

# DataSci 520

## lesson 3

combinatorics and probability  
distributions



PROFESSIONAL &  
CONTINUING EDUCATION  

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UNIVERSITY *of* WASHINGTON

## today's agenda

- discrete vs continuous random variables
- understand the 3 axioms of probability
- properties of basic probability distributions
- conditional, mutually exclusive, and independent events

# Introduction to Counting & Probability

## > Why counting?

- Counting is fundamental to probability theory.
- Probability is the extent or likelihood of an event or set of events.
  - > Depends heavily on the ability to COUNT up potential outcomes.

# Counting

- > Counting is one of the biggest areas of mathematics, called “Combinatorics”.
- > Example:
  - Subway has 4 different breads, 5 different meats, and 4 different toppings. How many sandwich combinations are there?
  - How many different 4-beer tasters can I have in a bar with 10 beers on tap?
- > Solve the above with the “Multiplication Principle”.

- Subway:

$$\frac{4}{\text{\# of Breads}} * \frac{5}{\text{\# of Meats}} * \frac{4}{\text{\# of Toppings}} = 80 \text{ Sandwiches}$$

- Beer Tasters:

$$\frac{10}{\text{\# Beer Choices}} * \frac{9}{\text{\# Beer Choices}} * \frac{8}{\text{\# Beer Choices}} * \frac{7}{\text{\# Beer Choices}} = 5,040 \text{ Beer Tasters}$$

# Multiplication Principle

- > If there are A ways of doing task-a, B ways of doing task-b, then there are  $A*B$  ways of completing both tasks.
- > Example:
  - If I have 5 books, how many ways can I ORDER them on the bookshelf

$$\begin{array}{ccccccccc} \underline{5} & & \underline{4} & & \underline{3} & & \underline{2} & & \underline{1} & & \\ \# \text{ Book Choices} & * & \# \text{ Book Choices} & * & \# \text{ Book Choices} & * & \# \text{ Book Choices} & * & \# \text{ Book Choices} & & =120 \text{ Book Orderings} \end{array}$$

# Factorials

## > Factorials

- Count # of ways to order  $N$  things =  $N!$

## > Factorials get VERY VERY large quickly.

- $21!$  Is larger than the biggest long-int in 64 bit.
  - >  $21! = 5.1\text{E}19$
  - > Biggest long int (64bit) =  $9.2\text{E}18$
- Fun fact, every 52 card shuffle is highly likely to be the only time that shuffle has ever occurred.

# Counting Subgroups

- > Revisit: 10 beers on tap, need a sample of 4 different beers.
- > Let's assume order matters.
  - E.g. Amber-Stout-Porter-Red is different than Red-Porter-Stout-Amber
- > Use "Permutations" or "Pick":

$$10 * 9 * 8 * 7 = \frac{10!}{6!} = \frac{10!}{(10 - 4)!} = 10P4 = P(10,4)$$

# Counting Subgroups

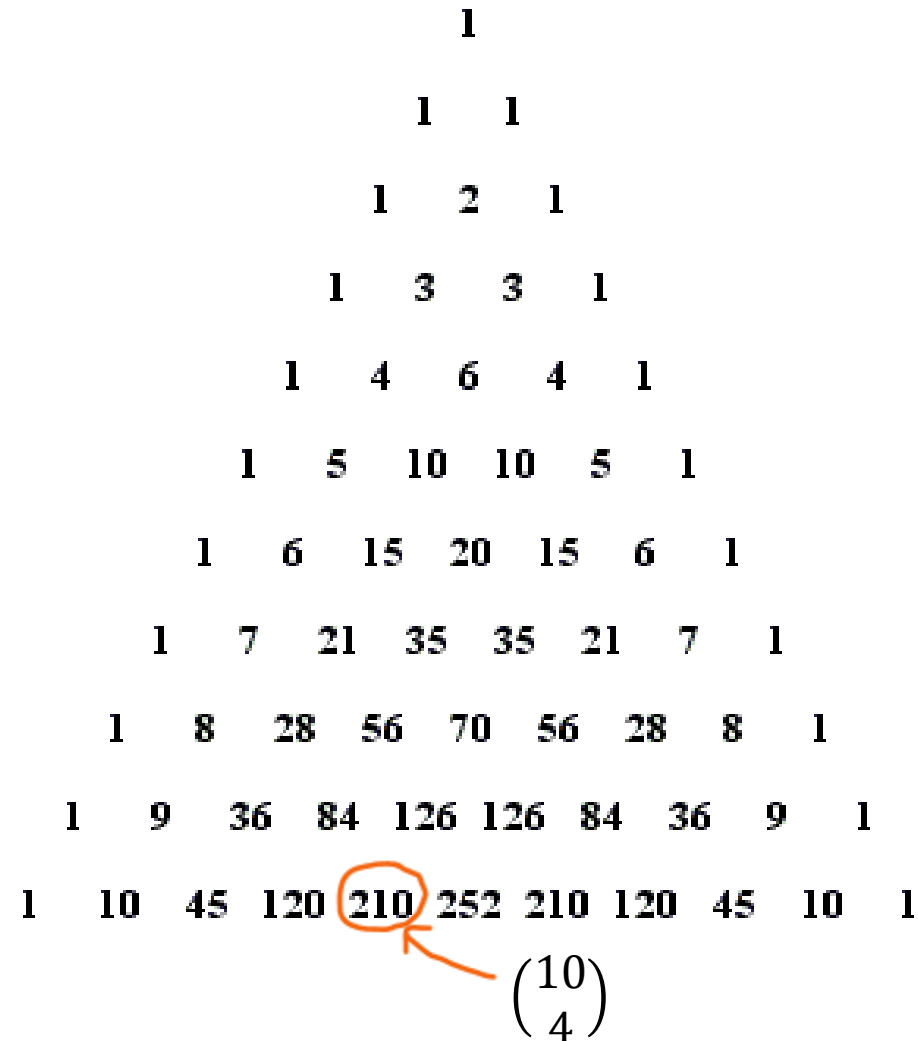
- > Now we assume order doesn't matter.
- > Use 'Combinations' or 'choose'.
- > Remember that  $10*8*7*6 =$  orderings of 4 beers.
- > Now we are counting some groups multiple times, so we divide out the multiple counts.

$$\frac{(10 * 9 * 8 * 7)}{(\# \text{ of ways to order 4 beers})} = \frac{P(10,4)}{4!} = \frac{10!}{(10 - 4)! 4!} = C(10,4) = \binom{10}{4}$$



# More About Combinations

- > Combinations appear on Pascal's Triangle!
- >  $C(N,x)$  appears on the Nth row, x-th number over (starting counting at 0).



