frac{1}{12}$ | F | T | F |
| $\frac{1}{12}$ | F | F | F |

$$Pr(Q | F) = \frac{Pr(Q \wedge F)}{Pr(F)} = \frac{\frac{1}{12}}{\frac{1}{6}} = \frac{1}{2} = \frac{1}{2}$$

| | Q | F | $Q \wedge F$ |
|-----------------|---|---|--------------|
| $\frac{4}{52}$ | T | T | T |
| 0 | T | F | F |
| $\frac{8}{52}$ | F | T | F |
| $\frac{40}{52}$ | F | F | F |

$$Pr(Q | F) = \frac{Pr(Q \wedge F)}{Pr(F)} = \frac{\frac{4}{52}}{\frac{12}{52}} = \frac{4}{12} = \frac{1}{3}$$

Note that:

$$Pr(F | Q) = 1$$

$$Pr(F) = \frac{12}{52}$$

$$Pr(Q) = \frac{4}{52}$$

$$Pr(Q | F) = Pr(F | Q) \frac{Pr(Q)}{Pr(F)} = 1 \times \frac{\frac{4}{52}}{\frac{12}{52}} = \frac{4}{12} = \frac{1}{3}$$