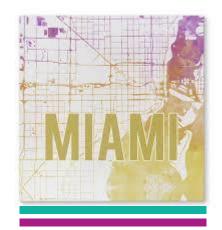


#### Probabilistic Models



- An experiment is an event whose outcome is not known with certainty
- The set of possible outcomes of an experiment is called the sample space which we will denote *S*
- The outcomes themselves are called sample points
- Examples of experiments
  - Flipping a coin  $\rightarrow S = \{H, T\}$
  - Tossing a die  $\rightarrow S = \{1,2,3,4,5,6\}$
  - Flipping a coin 10 times  $\rightarrow S = \{strings \ of \ length \ 10 \ with \ letters \ H \ \& \ T\}$
  - Time waiting on phone for airline to answer  $\rightarrow S = [0, \infty)$
  - Score in the next UNC basketball game  $\rightarrow S = \{(x, y): x, y \ge 0\}$
- Probability is a measure of how likely an event is to occur

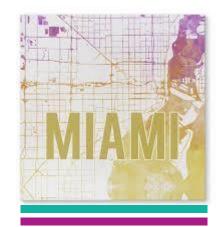


## Population vs. Samples



- The total population of an experiment is a set containing all observations
- A sample consists of a subset (usually randomly selected) of total population
- Total population of a random experiment that can be repeated an infinite number of times cannot be observed
- Q: What is an example of an experiment that can be infinitely repeated?
- If the total population is known, we can introduce randomness by considering the experiment of selecting one element (observation) of the population at a time
- The probability of selected an observation exhibiting "property x'' is

$$P(property \ x) = \frac{\# \ of \ elements \ exhibiting \ property \ x}{total \ population \ size}$$



## Interpretations of Probability

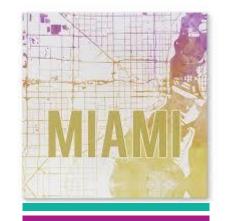
- Frequentist approach (classic)
  - Suppose we can repeat an experiment, under the exact conditions as many times as we want
  - We want to assign a value to how likely a specific outcome is
  - Compute the relative frequency of the desired outcome

$$\frac{\#\ of\ times\ outcome\ occurs}{\#\ of\ experiments}$$

• We can think of the probability as the limit of its relative frequency as the number of repetitions grows to infinity

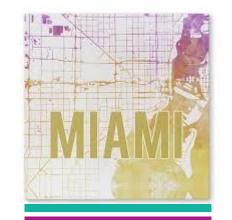
$$P(Outcome) = \lim_{n \to \infty} \frac{\# \ of \ times \ outcome \ occurs}{n}$$

where n is the number of times we repeat the experiment



# Interpretations of Probability

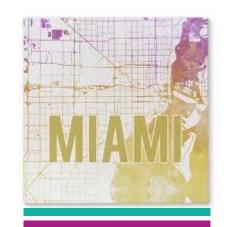
- Bayesian approach
  - Define probability as the degree of belief rather than the long-run frequency
  - Degree of belief is based off prior probability (subjective probability) and the relative frequency from observed data
  - Posterior probability is the updated belief on the probability of an event happening given the prior and data observed
- Difference between frequentist and Bayesian approach
  - Consider the experiment where we flip a coin
  - We want to find the probability of heads
  - Frequentist concludes probability is 0.5 under the belief that the relative frequency would get closer to 50% the more the coin is flipped
  - There is an assumption that out of the two outcomes both are equally likely
  - Bayesian would take the 50% as a prior belief with a lot of uncertainty until data has been gathered to back up the claim



#### **Probability Laws**



- A probability law P assigns to each event  $A \subseteq S$  a value in [0,1]
- Let  $\Omega = S$  be the universe and  $\emptyset$  denote the empty set
- Notation: U = "or",  $\cap = "and"$ , and  $A^c = "A complement"/"not A"$
- Axioms: Let A, B  $\subseteq \Omega$ 
  - $P(A) \ge 0$
  - If  $A \cap B = \emptyset$ , then  $P(A \cup B) = P(A) + P(B)$
  - $P(\Omega) = 1$
- Properties proven from axioms
  - If  $A \subseteq B = \emptyset$ , then  $P(A) \le P(B)$
  - $P(A \cup B) = P(A) + P(B) P(A \cap B)$
  - $P(\emptyset) = 0$
  - $P(A^c) = 1 P(A)$



