#### Stars and Bars Poll

#### Mark whats correct.

- (A) ways to split n dollars among k:  $\binom{n+k-1}{k-1}$
- (B) ways to split k dollars among n:  $\binom{k+n-1}{n-1}$
- (C) ways to split 5 dollars among 3:  $\binom{7}{5}$
- (D) ways to split 5 dollars among 3:  $\binom{5+3-1}{3-1}$
- All correct.

# CS70: On to probability.

Modeling Uncertainty: Probability Space

- 1. Key Points
- 2. Random Experiments
- 3. Probability Space

#### Poll

#### Mark whats corect.

- (A)  $|10 \text{ digit numbers}| = 10^{10}$
- (B)  $|k| = 2^k$
- (C)  $|10 \text{ digit numbers}| = 9 * 10^9$
- (D) |n| digit base m numbers  $|m| = m^n$
- (E) |n| digit base m numbers  $|=(m-1)m^{n-1}$
- (A) or (C)? (D) or (E)? (B) are correct.

# **Key Points**

- ▶ Uncertainty does not mean "nothing is known"
- ► How to best make decisions under uncertainty?
  - Buv stocks
  - ► Detect signals (transmitted bits, speech, images, radar, diseases, etc.)
  - ► Control systems (Internet, airplane, robots, self-driving cars, schedule surgeries in a hospital, etc.)
- ► How to best use 'artificial' uncertainty?
  - ► Play games of chance
  - Design randomized algorithms.
- Probability
  - ► Models knowledge about uncertainty
  - Optimizes use of knowledge to make decisions

# Refresh: Counting.

First Rule of counting: Objects from a sequence of choices:

 $n_i$  possibilitities for *i*th choice.

 $n_1 \times n_2 \times \cdots \times n_k$  objects.

Second Rule of counting: If order does not matter.

Count with order. Divide by number of orderings/sorted object. Typically:  $\binom{n}{\nu}$ .

Stars and Bars: Sample k objects with replacement from n.

Order doesn't matter. k stars n-1 bars.

Typically:  $\binom{n+k-1}{k}$  or  $\binom{n+k-1}{n-1}$ .

Inclusion/Exclusion: two sets of objects.

Add number of each and then subtract intersection of sets.

Sum Rule: If disjoint just add.

Combinatorial Proofs: Identity from counting same in two ways. Pascal's Triangle Example:  $\binom{n+1}{k} = \binom{n}{k-1} + \binom{n}{k}$ .

RHS: Number of subsets of n+1 items size k.

LHS:  $\binom{n}{k-1}$  counts subsets of n+1 items with first item.  $\binom{n}{k}$  counts subsets of n+1 items without first item. Disjoint – so add!

# The Magic of Probability

Uncertainty: vague, fuzzy, confusing, scary, hard to think about. Probability:

Precise, unambiguous, simple(!) way to reason about uncertainty.





Uncertainty = Fear

Probability = Serenity

Our mission: help you discover the serenity of Probability, i.e., enable you to think clearly about uncertainty.

Your cost: focused attention and practice on examples and problems.

# Random Experiment: Flip one Fair Coin

Flip a fair coin: (One flips or tosses a coin)





- ▶ Possible outcomes: Heads (H) and Tails (T) (One flip yields either 'heads' or 'tails'.)
- ► Likelihoods: *H* : 50% and *T* : 50%

# Random Experiment: Flip one Unfair Coin

Flip an unfair (biased, loaded) coin:



- ▶ Possible outcomes: Heads (H) and Tails (T)
- Likelihoods:  $H: p \in (0,1)$  and T: 1-p
- Frequentist Interpretation:

Flip many times  $\Rightarrow$  Fraction 1 – p of tails

- ▶ Question: How can one figure out *p*? Flip many times
- ► Tautology? No: Statistical regularity!

# Random Experiment: Flip one Fair Coin

Flip a fair coin:





What do we mean by the likelihood of tails is 50%? Two interpretations:

- ► Single coin flip: 50% chance of 'tails' [subjectivist] Willingness to bet on the outcome of a single flip
- ► Many coin flips: About half yield 'tails' [frequentist] Makes sense for many flips
- Question: Why does the fraction of tails converge to the same value every time? Statistical Regularity! Deep!

# Random Experiment: Flip one Unfair Coin

Flip an unfair (biased, loaded) coin: model



# 01-Probability Model

# Random Experiment: Flip one Fair Coin

Flip a fair coin: model





Physical Experiment

Probability Model

- ► The physical experiment is complex. (Shape, density, initial momentum and position, ...)
- ► The Probability model is simple:
  - ▶ A set Ω of outcomes:  $Ω = {H, T}$ .
  - ► A probability assigned to each outcome: Pr[H] = 0.5, Pr[T] = 0.5.

