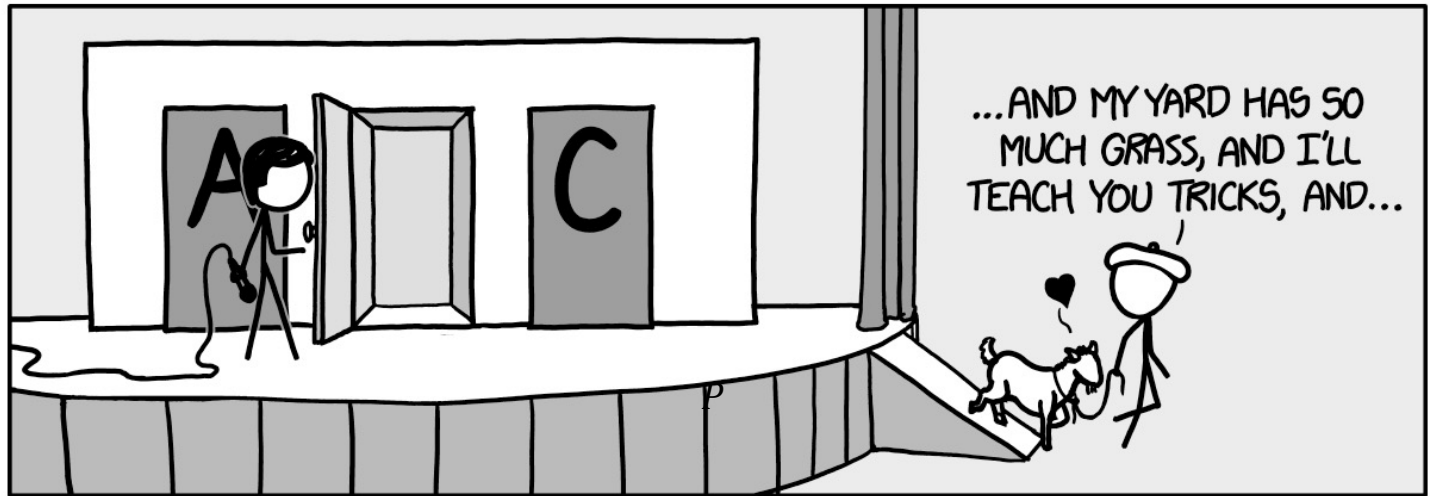


# Stat 88: Probability and Statistics in Data Science



<https://xkcd.com/1282/>

Lecture 5: 2/1/2022

Symmetry in Sampling, Bayes' Rule, Random variables

Sections 2.2, 2.3, 2.4, 3.2

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# Agenda

- § 2.2: Symmetries in sampling & counting
- § 2.3: Bayes' rule
- § 2.4: Use and interpretation of Bayes' rule
- § 3.2: Random variables: intro

## Review: Product rule and counting

- Recall the product rule for counting: if there are sequences constructed in  $n$  stages, with  $k_i$  options at each stage, then the total number of sequences is  $k_1 \times k_2 \times \cdots \times k_n$
- Count the number of outcomes for each stage and multiply them. (Recall the tree diagrams, and how we count outcomes.)

- Deal 5 cards from a deck. Number of possible sequences?

$$\underline{52 \cdot 51 \cdot 50 \cdot 49 \cdot 48} = \frac{52!}{47!}$$

- Number of outcomes from rolling three 6-sided dice?

$$6 \cdot 6 \cdot 6 = 6^3$$

- 10 students, choose 2 for committee (to be the president and secretary respectively). Number of possible committees?

$$\underline{10 \cdot 9}$$

What if only who is chosen matters, not <sup>3</sup> which position they are chosen

$$\text{for } \frac{(10 \cdot 9)}{2} = 45 = \frac{10!}{8! \cdot 2!}$$

## Example

- The English language has 26 letters. 5 letters are chosen **with** replacement. What is the chance that the middle three letters are all *different*, and the *first* and *last* are the same as each other, and also the same as one of the three middle letters.