# Counting Examples

- > There are 10 Light beers on tap, and 10 Dark beers on tap.
- > How many ways can Rick get a 4-beer sampler that contains exactly 1 light beer? (Order does not matter)

$$\begin{aligned} &= \frac{(\# \text{ of ways for 1 Light}) * (\# \text{ of ways for 3 D})}{(\# \text{ of ways to arrange 1L \& 3D})} = \frac{C(10,1) * C(10,3)}{C(4,1)} = \frac{10 * 120}{4} \\ &= 300 \end{aligned}$$












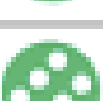
# Counting Examples

- > 6:5 Blackjack is dealt with a 6 shoe deck (52\*6=312 cards)
- > How many ways can someone get dealt two rank 10 cards?

$$\binom{6decks * 4ranks * 4suits}{2} = \binom{96}{2} = 4,560$$

# DICE!

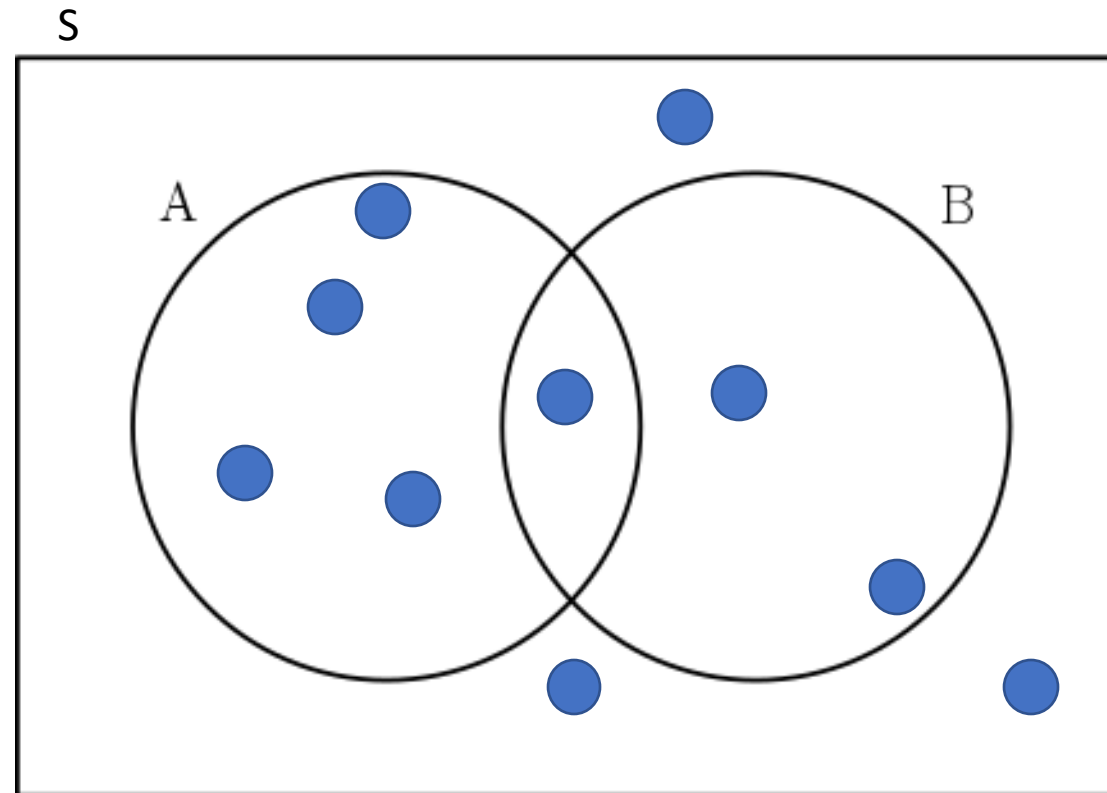
> How many ways can two dice be rolled to get a sum of 10?

						
	2	3	4	5	6	7
	3	4	5	6	7	8
	4	5	6	7	8	9
	5	6	7	8	9	10
	6	7	8	9	10	11
	7	8	9	10	11	12

# Probability

- > The Probability of an event, A, is the number of ways A can occur, divided by the number of total possible outcomes in a "Sample Space".
  - The sample space is the set of all possible outcomes.

$$P(A) = \frac{N(A)}{N(S)}$$



In this example, if '●' is an event, then:

- $P(A) = 6/10 = 3/5$
- $P(B) = 4/10$
- $P(S) = 10/10 = 1.0$

# Probability

> If '●' is an event, then

– Intersection:

$$P(A \cap B) = \frac{2}{10} = \frac{1}{5}$$

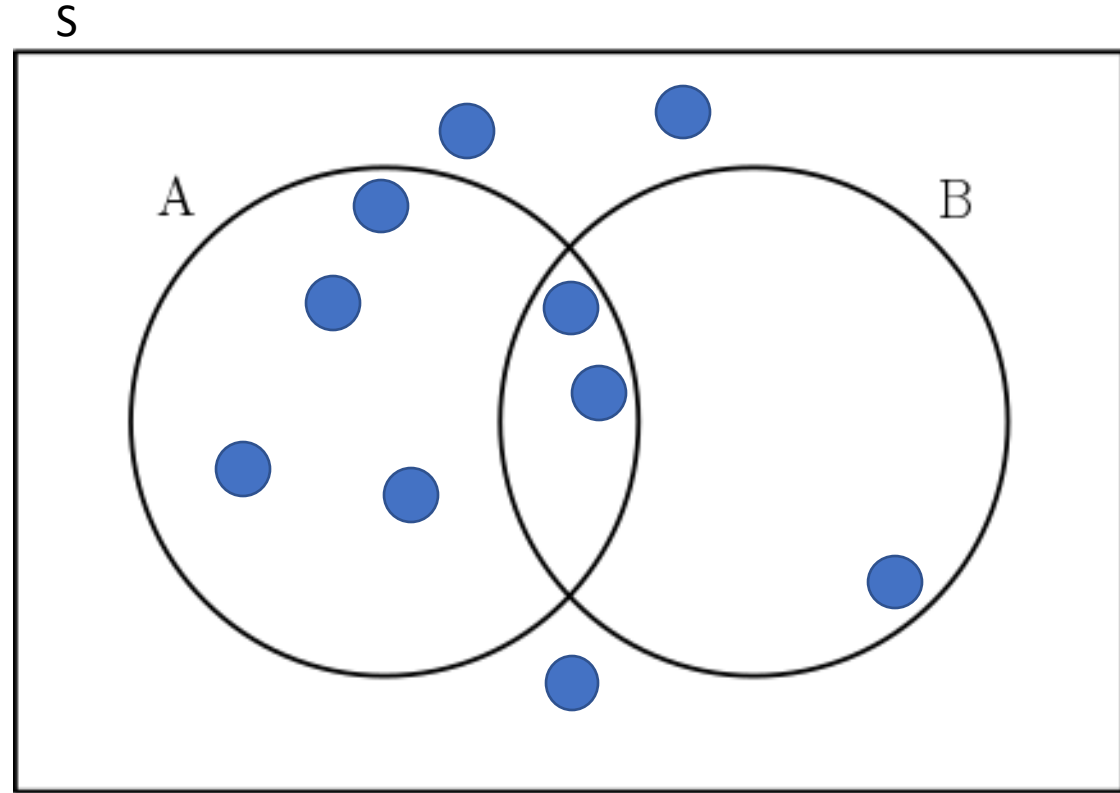
– Union:

$$P(A \cup B) = \frac{7}{10}$$

– Negation:

$$P((A \cup B)') = \frac{3}{10}$$

$$P(A' \cap B') = \frac{3}{10}$$



# Axioms of Probability

1. Probability is bounded between 0 and 1.

$$0 \leq P(A) \leq 1$$

Note: “percent” literally means per one hundred.

2. Probability of the Sample Space is equal to 1.

$$P(S) = 1$$

3. The probability of finite *mutually exclusive* events is the sum of their probabilities.

$$P(A \cup B) = P(A) + P(B) \text{ if } A \text{ and } B \text{ are Mutually Exclusive}$$

## Quick Review – Probabilities

- > What is the probability of rolling a 10?
- > What is the probability of rolling an even #?
- > What is the probability of doubles?

