

Bayesian Inference

1. Frequentists vs. Bayesian
2. Bayes' Rule
3. Prior, likelihood, posterior distributions

What is the probability that it rained in my city last night?

(No info is given about which city I'm currently in.)

$$P(\text{rain}) = 0.1$$

What is the probability that it rained in my city last night given that I'm in Seattle?

$$P(\text{rain}|\text{Seattle}) = 0.65$$

What is the probability that it rained in my city last night?

(No info is given about which city I'm currently in.)

$$P(\text{rain}) = 0.1$$

What is the probability that it rained in my city last night given that I live in Seattle and I see that the road is wet?

$$P(\text{rain} | \text{Seattle, wet roads}) = 0.97$$

Frequentist vs. Bayesian

Frequentist Probability

“Long Run” frequency of an outcome

Subjective Probability

A measure of degree of belief

Bayesians consider both types

Experiment 1:

A fine classical musician says he's able to distinguish Haydn from Mozart.
Small excerpts are selected at random and played for the musician.
Musician makes 10 correct guesses in exactly 10 trials.



Experiment 2:

Drunken man says he can correctly guess what face of the coin will fall down, mid air.
Coins are tossed and the drunken man shouts out guesses while the coins are mid air.
Drunken man correctly guesses the outcomes of the 10 throws.

