

# Module 2 : Probability

Frequentist and Bayesian building blocks

## Agenda:

- Bayesian Probability
  - Structure of Bayes' Theorem: 
$$P[A|B] = \frac{P[A \cap B]}{P[B]} = \frac{P[B|A]P[A]}{P[B]}$$
  - The Monty Hall Problem: illustrating the philosophical difference with Frequentist camp - ability to update probability with new information
  - Examples:
    - Pedigree Analysis

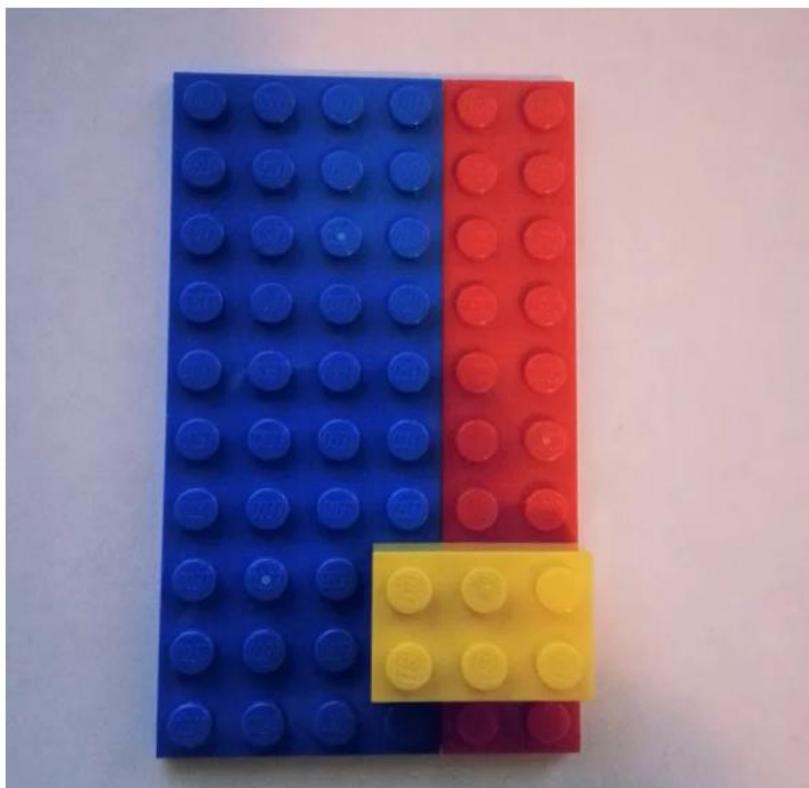
<https://www.countbayesie.com/blog/2015/2/18/bayes-theorem-with-lego>

### Introducing Bayes' Theorem

Bayes' Theorem states:

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

As far as formulae go this one isn't too scary, it doesn't even have a  $\Sigma$ ! But what is actually happening here? Let's pull out some Lego bricks and put some concrete questions to our equation.



Lego Brick Probability Space

$$P[A | B] = \frac{P[A]P[B | A]}{P[B]}$$

You can interpret Bayes as reducing the state space, like the following equations which use the proportion of A intersecting with B over the WHOLE universe (i.e., black die and red die both equal 1 is 1/36) and then reduce the proportion by dividing by the probability of the first event