1	2	3	4	5	6	7
2	3	4	5	6	7	8
3	4	5	6	7	8	9
4	5	6	7	8	9	10
5	6	7	8	9	10	11
6	7	8	9	10	11	12

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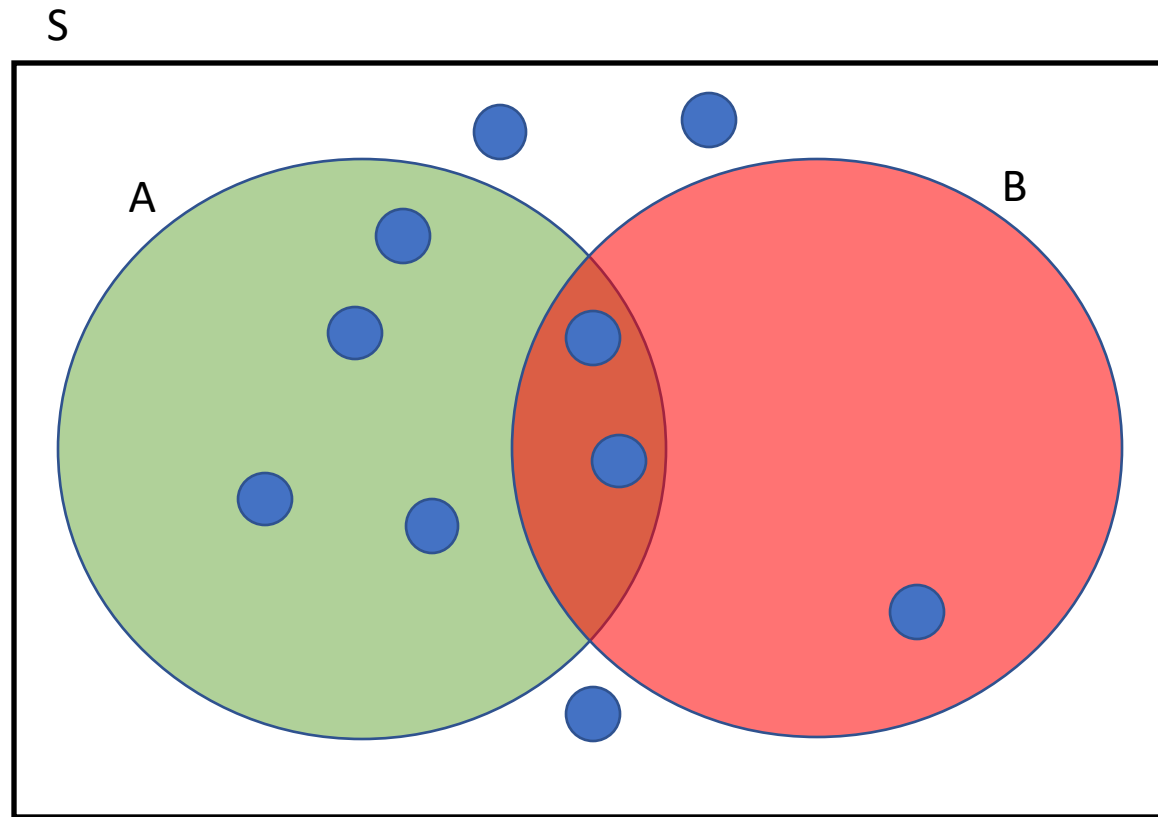
# “At Least” or “At Most” Probabilities

- > When dealing with problems that are “At Least” or “At Most”, we can think of what else can happen and do 1 minus that.
- > Ex:
  - What is the probability that the sum is at least 3?
    - >  $1 - P(\text{sum} = 2) = 1 - (1/36) = 35/36$ .
  - If it is always 40% chance of rain each day, what is the probability that it rains on at most 6 days this week?
    - >  $1 - P(\text{rains all week}) = 1 - (0.4)^7$

	2	3	4	5	6	7
	3	4	5	6	7	8
	4	5	6	7	8	9
	5	6	7	8	9	10
	6	7	8	9	10	11
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Why is this false?

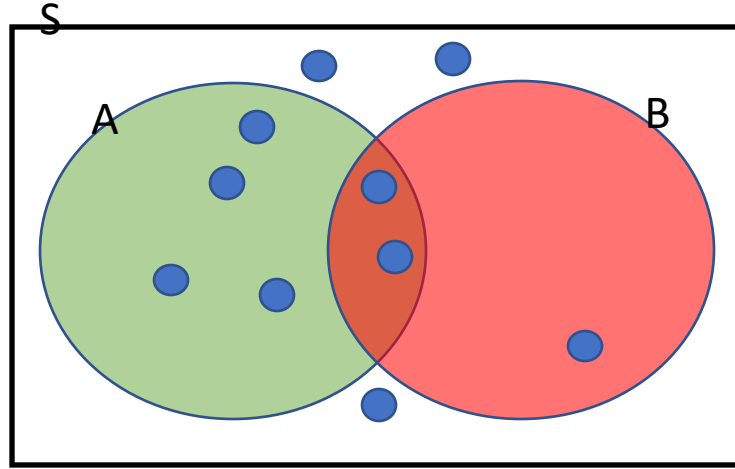
$$P(A \cup B) = P(A) + P(B)$$



# Mutually Exclusive Events

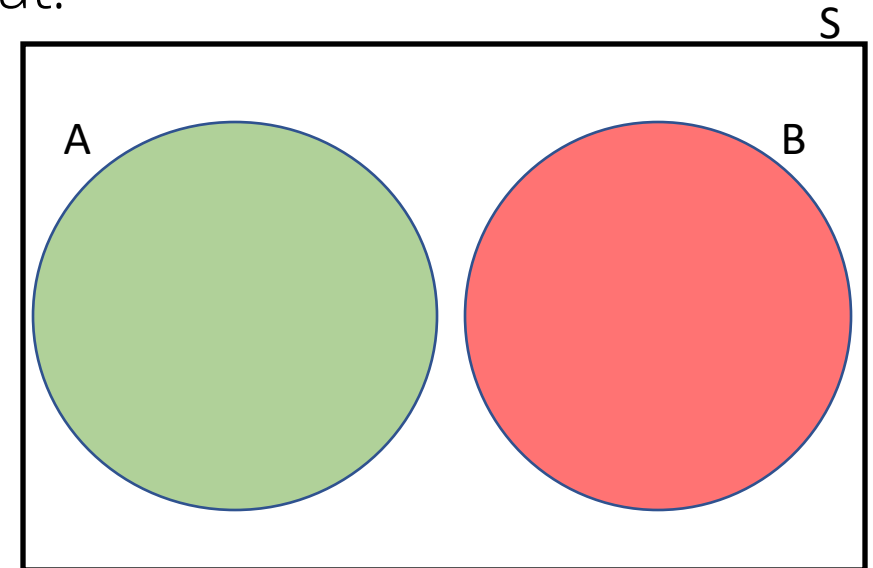
> In all cases, the probability of the union of A and B takes the form:

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$



> If A and B are *mutually exclusive* then means that:

$$P(A \cap B) = 0$$
$$P(A \cup B) = P(A) + P(B)$$



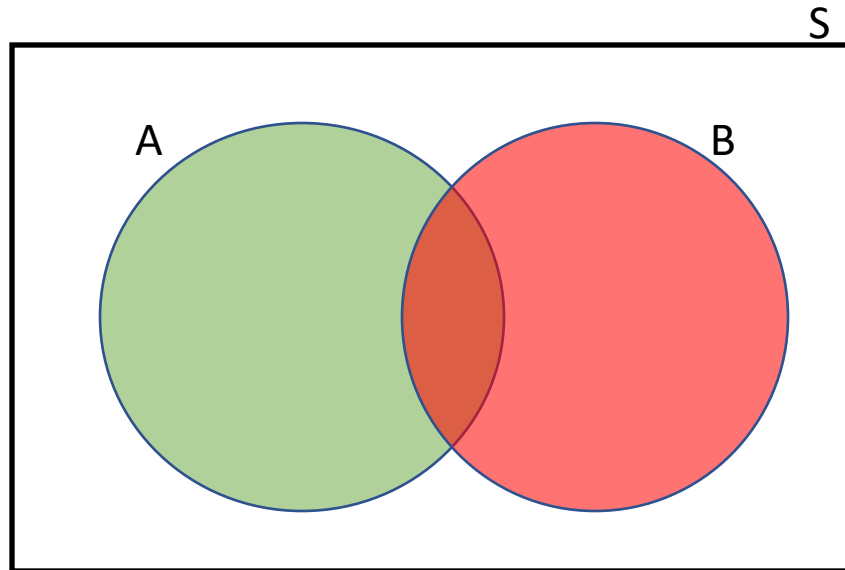
# Conditional Probability

> The Probability of A *given* B is written:

$$P(A|B)$$

> And is equal to:

$$P(A|B) = \frac{P(A \cap B)}{P(B)}, \text{ compare to: } P(E) = \frac{P(E)}{P(S)}$$



# Independence VS Mutually Exclusive

- > These concepts are not similar AT ALL, and in fact, are nearly opposite ideas.
- > If A is M.E. of B, then  $P(A|B) = 0$



The event, B, has a HUGE effect on the occurrence of event A.