# Flip Two Fair Coins

- ▶ Possible outcomes:  $\{HH, HT, TH, TT\} \equiv \{H, T\}^2$ .
- Note:  $A \times B := \{(a,b) \mid a \in A, b \in B\}$  and  $A^2 := A \times A$ .
- Likelihoods: 1/4 each.











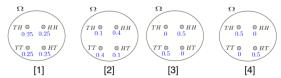
# Flip Glued Coins

Flips two coins glued together side by side:



- ▶ Possible outcomes: {HT, TH}.
- Likelihoods: *HT*: 0.5, *TH*: 0.5.
- Note: Coins are glued so that they show different faces.

# Flipping Two Coins



#### Important remarks:

- ► Each outcome describes the two coins.
- ▶ E.g., HT is one outcome of each of the above experiments.
- $\blacktriangleright$  Wrong to think that outcomes are  $\{H, T\}$  and that one picks twice from that set.
- Indeed, this viewpoint misses the relationship between the two
- ▶ Each ω ∈ Ω describes one outcome of the complete experiment.
- $ightharpoonup \Omega$  and the probabilities specify the random experiment.

# Flip two Attached Coins

Flips two coins attached by a spring:

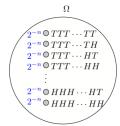


- ▶ Possible outcomes: {HH, HT, TH, TT}.
- Likelihoods: HH: 0.4, HT: 0.1, TH: 0.1, TT: 0.4.
- Note: Coins are attached so that they tend to show the same face, unless the spring twists enough.

# Flipping *n* times

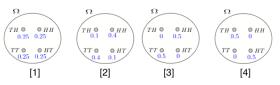
Flip a fair coin n times (some  $n \ge 1$ ):

- ▶ Possible outcomes:  $\{TT \cdots T, TT \cdots H, \dots, HH \cdots H\}$ . Thus,  $2^n$  possible outcomes.
- Note:  $\{TT \cdots T, TT \cdots H, \dots, HH \cdots H\} = \{H, T\}^n$ .  $A^n := \{(a_1, \dots, a_n) \mid a_1 \in A, \dots, a_n \in A\}. \ |A^n| = |A|^n.$
- ▶ Likelihoods: 1/2<sup>n</sup> each.



# Flipping Two Coins

Here is a way to summarize the four random experiments:

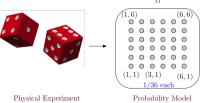


- Ω is the set of possible outcomes;
- Each outcome has a probability (likelihood);
- The probabilities are ≥ 0 and add up to 1;
- Fair coins: [1]; Glued coins: [3],[4]; Spring-attached coins: [2];

## Roll two Dice

Roll a balanced 6-sided die twice:

- ▶ Possible outcomes:  $\{1,2,3,4,5,6\}^2 = \{(a,b) \mid 1 \le a,b \le 6\}.$
- Likelihoods: 1/36 for each.



# Probability Space.

#### 1. A "random experiment":

- (a) Flip a biased coin;
- (b) Flip two fair coins;
- (c) Deal a poker hand.

#### 2. A set of possible outcomes: $\Omega$ .

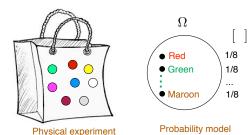
3. Assign a probability to each outcome:  $Pr: \Omega \rightarrow [0,1]$ .

(a) 
$$Pr[H] = p, Pr[T] = 1 - p$$
 for some  $p \in [0, 1]$   
(b)  $Pr[HH] = Pr[HT] = Pr[TH] = Pr[TT] = \frac{1}{4}$ 

(c)  $Pr[\underline{A \spadesuit A \lozenge A \clubsuit A \heartsuit K \spadesuit}] = \cdots = 1/\binom{52}{5}$ 

# Probability Space: Formalism

Simplest physical model of a uniform probability space:



A bag of identical balls, except for their color (or a label). If the bag is well shaken, every ball is equally likely to be picked.

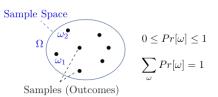
$$\Omega = \{ \text{white, red, yellow, grey, purple, blue, maroon, green} \}$$
 
$$Pr[\text{blue}] = \frac{1}{8}.$$

# Probability Space: formalism.

 $\Omega$  is the sample space.

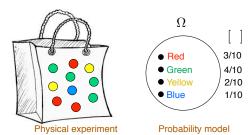
 $\omega \in \Omega$  is a **sample point**. (Also called an **outcome**.) Sample point  $\omega$  has a probability  $Pr[\omega]$  where

- ▶  $0 \le Pr[\omega] \le 1$ ;
- $\sum_{\omega \in \Omega} Pr[\omega] = 1.$



# Probability Space: Formalism

Simplest physical model of a non-uniform probability space:



$$\Omega = \{\text{Red, Green, Yellow, Blue}\}\$$

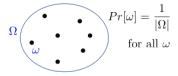
$$Pr[\text{Red}] = \frac{3}{10}, Pr[\text{Green}] = \frac{4}{10}, \text{ etc.}$$

Note: Probabilities are restricted to rational numbers:  $\frac{N_k}{N}$ .