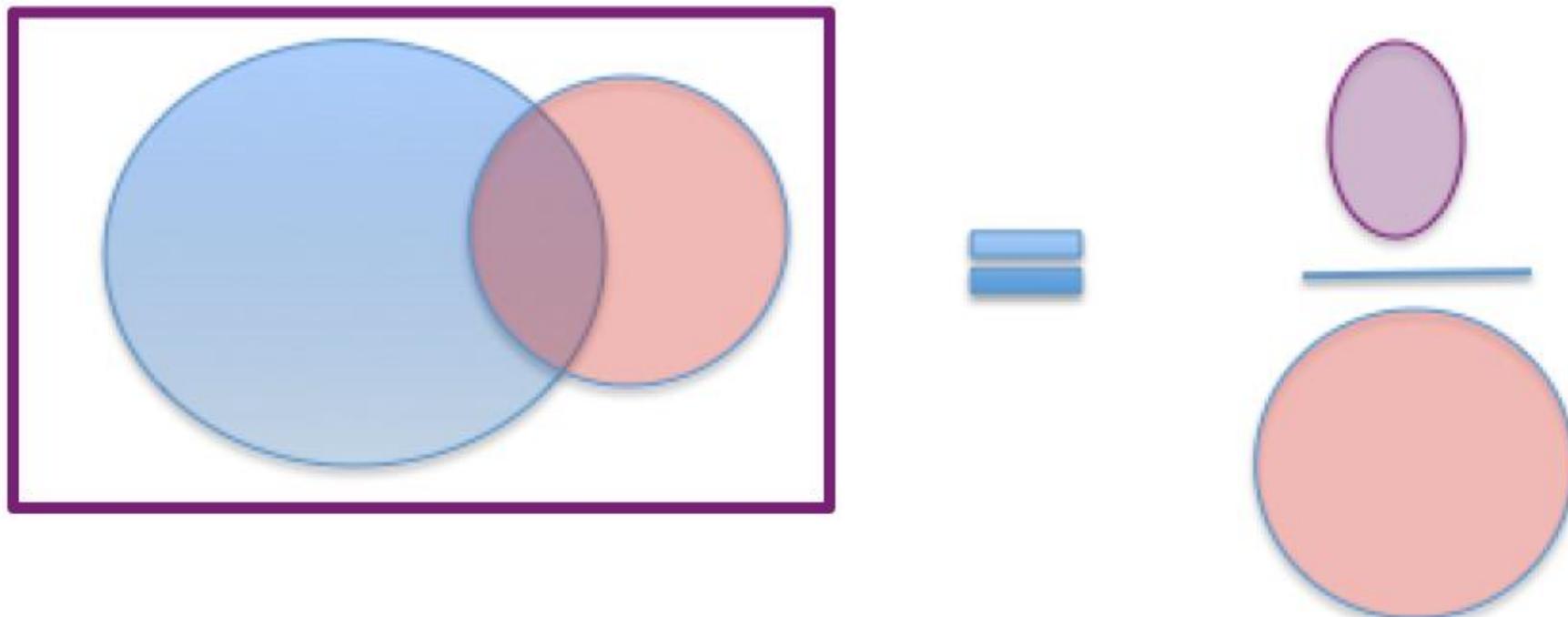
$$P[A | B] = \frac{P[A \cap B]}{P[B]}$$

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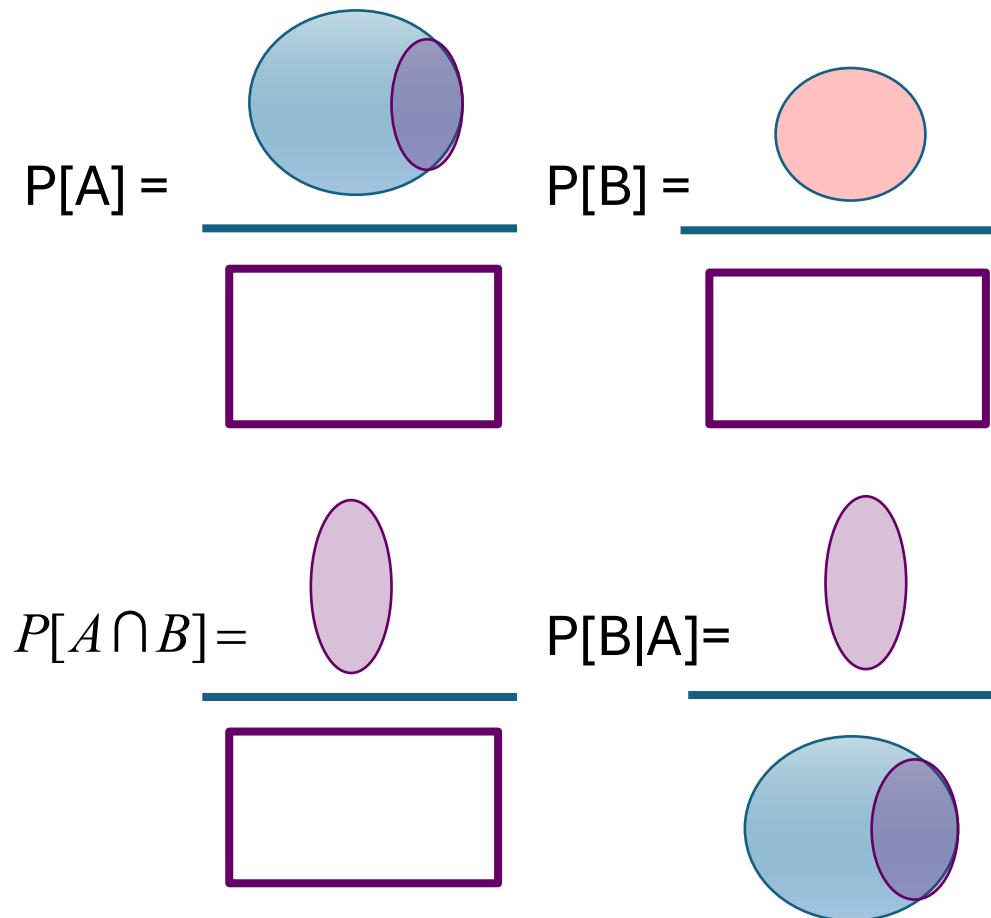
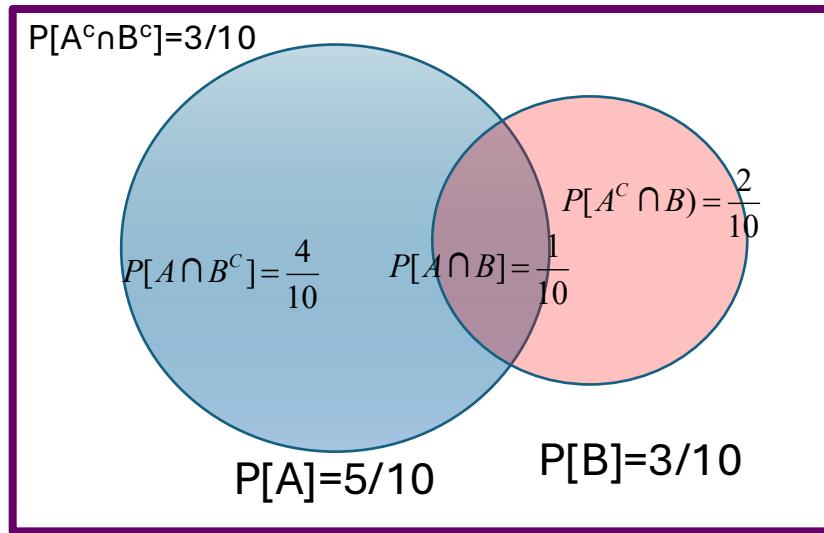
You might notice that both equations involve the SAME **numerator** whereas the **denominator** changes based on what event has happened first i.e., what we already know and what we still want to know.