- > If A is independent of B, then  $P(A|B) = P(A)$

**Example:** The probability the sidewalk is wet given it is raining is very high, but the probability that it is raining given the sidewalk is wet could be quite low, if I run my sprinklers often.

# ODDS

- > Odds are probabilities expressed as
  - (Count of events in favor):(Count of events not in favor), with the reduced form of probability in fractional form.

$$P(A) = \frac{n}{m} = \frac{n}{n + (m - n)}$$

This implies the odds are **n:(m-n)**

Example:

- > If  $P(A)=50\%$ , then the odds are 1:1 or “one to one”.
- > If  $P(A)=1/3$ , then the odds are 1:2 or “one to two”.
- > If  $P(A)=5/6$ , then the odds are 5:1. ‘five to one’.
- > If the odds are 3:20, then  $P(A)=3/23$
- > A straight up sports bet in Vegas has odds 1:1 (50%) (but pays 0.95Xbet.)

# Independent Events

- > Event  $A$  is *independent* of event  $B$  if and ONLY if:

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

- >  $A$  being independent of  $B$  does NOT imply  $B$  is independent of  $A$

$$P(A|B) = P(A) \not\Rightarrow P(B|A) = P(B)$$

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

$$P(A)P(B) = P(A \cap B)$$

**Used in a sentence:** The event that my boss takes a vacation has an impact on when I take vacation. But when I take vacation has no impact on when my boss takes a vacation.

## discrete vs continuous random variables

	discrete	continuous
sample space	counted	measured
math	summation	integration
$P(A)$	probability mass function $P(A) = \sum_{a \in A} P(A = a)$	probability density function $P(A) = \int_{a \in A} P(A)$
examples	dice roll, number of arrivals, etc.	price, height, distance
visualizations	bar plot, n-way tables, ...	histograms, box plots, ...



Consider the probabilities of all potential sums of the two die:

$$P(2) = 1/36$$

$$P(3) = 2/36$$

$$P(4) = 3/36$$

$$P(5) = 4/36$$

$$P(6) = 5/36$$

$$P(7) = 6/36$$

$$P(8) = 5/36$$

$$P(9) = 4/36$$

$$P(10) = 3/36$$

$$P(1,1) = 2/36$$

$$P(1\ 2)=1/36$$

	1	2	3	4	5	6
1	2	3	4	5	6	7
2	3	4	5	6	7	8
3	4	5	6	7	8	9
4	5	6	7	8	9	10
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6	7	8	9	10	11	12

If we consider all the event possibilities together, this is called a “**Probability Distribution**”.

# Discrete Distributions

- > Properties of Discrete Distributions:
  - Sum of all events MUST be equal to 1.
  - Probability of event,  $E$ , is equal to the value of the distribution at that point.
  - No negative values or values greater than 1.

# Discrete Distribution: Bernoulli Distribution (1 coin flip)

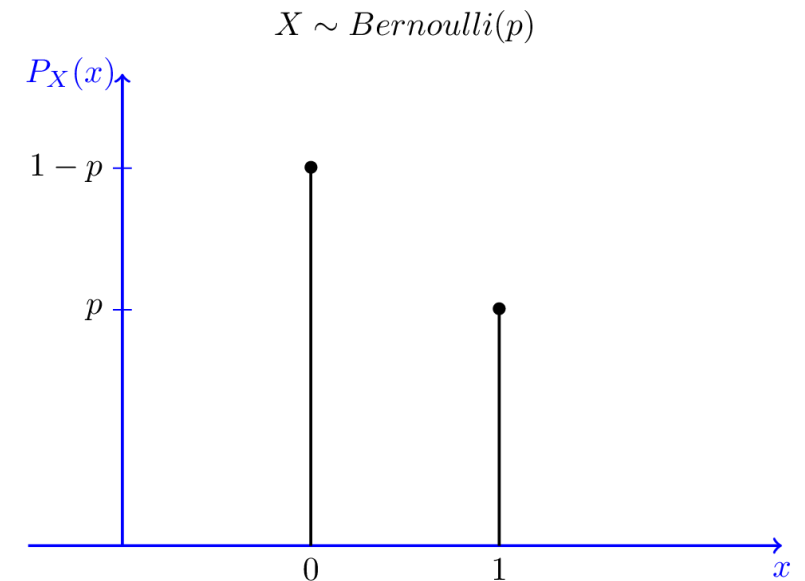
> Bernoulli (1 event, example: 1 coin flip)

$$P(x) = \begin{cases} p & \text{if } x = 1 \\ (1 - p) & \text{if } x = 0 \end{cases}$$

$$P(x) = p^x (1 - p)^{(1-x)} \text{ for } x \in \{0, 1\}$$

> Mean =  $p$

> Variance =  $p(1-p)$



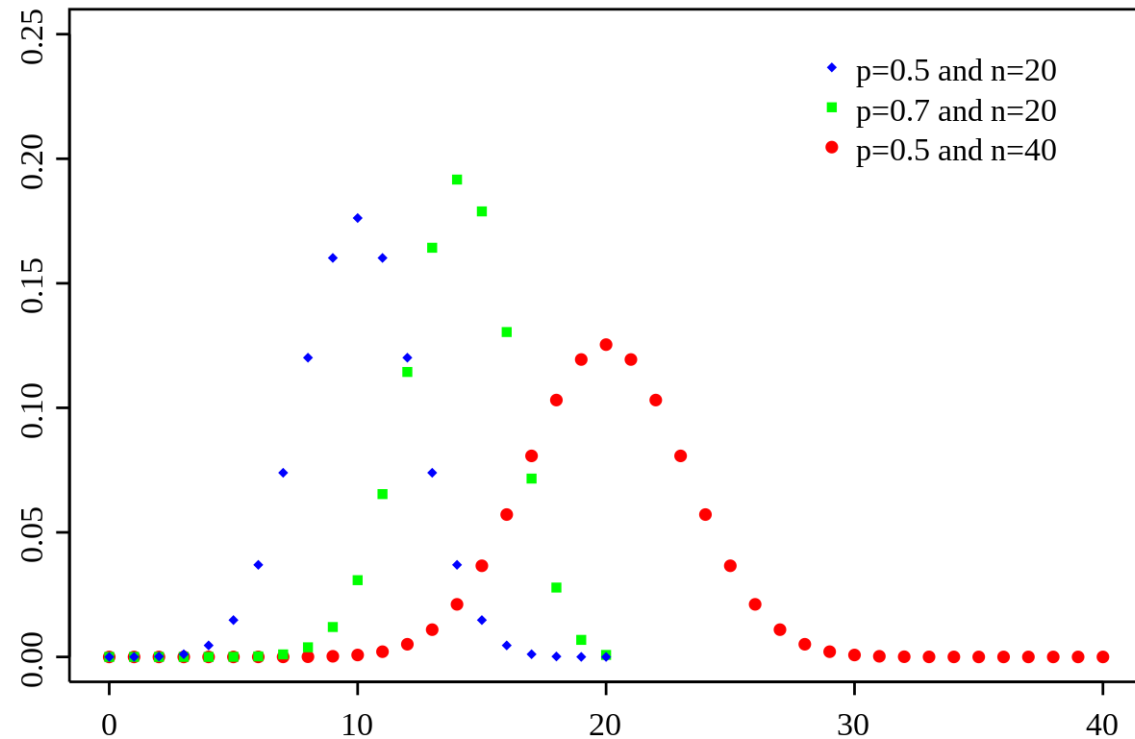
# Discrete Distribution: Binomial Distribution (Multiple Coin Flips)

> Multiple independent events = Product of Bernoulli probabilities.

$$P(x|N, p) = \binom{N}{x} p^x (1 - p)^{(N-x)}$$

> Mean =  $np$

> Variance =  $np(1-p)$



> Note: for larger n, we approximate this by a normal distribution.