# Probability Space: Formalism.

In a **uniform probability space** each outcome  $\omega$  is equally probable:  $Pr[\omega] = \frac{1}{|\Omega|}$  for all  $\omega \in \Omega$ .

### Uniform Probability Space

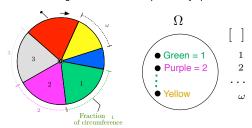


#### Examples:

- Flipping two fair coins, dealing a poker hand are uniform probability spaces.
- Flipping a biased coin is not a uniform probability space.

# Probability Space: Formalism

Physical model of a general non-uniform probability space:



Physical experiment

Probability model

The roulette wheel stops in sector  $\omega$  with probability  $p_{\omega}$ .

$$\Omega = \{1, 2, 3, ..., N\}, Pr[\omega] = p_{\omega}.$$

# An important remark

- ightharpoonup The random experiment selects one and only one outcome in  $\Omega$ .
- For instance, when we flip a fair coin twice
  - $ightharpoonup \Omega = \{HH, TH, HT, TT\}$
  - ▶ The experiment selects *one* of the elements of Ω.
- ▶ In this case, its wrong to think that  $\Omega = \{H, T\}$  and that the experiment selects two outcomes.
- Why? Because this would not describe how the two coin flips are related to each other.
- ➤ For instance, say we glue the coins side-by-side so that they face up the same way. Then one gets *HH* or *TT* with probability 50% each. This is not captured by 'picking two outcomes.'

#### CS70: On to Events.

Events, Conditional Probability, Independence, Bayes' Rule

Today: Events.

# Summary of Probability Basics

Modeling Uncertainty: Probability Space

- 1. Random Experiment
- 2. Probability Space:  $\Omega$ ;  $Pr[\omega] \in [0,1]$ ;  $\sum_{\omega} Pr[\omega] = 1$ .
- 3. Uniform Probability Space:  $Pr[\omega] = 1/|\Omega|$  for all  $\omega \in \Omega$ .

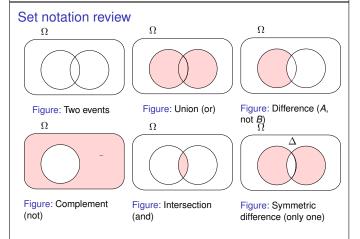
# **Probability Basics Review**

#### Setup:

- Random Experiment. Flip a fair coin twice.
- Probability Space.
  - ► Sample Space: Set of outcomes, Ω.  $Ω = \{HH, HT, TH, TT\}$  (Note: Not  $Ω = \{H, T\}$  with two picks!)
  - ▶ **Probability:**  $Pr[\omega]$  for all  $\omega \in \Omega$ .  $Pr[HH] = \cdots = Pr[TT] = 1/4$ 1.  $0 \le Pr[\omega] \le 1$ . 2.  $\sum_{\omega \in \Omega} Pr[\omega] = 1$ .

# Onwards in Probability.

Events, Conditional Probability, Independence, Bayes' Rule



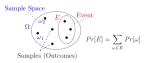
# Probability of exactly one 'heads' in two coin flips?

Idea: Sum the probabilities of all the different outcomes that have exactly one 'heads': HT, TH.

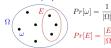
This leads to a definition!

#### Definition:

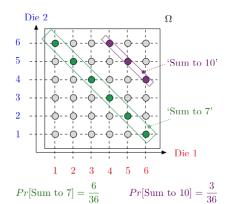
- ightharpoonup An **event,** E, is a subset of outcomes:  $E \subset \Omega$ .
- ▶ The **probability of** *E* is defined as  $Pr[E] = \sum_{\omega \in E} Pr[\omega]$ .



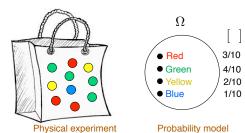
#### Uniform Probability Space



#### Roll a red and a blue die.



## **Event: Example**



$$\Omega = \{ \text{Red, Green, Yellow, Blue} \}$$
  
 $Pr[\text{Red}] = \frac{3}{10}, Pr[\text{Green}] = \frac{4}{10}, \text{ etc.}$ 

$$E = \{\textit{Red}, \textit{Green}\} \Rightarrow \textit{Pr}[E] = \frac{3+4}{10} = \frac{3}{10} + \frac{4}{10} = \textit{Pr}[\mathsf{Red}] + \textit{Pr}[\mathsf{Green}].$$

# Example and Polls: 20 coin tosses.

20 coin tosses

Sample space:  $\Omega = \text{set of 20 fair coin tosses.}$  $\Omega = \{T, H\}^{20} \equiv \{0, 1\}^{20}$ :  $|\Omega| = 2^{20}$ .