$$\frac{3}{26} \cdot \frac{26}{26} \cdot \frac{25}{26} \cdot \frac{24}{26} \cdot \frac{1}{26} = \frac{1}{26^5}$$

## Probabilities of dealing cards:

- Deal 2 cards from top of the deck.
  - How many possible sequences of 2 cards?  $52 \cdot 51$
  - What is the chance that the second card is red?  $\frac{26}{52}$  (no info about 1st card)

- $P(5^{\text{th}} \text{ card from top is red}) = \frac{26}{52} = \frac{1}{2}$

- 21<sup>st</sup> card and 35<sup>th</sup> cards are red =  $P(R_{21} \cap R_{35})$  = (write it using conditional prob)

$$P(R_{21}) P(R_{35} | R_{21}) = \frac{26}{52} \cdot \frac{25}{51}$$

- $P(7^{\text{th}} \text{ card is a queen}) = \frac{4}{52}$

- $P(B_{52}^A | R_{21} R_{35}) = \frac{26}{50}$

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

# Counting permutations & combinations

- Recall # of ways to rearrange  $n$  things, taking them 1 at a time is  $n!$
- If we have only  $k \leq n$  spots to fill, then  $n \cdot (n-1) \cdot \dots \cdot (n-(k-1))$
- # of perm. of  $n$  things taken  $k$  at a time. *when order matters*

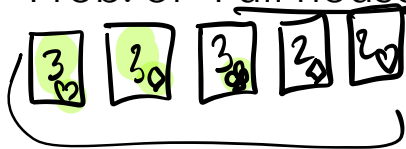
$$\underbrace{n \cdot (n-1) \cdot (n-2) \cdot (n-3)}_{4 \text{ spots}}$$

- If we don't care about order, then we are counting subsets, and this number is denoted by  $\binom{n}{k}$ , which we get by dividing:  $n \cdot (n-1) \cdot \dots \cdot (n-(k-1))$  by  $k!$

$$\frac{n!}{(n-k)!} \cdot \frac{1}{k!}$$

- Note:  $\binom{n}{n} = 1$ ,  $\binom{n}{0} = 1$

- Prob. of "Full house" =  $P(\text{triple \& pair})$



1/31/22

$$\frac{\binom{13}{1} \binom{4}{3} \binom{12}{1} \binom{4}{2}}{\binom{52}{5}}$$

26 choices  
pick 3  $\frac{26 \cdot 25 \cdot 24}{6}$

$abc = acb =$   
 $bac = bca =$   
 $cab = cba$

it matters which  
the triple & which

## Section 2.3: Bayes' Rule:

- I have two containers: a jar and a box. Each container has five balls: The jar has three red balls and two green balls, and the box has one red and four green balls.
- Say I pick one of the containers at random, and then pick a ball at random. What is the chance that I picked the box, if I ended with a red ball?

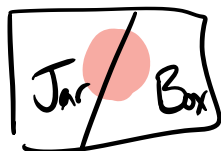


$$P(\text{jar}) = P(\text{box}) = \frac{1}{2} \text{ (priors)}$$

$$P(\text{jar} | \text{red ball}) = \text{posterior prob.}$$

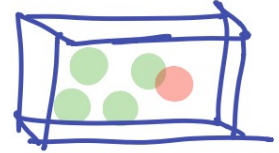
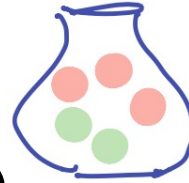
$$P(\text{red ball} | \text{jar}) = \frac{3}{5}, \quad P(\text{red ball} | \text{box}) = \frac{1}{5} \quad \text{likelihooods}$$

$$P(\text{box} | \text{red ball}) = \frac{P(B \cap R)}{P(R)} = \frac{P(R | B) P(B)}{P(R)} = \frac{(\frac{1}{5})(\frac{1}{2})}{P(R)}$$