# Discrete Distribution: Poisson Distribution (Counting)

> Count number of events in a time span.

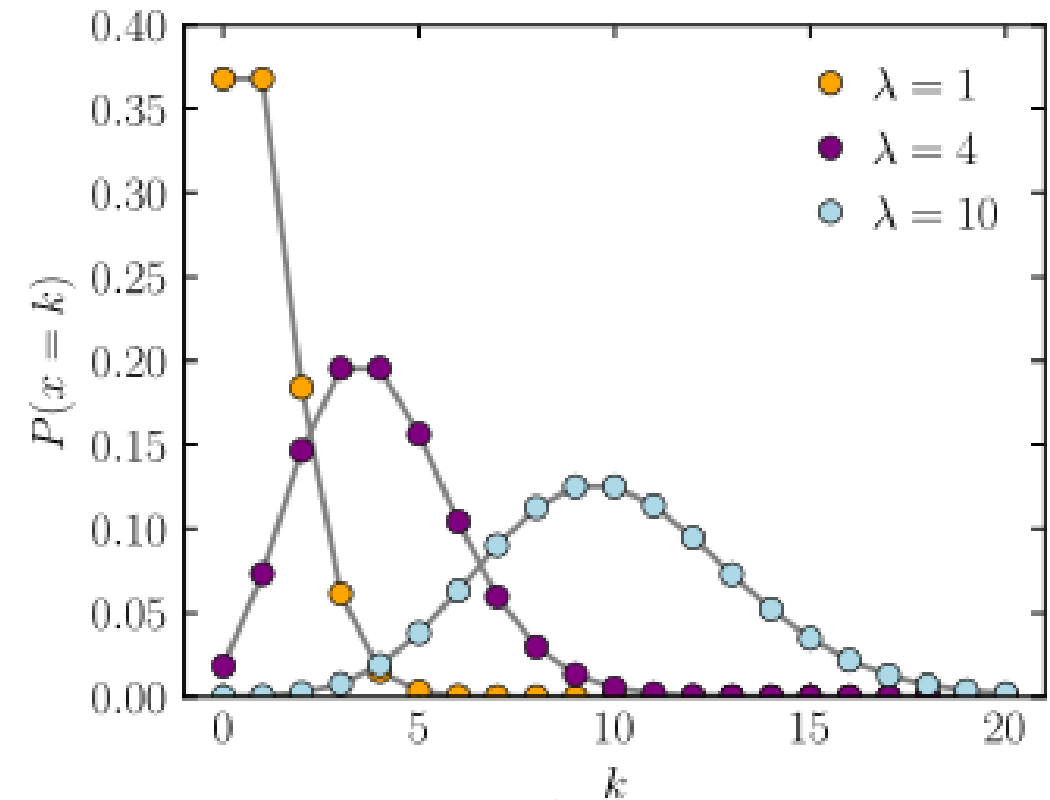
$$P(x|\lambda) = \frac{\lambda^x}{x!} e^{-\lambda}$$

> Mean =  $\lambda$

> Variance =  $\lambda$

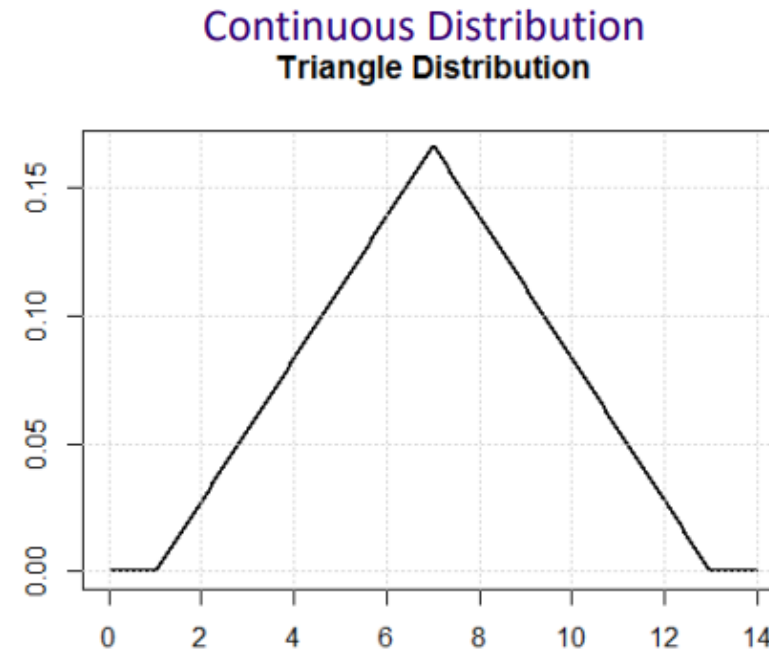
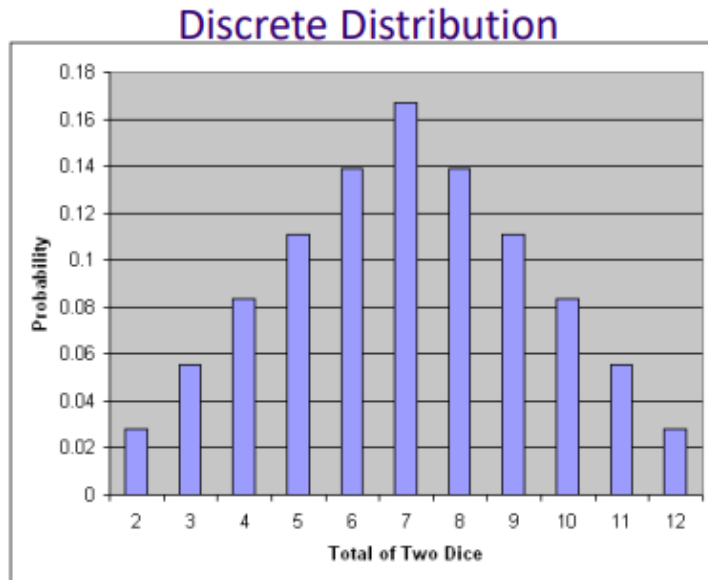
> Interpretation: The rate of occurrence of an event is equal to lambda in a finite period of time

> Can also approximate by a normal at large lambda.



# Continuous Distributions

- > Properties of Continuous Distributions:
  - AREA under the curve MUST be equal to 1.
  - Probability of event,  $E$ , is equal to the AREA under the curve between two points.
  - No negative values or values greater than 1.
  - Probability of a single, exact value is 0.