What is more likely?

(B)  $\omega_2 := (1,0,1,1,0,0,0,1,0,1,0,1,1,0,1,1,1,0,0,0)$ ?

What is more likely?

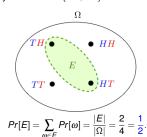
(A)  $(E_1)$  Twenty Hs out of twenty, or

(B) (E<sub>2</sub>) Ten Hs out of twenty?

# Probability of exactly one heads in two coin flips?

Sample Space,  $\Omega = \{HH, HT, TH, TT\}.$ 

Uniform probability space:  $Pr[HH] = Pr[HT] = Pr[TH] = Pr[TT] = \frac{1}{4}$ . Event, E, "exactly one heads":  $\{TH, HT\}$ .



# Example: 20 coin tosses.

20 coin tosses

Sample space:  $\Omega = \text{set of 20 fair coin tosses.}$  $\Omega = \{T, H\}^{20} \equiv \{0, 1\}^{20}; \ |\Omega| = 2^{20}.$ 

What is more likely?

Answer: Both are equally likely:  $Pr[\omega_1] = Pr[\omega_2] = \frac{1}{|\Omega|}$ .

What is more likely?

(E<sub>1</sub>) Twenty Hs out of twenty, or

(E<sub>2</sub>) Ten Hs out of twenty?

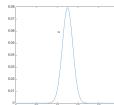
Answer: Ten Hs out of twenty.

Why? There are many sequences of 20 tosses with ten Hs; only one with twenty Hs.  $\Rightarrow Pr[E_1] = \frac{1}{|\Omega|} \ll Pr[E_2] = \frac{|E_2|}{|\Omega|}$ .

$$|E_2| = {20 \choose 10} = 184,756.$$

# Probability of *n* heads in 100 coin tosses.

$$\Omega = \{H,T\}^{100}; \ |\Omega| = 2^{100}.$$

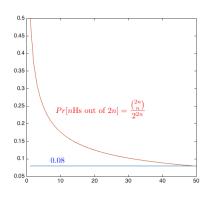


Event 
$$E_n$$
 = ' $n$  heads';  $|E_n| = \binom{100}{n}$ 

$$p_n := Pr[E_n] = \frac{|E_n|}{|\Omega|} = \frac{\binom{100}{n}}{2^{100}}$$

- Concentration around mean: Law of Large Numbers;
- ► Bell-shape: Central Limit Theorem.

# Exactly 50 heads in 100 coin tosses.



Exactly 50 heads in 100 coin tosses. Sample space:  $\Omega = \text{set of } 100 \text{ coin tosses} = \{H, T\}^{100}$ .  $|\Omega| = 2 \times 2 \times \cdots \times 2 = 2^{100}$ .

Uniform probability space:  $Pr[\omega] = \frac{1}{2^{100}}$ .

Event E = "100 coin tosses with exactly 50 heads"

|*E*|? Poll: (A) 100<sup>5</sup>0

(B) 2<sup>5</sup>0, (C) (<sup>50</sup><sub>100</sub>) (D) (<sup>100</sup><sub>50</sub>) (E) IDK - TBH.

Choose 50 positions out of 100 to be heads.

$$|E| = \binom{100}{50}$$
.

$$Pr[E] = \frac{\binom{100}{50}}{2^{100}}.$$

# Summary.

- 1. Random Experiment
- 2. Probability Space:  $\Omega$ ;  $Pr[\omega] \in [0,1]$ ;  $\sum_{\omega} Pr[\omega] = 1$ .
- 3. Uniform Probability Space:  $Pr[\omega] = 1/|\Omega|$  for all  $\omega \in \Omega$ .
- 4. Event: "subset of outcomes."  $A \subseteq \Omega$ .  $Pr[A] = \sum_{w \in A} Pr[\omega]$
- Some calculations.

#### Calculation.

Stirling formula (for large *n*):

$$n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$$
.

$$\binom{2n}{n} \approx \frac{\sqrt{4\pi n} (2n/e)^{2n}}{[\sqrt{2\pi n} (n/e)^n]^2} \approx \frac{4^n}{\sqrt{\pi n}}.$$

$$Pr[E] = \frac{|E|}{|\Omega|} = \frac{|E|}{2^{2n}} = \frac{1}{\sqrt{\pi n}} = \frac{1}{\sqrt{50\pi}} \approx .08.$$