$$P(R) = P(R \cap B) + P(R \cap J) \\ = P(R | B) P(B) + P(R | J) P(J)$$

## Jars and boxes



$$P(R) = \left(\frac{1}{5}\right)\left(\frac{1}{2}\right) + \left(\frac{3}{5}\right)\left(\frac{1}{2}\right)$$
$$= \frac{1 + 3}{10} = \frac{4}{10} = \frac{2}{5}$$

$$P(B | R) = \frac{\left(\frac{1}{\cancel{5}}\right)\left(\frac{1}{2}\right)}{\left(\frac{2}{\cancel{5}}\right)} = \frac{1}{4}$$



# Prior and Posterior probabilities

- The **prior** probability of drawing the box = 0.5 (before we knew anything about the balls drawn)
- The **posterior** probability of drawing the box = 0.25 (this is after we *updated* our probability, *given* the information about which ball was drawn)

## Computing Posterior Probabilities: Bayes' Rule

- We want the *posterior* probability. That is, the conditional prob for the first stage  $A$ , *given* the second stage  $B$ .

- Division rule (for conditional probability) =  
$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

- Using the multiplication rule on  $P(A \cap B)$ , we get:

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

- Rule first written down by Rev. Thomas Bayes in the 18<sup>th</sup> century. Helps us compute posterior probability, given prior prob. And **likelihoods** (which are conditional probabilities for the second stage given the first, which are generally easier to compute.)

## Exercise 2.6.9

A factory has two widget-producing machines. Machine I produces 80% of the factory's widgets and Machine II produces the rest. Of the widgets produced by Machine I, 95% are of acceptable quality. Machine II is less reliable - only 85% of its widgets are acceptable.

Suppose you pick a widget at random from those produced at the factory.

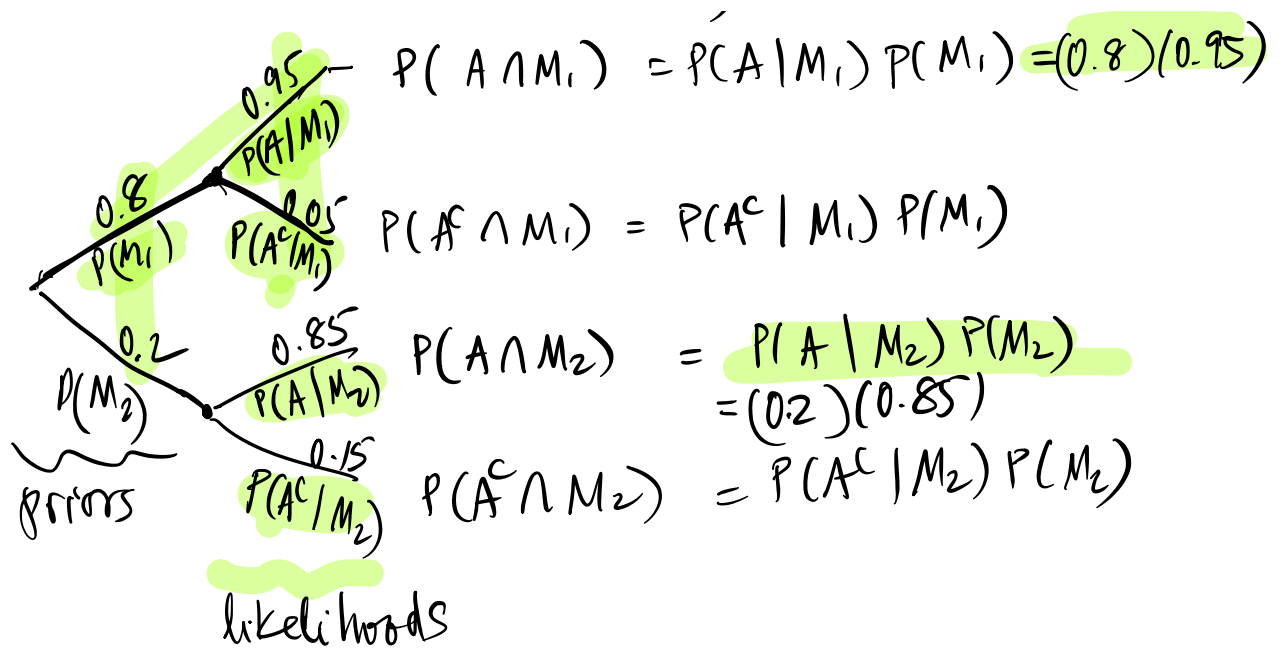
a) Find the chance that the widget is acceptable, given that it is produced by Machine I. (likelihood)  $P(A | M_1)$