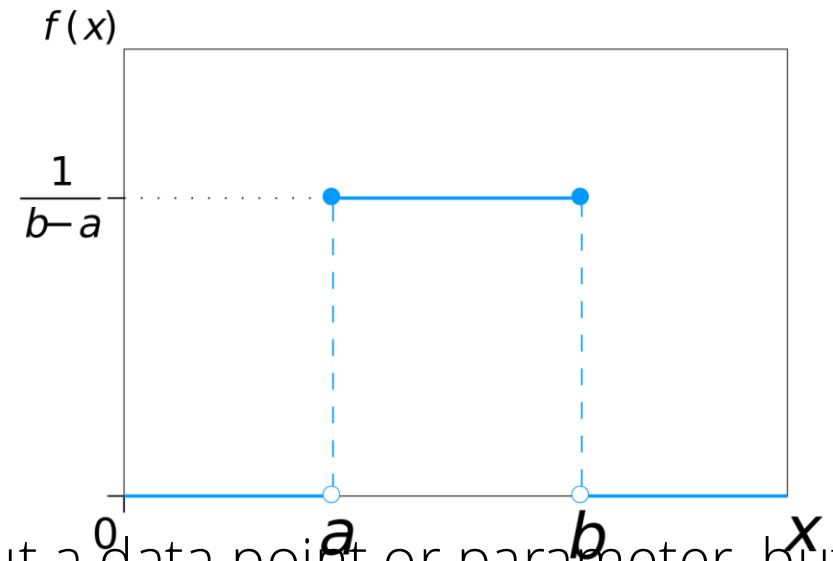


## Continuous Distribution: Uniform Distribution (Equal Probability)

- > Most commonly *used* in machine learning. (not most commonly occurring in data)
- > Uninteresting distribution, all events are equal.
- > Flat & Bounded.

$$f(x) = \begin{cases} \frac{1}{(b-a)} & \text{if } a \leq x \leq b \\ 0 & \text{if } x < a \text{ or } x > b \end{cases}$$

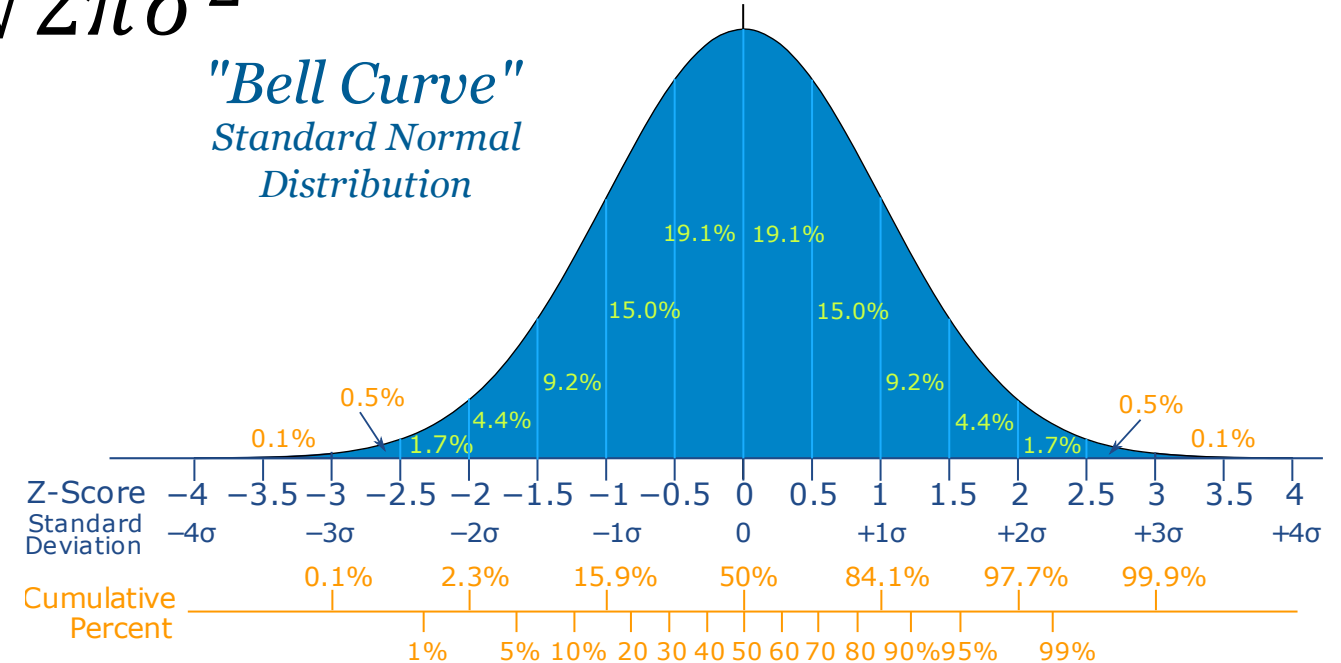
- > Mean =  $\frac{(a+b)}{2}$
- > Variance =  $(\frac{1}{12})(b-a)^2$
- > Very useful when we *need* to put in information about a data point or parameter, but we don't know much about it.



# Continuous Distribution: Normal Distribution (Gaussian/Bell Curve)

- > Most common & naturally occurring in data/nature.
- > Defined by a mean and variance only.

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{\frac{-(x-\mu)^2}{2\sigma^2}}$$