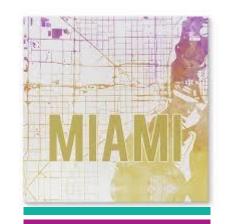
### Ex: Letter Grades in School



- School collected records of its 3,000 students
- Students in the science class have the following grade distribution (probability law)

Grade	Number of students	Probability
А	300	.10
В	600	.20
C	1500	.50
D	450	.15
F	150	.05

- Experiment = choose at random one of the 3000 students
- Q: What is the probability the student's grade in the science class is an A?
- Q: What is the probability the student's grade is C or higher?  $P(A \cup B \cup C) = P(A) + P(B) + P(C) = 0.1 + 0.2 + 0.5 = 0.8 = 1 P(D \cup F)$



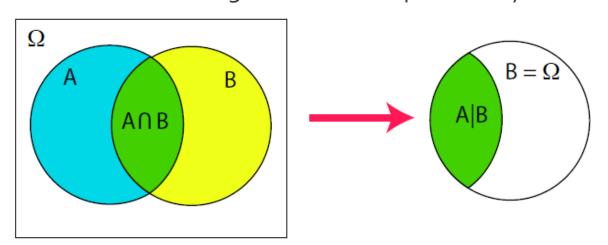
### Conditional Probability

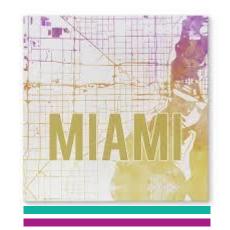


• For any events A and B in the sample space, with P(B) > 0, the conditional probability of event A given B is defined according to the formula

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

• Visual understanding of conditional probability





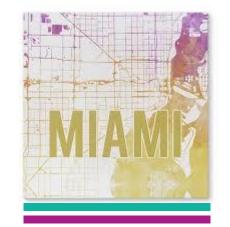
## Relationships Between Events



- Two events A and B are mutually exclusive if  $A \cap B = \emptyset$
- Mutually exclusive refers to events that cannot occur simultaneously
- Events of getting a 3 on a die roll and 4 on the same die roll are mutually exclusive
- Two events A and B are independent if  $P(A \cap B) = P(A) \times P(B)$
- If two events are independent,

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

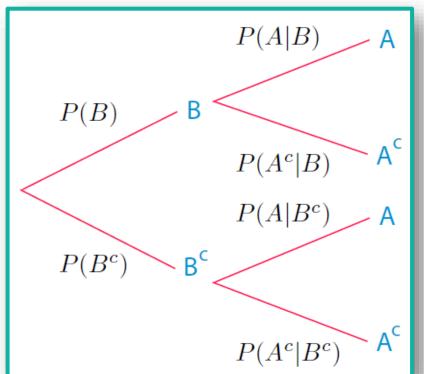
- Independence implies that the probability of a random event is not impacted at all by the occurrence of another event
- Events of getting a 3 on a die roll and a 4 on another die roll are independent

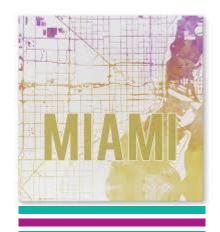


#### **Probability Trees**



- A probability tree is a diagram used to represent a probability space from a series of experiments (different or repetitive)
- Each path leads to a different outcome
- Numbers on path indicate probability
- Visualization of conditional probability
- Multiply probabilities along path to find the probabilities of different outcomes

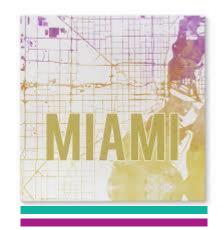




#### Ex: Flippin' Unfair Coins



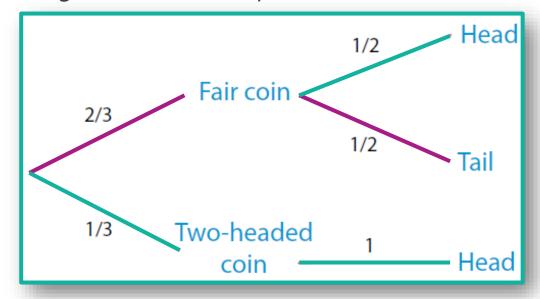
- A friend of yours has 3 coins in her pocket, two fair coins and one two-headed
- The two of you are trying to decide whether to watch "The Greatest Showman" or "Pitch Perfect" tonight
- You decide to flip a coin and go see "The Greatest Showman" if it is heads
- Your friend takes out one of the coins without looking and flips it
- Q: What is the probability that you go see "Pitch Perfect"?
- Q: What is the probability that you go see "The Greatest Showman"?