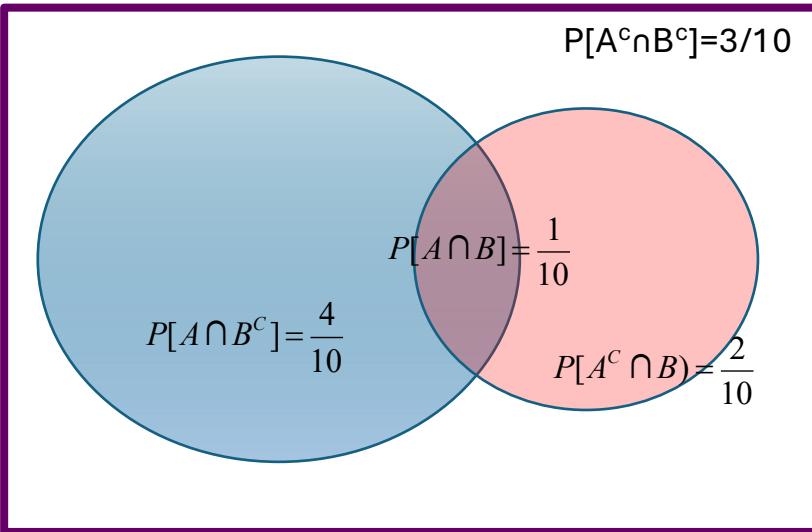
**This is a sophisticated rearrangement of the multiplication rule.**

$$P[A|B] = \frac{P[A \cap B]}{P[B]} = \frac{P[B|A]P[A]}{P[B]}$$



$$P[A | B] = \frac{P[A]P[B | A]}{P[B]}$$





$$P[A | B] = \frac{P[A]P[B | A]}{P[B]}$$

$$P[A] = \frac{5}{10}$$

$$P[A|B] = \frac{1/10}{3/10} = \frac{1}{3}$$

$$P[B] = \frac{3}{10}$$

$$P[A \cap B] = \frac{1}{10}$$

$$P[B|A] = \frac{1/10}{5/10} = \frac{1}{5}$$

**Bayes' theorem in words:** The conditional probability of **A (the hypothesis) given B (the data)** is the conditional probability of **B (data) given A (hypothesis)** scaled by the relative probability of **A compared to B**

$$P[\text{Hypothesis}|\text{DATA}] = \frac{P[\text{Data}|\text{Hypothesis}]P[\text{Hypothesis}]}{P[\text{Data}]}$$

The **PRIOR** hypothesis:  
The original probability of the hypothesis without any additional information

The **LIKELIHOOD** interpreted as:  
P(observation GIVEN the hypothesis)

$$P[A | B] = \frac{P[A \cap B]}{P[B]} = \frac{P[A]P[B | A]}{P[B]}$$

the **POSTERIOR probability** interpreted as  
the P(hypothesis GIVEN the observation)

The **observation/data/  
Evidence** that has been observed