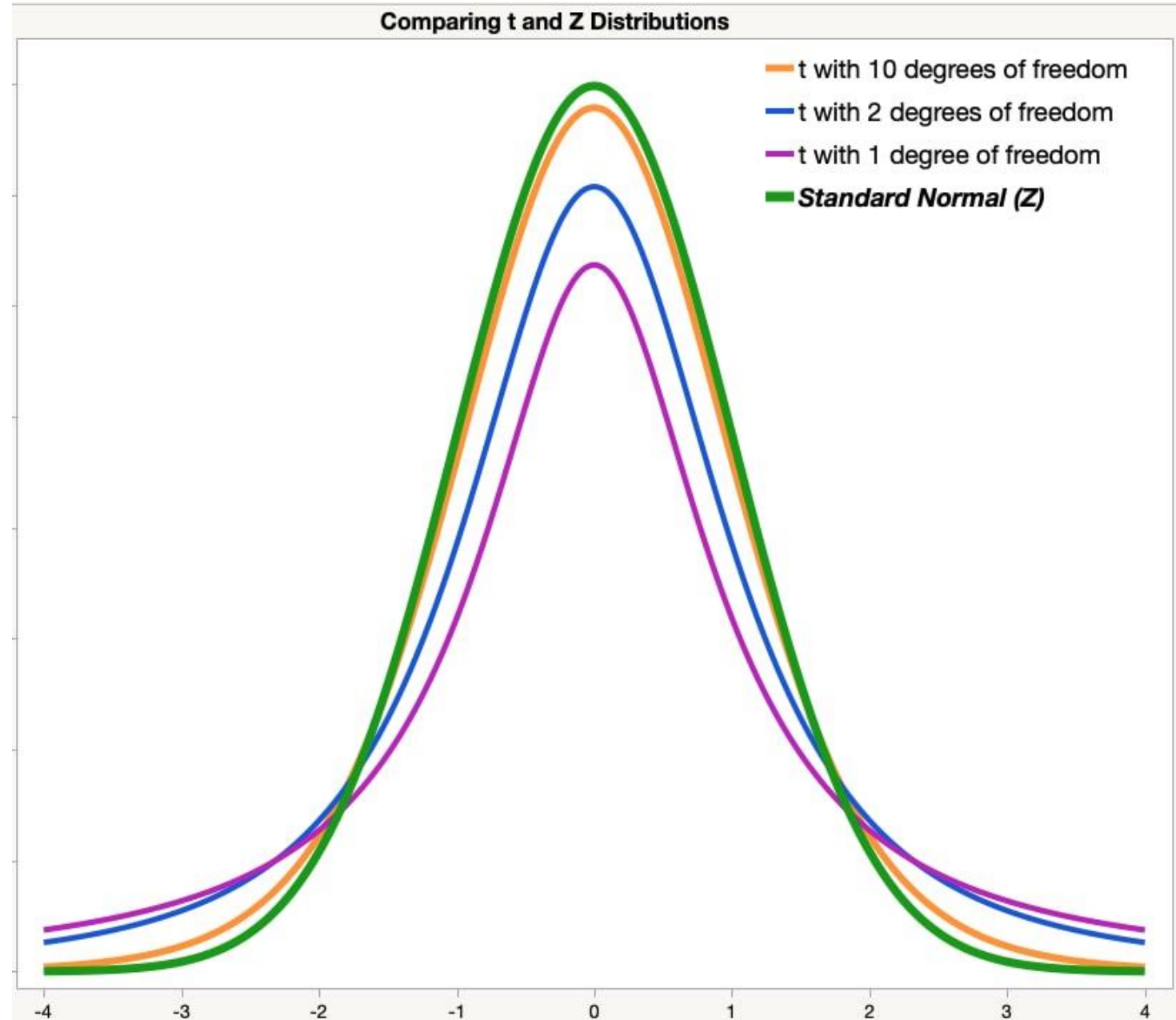


- > Has very nice properties.
- > Testing data for normality is VERY important.



# Continuous Distribution: Student's T Distribution

- > Important for investigating / testing smaller data sample sizes.
- > Used specifically for:
  - Testing of a mean-value when st.dev. is unknown.
  - Testing difference between two distribution means.
- > Looks very similar to the Normal distribution, but has longer tails.



# Distribution Transformations

- > The purpose of transforming a variable is to make it easier to distinguish between values (graphically, statistically, ML models...)
  - Most commonly, we are looking to transform it to a normal distribution.
- > Useful transformations are generally MONOTONIC
  - Functions used are never changing directions.
- > Common Transformations to consider:
  - Log based
    - >  $\log(x)$ ,  $\log(x+1)$ ,  $\log(x-\min(x)+1)$
  - N-th root based (of positive #s)
    - >  $X^{(1/n)}$
  - Any combination you can think of- but keep it monotonic.
- > How to determine? We will cover normality tests in a later class.

**notebook time**

**we return to the lecture later**

## the Monte Hall game

- you are presented with three doors: two have goats behind, one has a car behind
- the host knows which door has a car behind it, but you don't
- you select a door, then host would open one of the two other doors to reveal a goat
- you have a choice between sticking with the door you selected, or switching to the other door the host didn't open
- does it matter if you switch or not?

# Monty Hall Problem

- > Start with 3 doors. One prize behind unknown door. Pick a door.
- > Host reveals a separate door with no prize.
- > Then contestant can switch. Should they?

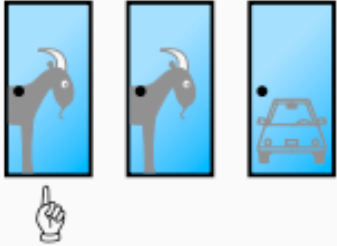
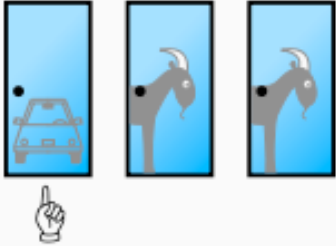
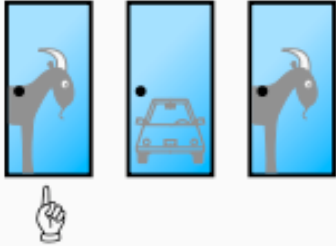
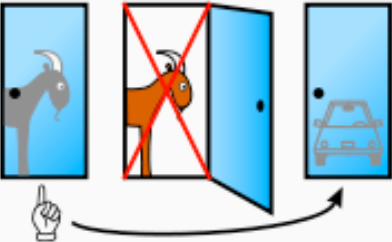
Car hidden behind Door 3	Car hidden behind Door 1	Car hidden behind Door 2
Player initially picks Door 1		
		
Host must open Door 2	Host randomly opens either goat door	Host must open Door 3



Let's consider this situation first

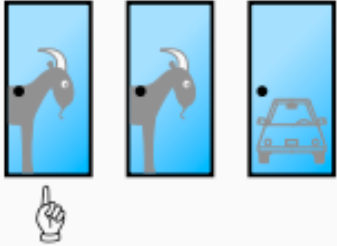
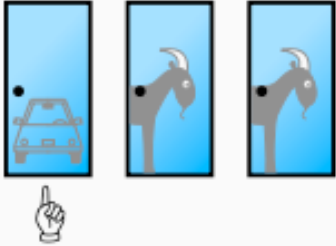
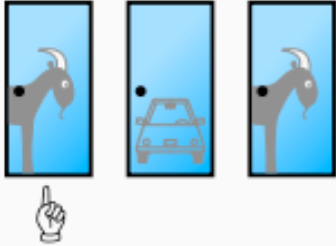
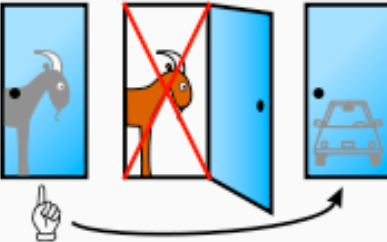

# Monty Hall Problem

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Car hidden behind Door 3	Car hidden behind Door 1	Car hidden behind Door 2
Player initially picks Door 1		
		
Host must open Door 2	Host randomly opens either goat door	Host must open Door 3
		
Probability 1/3		
Switching wins		

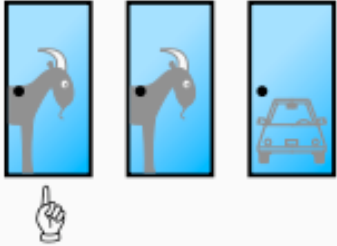
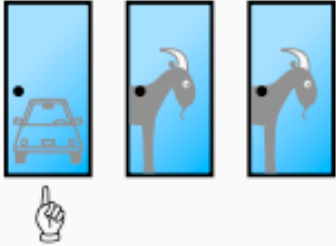
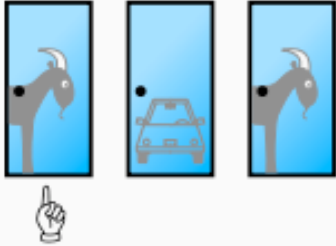
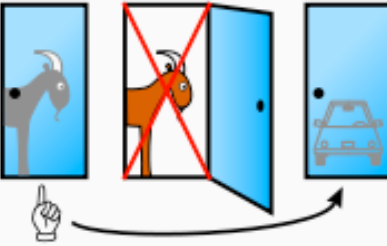
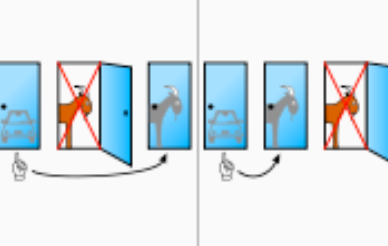
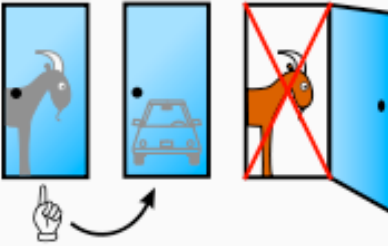
# Monty Hall Problem