b) Find the chance that the widget is produced by Machine I, given that it is acceptable. (posterior)

$M_1, M_2, A$ : widget is acceptable  
Machine I, Machine II,

$$P(A | M_1) = 0.95, \quad P(A | M_2) = 0.85$$

$$P(M_1 | A) = \frac{P(A \cap M_1)}{P(A)}$$



$$\underset{\text{posterior}}{P(M_1 | A)} = \frac{P(A \cap M_1)}{P(A)} = \frac{(0.8)(0.95)}{(0.8)(0.95) + (0.2)(0.85)}$$

## Example: Binge drinking & Alcohol related accidents

(This example is from the text *Intro Stats* by De Veaux, Velleman, and Bock)

For men, binge drinking is defined as having 5 or more drinks in a row and for women as having 4 or more drinks in a row.

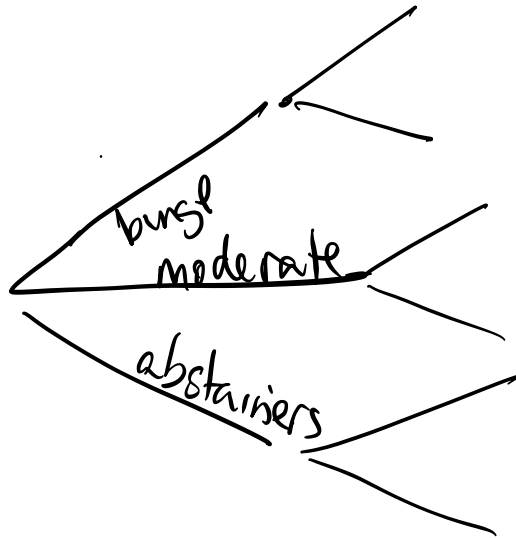
(The difference is because of the average difference in weight.)

According to a study by the Harvard School of Public Health (H. Wechsler, G. W. Dowdall, A. Davenport, and W. Dejong, "*Binge Drinking on Campus: Results of a National Study*"):

- 44% of college students engage in binge drinking, 37% drink moderately, and 19% abstain entirely. (*priors*)
- Another study, published in American journal of Health Behavior, finds that among binge drinkers aged 21 to 34, 17% have been involved in an alcohol-related automobile accident, while among nonbingers of the same age, only 9% have been involved in such accidents. (*likelihoods*)
- Given that a student has been in a car crash, what is the chance that they were a binge drinker? (*posterior*)

## Example: Binge drinking & Alcohol related accidents

- Make a tree diagram. What are we given? What do we want to compute?



## 2.4: Use and interpretation of Bayes' rule

- Harvard study: 60 physicians, students, and house officers at the Harvard Medical school were asked the following question:
- "If a test to detect a disease whose **prevalence** is 1/1,000, has a false positive rate of 5 per cent, what is the chance that a person found to have a positive result actually has the disease, assuming that you know nothing about the person's symptoms or signs?"
- *Prevalence* aka *Base Rate* = fraction of population that has disease.
- *False positive rate*: fraction of positive results among people who don't have the disease
- *Positive result*: test is positive
- What is your guess – without any computations?