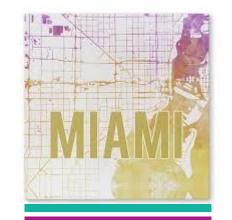
#### Ex: Flippin' Unfair Coins



• Diagram of this example



- Purple indicates the path to watching "Pitch Perfect"
- Teal indicates the path to watching "The Greatest Showman"



#### Ex: Flippin' Unfair Coins





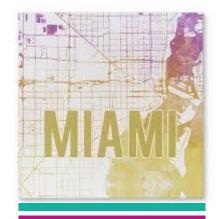
$$P(Pitch\ Perfect) = P(Tails) = P(Tails \cap Fair\ Coin)$$
$$= P(Tails|Fair\ Coin)P(Fair\ Coin) = \frac{1}{2} \times \frac{2}{3} = \frac{1}{3} = 0.3333$$

Probability of "The Greatest Showman"

 $P(The\ Greatest\ Showman) = P(Heads)$ 

- $= P(Heads \cap Fair\ Coin) + P(Heads \cap Unfair\ Coin)$
- $= P(Heads|Fair\ Coin)P(Fair\ Coin) + P(Heads|Unfair\ Coin)P(Unfair\ Coin)$

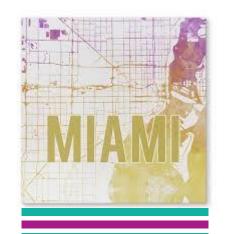
$$=\frac{1}{2}\times\frac{2}{3}+1\times\frac{1}{3}=\frac{1}{3}+\frac{1}{3}=\frac{2}{3}=0.6667=1-0.3333$$



#### Binomial Probability

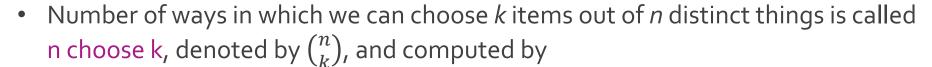


- Consider tossing a coin with probability of heads equal to p a total of 6 times
- Q: What is the probability that we get exactly 3 heads?
- We could express all possible outcomes of tossing a coin 6 times using a tree diagram that goes on forever but we all have lives
- Let's consider a few of the outcomes (sequences) where we get exactly 3 heads
- If A = Event of Exactly 3 Heads, then A =  $\{HHHTTT, TTTHHHH, HTHTHT, \dots\}$
- For each outcome where A occurs, the probability is  $F^3(1-p)^3$  because each coin flip is independent
- Q: How many such sequences exist where A occurs?



#### Binomial Probability





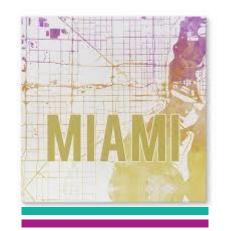
$$\binom{n}{k} = \frac{n!}{(n-k)! \, k!}$$

where 
$$n! = n \times (n-1) \times (n-2) \cdots 3 \times 2 \times 1$$
 (n factorial)

• The numbers  $\binom{n}{0}$ ,  $\binom{n}{1}$ ,  $\binom{n}{2}$ ,  $\cdots$ ,  $\binom{n}{n-1}$ ,  $\binom{n}{n}$  are called binomial coefficients, since

$$(a+b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$$

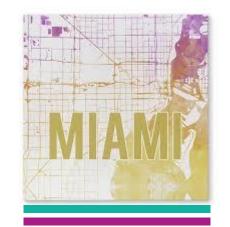
• From coin example, P(Exactly 3 Heads) = P(A) =  $\binom{6}{3}p^3(1-p)^3$ 



### Binomial Probability



- From coin example,  $P(\text{Exactly 3 Heads}) = P(A) = {6 \choose 3} p^3 (1 - p)^3$
- Bernouilli process is a repetition of fixed number of independent trials with a binary outcome where the probability of each outcome remains constant
- Each trial/experiment is called a Bernouilli trial
- For a Bernouilli process, the probability of k successes in n trials is  $\binom{n}{k}p^n(1-p)^{n-k}$
- These probabilities build the binomial distribution
- Excel formula is = BINOM.DIST(n, k, p, FALSE)









## The End





