

# A Brief Introduction to Bayesian Inference

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# Two Schools of Thought

- Classical (Frequentist)
  - Focused on computing some quantity (mean, std dev, confidence interval, etc.) from a set of observations.
- Bayesian
  - Focused on estimating the probability of some parameter (observed or from a model) given some prior knowledge of the domain.

# Some Terminology

- Probability Density (distribution) Function
  - Usually just called the pdf. It gives the probability of obtaining a value  $X$ . Written  $P(X)$ .
- Conditional Probability
  - The probability of obtaining some value  $X$  given that you have obtained some value  $Y$ . Written as  $P(X|Y)$ .
- Joint Probability
  - The probability of  $A$  and  $B$  happening together. It has the definition  $P(AB) = P(A|B)P(B)$

# An Identity

- Since A and B are arbitrary, then it follows that  $P(AB) = P(BA)$
- Applying our definition of joint probability, we find

$$P(AB) = P(A|B)P(B) = P(B|A)P(A) = P(BA)$$

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

# Bayes' Theorem

Comparing my observed data,  $D$ , to some model,  $M$ .

Having accounted for my data,  
this is my updated belief in my  
model; the *posterior*.

*likelihood*

My initial belief in my  
model; the *prior*.

$$P(M|D) = \frac{P(D|M)}{P(D)} P(M)$$

The support that the data  
provides for the model.

# Using Bayes' Theorem

- Developing Bayes' Theorem was nearly trivial, but employing it in real calculations can be a very daunting task.
- There are several steps which must be done to utilize Bayes' Theorem correctly.

# Using Bayes' Theorem (cont.)

- Formulate the data likelihood function  $P(D|M)$
- Choose a prior
- Determine the posterior  $P(M|D)$
- Search the model space to find the maximum a posteriori estimate.
- Quantify uncertainty using credible regions (these are much like confidence intervals).