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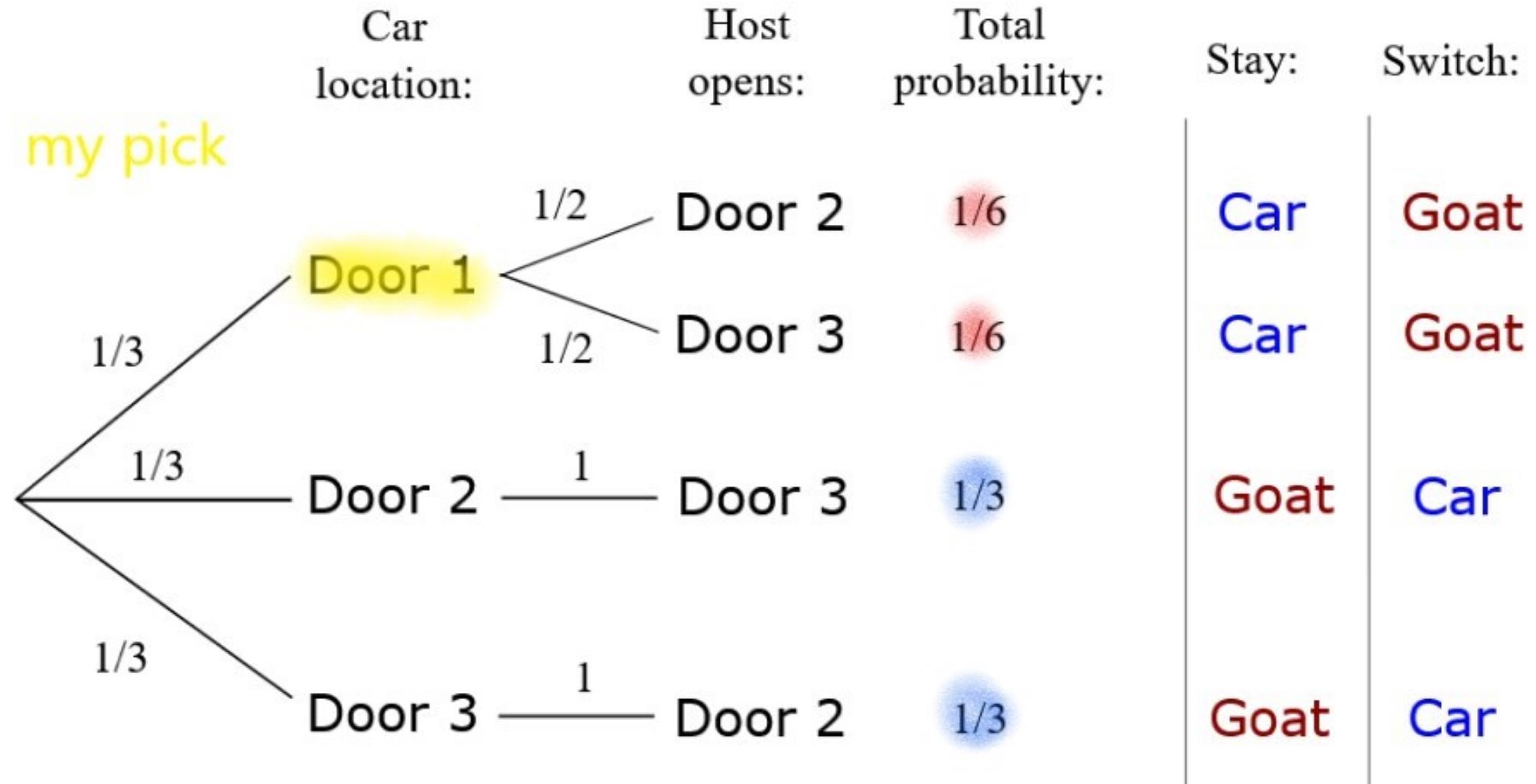
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Host must open Door 2	Host randomly opens either goat door	Host must open Door 3
		
Probability 1/3	Probability 1/6	Probability 1/6
Switching wins	Switching loses	Switching loses

# Monty Hall Problem

- > Start with 3 doors. One prize behind unknown door. Pick a door.
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Car hidden behind Door 3	Car hidden behind Door 1	Car hidden behind Door 2
Player initially picks Door 1		
		
Host must open Door 2	Host randomly opens either goat door	Host must open Door 3
		
Probability 1/3	Probability 1/6	Probability 1/6
Switching wins	Switching loses	Switching loses
If the host has opened Door 2, switching wins twice as often as staying		If the host has opened Door 3, switching wins twice as often as staying





- let  $C = \{1, 2, 3\}$  represent where the car is
- let  $H = \{1, 2, 3\}$  represent the door opened by host
- **assume by symmetry that we pick door 1**, and compute conditionals probabilities:
  - note that  $P(H = 1|C = 1) = 0$ , which leaves us with  $P(H = 2|C = 1) = 0.5$  and  $P(H = 3|C = 1) = 0.5$
  - note that  $P(H = 1|C = 2) = P(H = 1|C = 3) = 0$  because Montey won't open a door containing the car, leaving us with  $P(H = 3|C = 2) = 1$  and  $P(H = 2|C = 3) = 1$  respectively
- now we compute the joint probabilities  $P(H \text{ and } C) = P(H|C)P(C)$
- finally, we add up joint probabilities for goat and car under switch scenario

# Dealing with Missing Data (Intro)

- > Reasons for missing data
  - Recording failure (mechanical/software failures)
  - Reporting failure (human decisions)
  - Translation failure (data parsing / data transferring errors)
- > Many shapes and types of missing data:
  - Block missing – Physical shapes of data missing
    - > Examples: Time series- collection was down for some time.
  - Regular missing
    - > Rule based missing: Temperature gauge doesn't work below freezing.
  - Random missing
    - > Probability of missing data point equal across all data points.
  - Sparsity
    - > Very low observations. Hard to observe data (endangered species, rare events)
- > NOTE: Outliers *may* be treated as missing data. Depending on reasons.

# Dealing with Missing Data

Type	Benefits	Disadvantages	Notes
Drop Missing	Speed	Data Loss	
Mean/Median/Mode Fill	No Data Loss	Variation Reduction	
$X \sim F(\text{independents})$	More accurate No Data Loss	Slower	Needs most columns to be filled out Harder on truly independent data
KNN	More accurate No Data Loss	Slower Dependent on a Distance Function	
$X \sim F(y, \text{independents})$	Very Accurate No Data Loss	Slower Need Y	Can ONLY do this on a TRAINING dataset.

# Dealing with Missing Data in the Real World: Using Outside or New Data Sources as Supplements

- > Don't forget to explore outside or new data sources to help with missing data.
- > With the advent of free public data and bigger data sources, this is gaining popularity as a tool for imputation.
- > This includes using unstructured text as a major source of data.
- > Will eventually include asking Large-Language-Models to generate missing data.
- > Ex
  - Caesar's uses public reviews on websites to mine for customer sentiment about hotel rooms.
  - Zillow uses text descriptions of properties to fill in missing data about homes (# BDs, # BTHMs, sq footage, etc)
  - Subject to human stupidity.:

**Yelp Rating for Circus-Circus: 2/5**

**Text Description: "My son and I stayed here. The service and room was great, but it turns out my son is deathly afraid of clowns.**

# Breakout Discussion

- > Below are some example situations for missing data.
  - How would / could we address these? (Both statistically AND business wise)

Situations
An intern at your company accidentally deleted 10 rows in a large user-events table.
A set of door-to-door poll workers in a county were all sick over the same time period. (They are interviewing residents about news topic sentiments).
A malicious SQL injection attack dropped data from user-sign in logging records. Only from users who have a email that starts with the letter 'A' or 'a'.
The IoT coffee maker company you work for has a defect in the devices – it disconnects from wifi when boiling water is spilled in a certain spot.
A rental car company cannot collect GPS car locations during severe weather.

**the end**