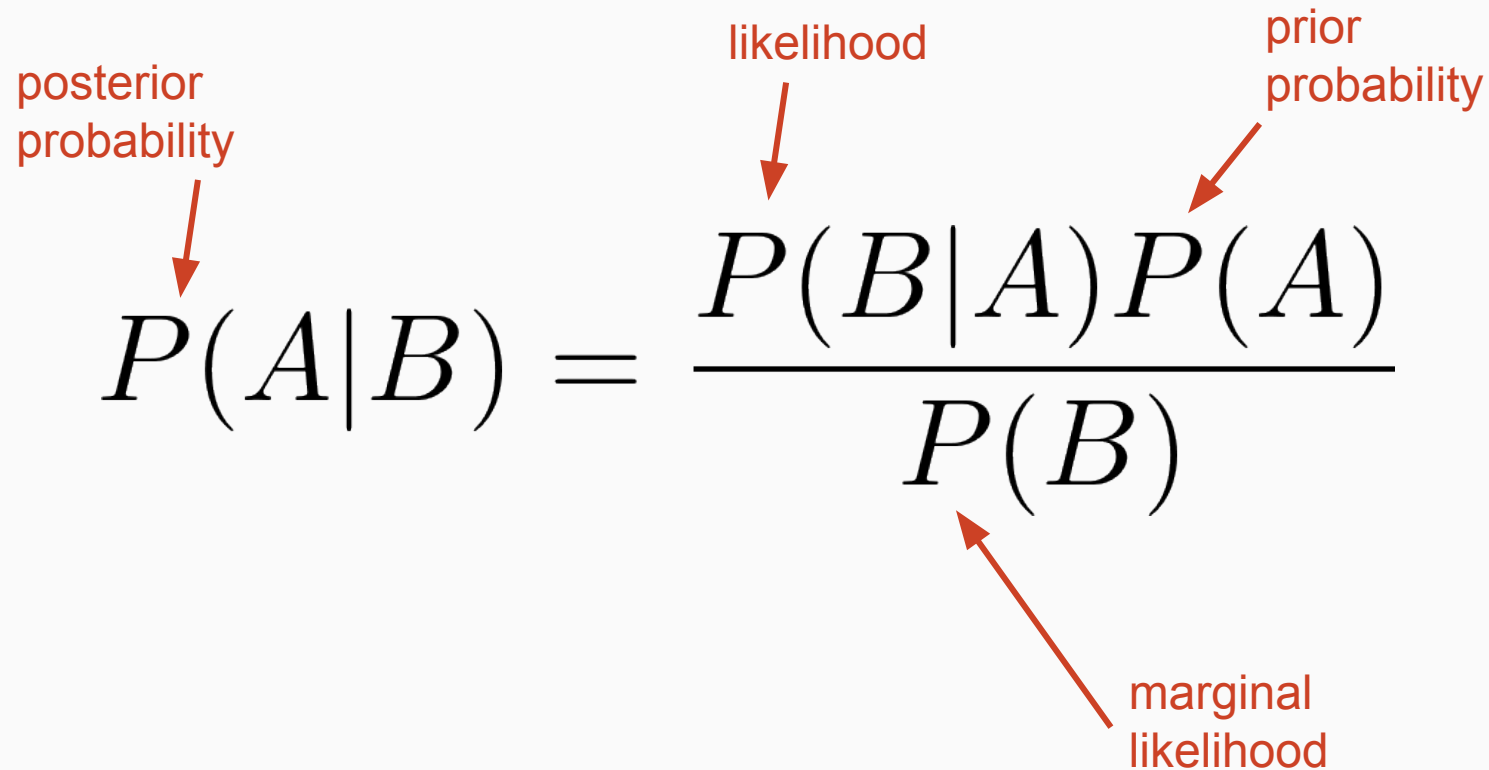




Frequentist: “They’re both so skilled! I have **as much confidence** in musician’s ability to distinguish Haydn and Mozart as I do the drunk’s to predict coin tosses”

Bayesian: “I’m not convinced by the drunken man...”

The Bayesian approach is to incorporate prior knowledge into the experimental results.



The image displays the formula for Bayes' Rule with three red arrows pointing to specific parts of the equation. The arrow from 'posterior probability' points to $P(A|B)$. The arrow from 'likelihood' points to $P(B|A)$. The arrow from 'prior probability' points to $P(A)$. The arrow from 'marginal likelihood' points to $P(B)$.

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

posterior probability

likelihood

prior probability

marginal likelihood

$$P(\text{psychic}|\text{correct}) = \frac{P(\text{correct}|\text{psychic})P(\text{psychic})}{P(\text{correct})}$$

$$= \frac{1.0 * 0.0001}{0.5^{10}}$$

$$= 10.2\%$$

Very subjective!



DID THE SUN JUST EXPLODE?
(IT'S NIGHT, SO WE'RE NOT SURE.)

THIS NEUTRINO DETECTOR MEASURES
WHETHER THE SUN HAS GONE NOVA.

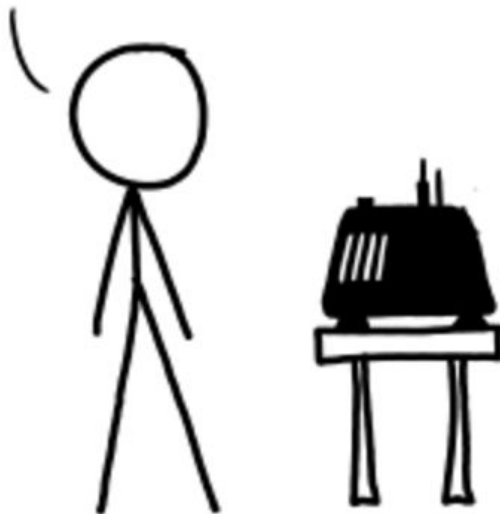
THEN, IT ROLLS TWO DICE. IF THEY
BOTH COME UP SIX, IT LIES TO US.
OTHERWISE, IT TELLS THE TRUTH.

LET'S TRY.
DETECTOR! HAS THE
SUN GONE NOVA?



FREQUENTIST STATISTICIAN:

THE PROBABILITY OF THIS RESULT
HAPPENING BY CHANCE IS $\frac{1}{36} = 0.027$.
SINCE $p < 0.05$, I CONCLUDE
THAT THE SUN HAS EXPLODED.



BAYESIAN STATISTICIAN:

BET YOU \$50
IT HASN'T.



Bayes' Rule:

$$P(A|B) = \frac{P(B|A)P(A)}{P(B|A)P(A) + P(B|A^C)P(A^C)}$$

Posterior Distribution:

$$\pi(\theta|x) = \frac{f(x|\theta)\pi(\theta)}{\int f(x|\theta')\pi(\theta')d\theta'}$$

- **Prior distribution**: describes our current (prior to seeing new data) knowledge about A
- **Likelihood**: the probability of the data given A
- **Posterior distribution**: updated knowledge about A after seeing data

Example 1: bag of coins, HH, HT, TT

Example 2: coin with unknown heads probability

Bayesian Updates

Bayesian updating of posterior probabilities

