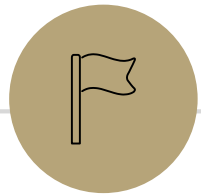


# Random Variables

CSE 312 Spring 24  
Lecture 10



# Random Variables



# Random Variable

What's a random variable?

Formally

## Random Variable

$X: \Omega \rightarrow \mathbb{R}$  is a random variable  
 $X(\omega)$  is the summary of the outcome  $\omega$

Informally: A random variable is a way to summarize the important (numerical) information from your outcome.

Support  $\Omega_X$   
the set of values  
 $X$  can take.

Probability Mass  
Function  
(pmf  $p_X(x)$ )  
on input  $x$ , tells  
you  $\mathbb{P}(X = x)$ .

$$p_X(7) = \frac{1}{6}$$

# Describing a Random Variable

The most common way to describe a random variable is the PMF.

But there's a second representation:

The cumulative distribution function (CDF) gives the probability  $X \leq x$

More formally,  $\mathbb{P}(\{\omega: X(\omega) \leq x\})$

Often written  $F_X(x) = \mathbb{P}(X \leq x)$

$$F_X(x) = \sum_{i:i \leq x} p_X(i)$$

# Try it yourself

What is the CDF of  $X$  where

$X$  be the largest value among the three balls. (Drawing 3 of the 20 without replacement)

$$\frac{\binom{7}{3}}{\binom{20}{3}}$$

A handwritten number 7 is written above the fraction, with a blue arrow pointing from it to the numerator  $\binom{7}{3}$ .

$$P(X \leq x)$$



$$P(B_1 \leq x \cap B_2 \leq x \cap B_3 \leq x)$$

$$F_X(x) =$$

$$\begin{cases} 0 & \text{if } x < 3 \\ \frac{\binom{\lfloor x \rfloor}{3}}{\binom{20}{3}} & \\ 1 & \text{if } x > 20 \end{cases}$$

# Try it yourself

What is the CDF of  $X$  where

$X$  be the largest value among the three balls. (Drawing 3 of the 20 without replacement)

$$F_X(x) = \begin{cases} 0 & \text{if } x < 3 \\ \binom{\lfloor x \rfloor}{3} / \binom{20}{3} & \text{if } 3 \leq x \leq 20 \\ 1 & \text{otherwise} \end{cases}$$

# Try it yourself

What is the CDF of  $X$  where

$X$  be the largest value among the three balls. (Drawing 3 of the 20 without replacement)

$$F_X(x) = \begin{cases} 0 & \text{if } x < 3 \\ \binom{\lfloor x \rfloor}{3} / \binom{20}{3} & \text{if } 3 \leq x \leq 20 \\ 1 & \text{otherwise} \end{cases}$$

Good checks: Is  $F_X(\infty) = 1$ ? If not, something is wrong.

Is  $F_X(x)$  increasing? If not something is wrong.

Is  $F_X(x)$  defined for all real number inputs? If not something is wrong.

# Two descriptions

## PROBABILITY MASS FUNCTION

Defined for all  $\mathbb{R}$  inputs.

Usually has "0 otherwise" as an extra case.

$$\sum_x p_X(x) = 1$$

$$0 \leq p_X(x) \leq 1$$

$$\sum_{z:z \leq x} p_X(z) = F_X(x)$$

## CUMULATIVE DISTRIBUTION FUNCTION

Defined for all  $\mathbb{R}$  inputs.

Often has "0 otherwise" and 1 otherwise" extra cases

Non-decreasing function

$$0 \leq F_X(x) \leq 1$$

$$\lim_{x \rightarrow -\infty} F_X(x) = 0$$

$$\lim_{x \rightarrow \infty} F_X(x) = 1$$



# More Random Variable Practice

Roll a fair die  $n$  times. Let  $Z$  be the number of rolls that are 5s or 6s.

What is the pmf?

Don't try to write the CDF...it's a mess...

Or try for a few minutes to realize it isn't nice.

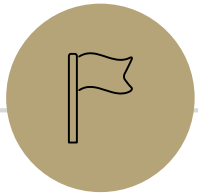
$$P_Z(z) = \binom{n}{z} \left(\frac{1}{3}\right)^z \left(\frac{2}{3}\right)^{n-z}$$

# More Random Variable Practice

Roll a fair die  $n$  times. Let  $Z$  be the number of rolls that are 5s or 6s.

What's the probability of getting exactly  $k$  5's/6's? Well we need to know which  $k$  of the  $n$  rolls are 5's/6's. And then multiply by the probability of getting exactly that outcome

$$p_Z(z) = \begin{cases} \binom{n}{z} \cdot \left(\frac{1}{3}\right)^z \left(\frac{2}{3}\right)^{n-z} & \text{if } z \in Z, 0 \leq z \leq n \\ 0 & \text{otherwise} \end{cases}$$



# Expectation

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# Expectation



## Expectation

The "expectation" (or "expected value") of a random variable  $X$  is:

$$\mathbb{E}[X] = \sum_k k \cdot \mathbb{P}(X = k)$$

Intuition: The weighted average of values  $X$  could take on.

Weighted by the probability you actually see them.

# Example 1

Flip a fair coin twice (independently)

Let  $X$  be the number of heads.

$$\Omega_X = \{0, 1, 2\}$$

$\Omega = \{TT, TH, HT, HH\}$ ,  $\mathbb{P}()$  is uniform measure.

$$\mathbb{E}[X] = \frac{1}{4} \cdot 0 + \frac{1}{2} \cdot 1 + \frac{1}{4} \cdot 2 = 0 + \frac{1}{2} + \frac{1}{2} = 1.$$

# Example 2

You roll a biased die.

It shows a 6 with probability  $\frac{1}{3}$ , and 1,...,5 with probability  $\frac{2}{15}$  each.

Let  $X$  be the value of the die. What is  $\mathbb{E}[X]$ ?

$$\begin{aligned} & \frac{1}{3} \cdot 6 + \frac{2}{15} \cdot 5 + \frac{2}{15} \cdot 4 + \frac{2}{15} \cdot 3 + \frac{2}{15} \cdot 2 + \frac{2}{15} \cdot 1 \\ & = 2 + \frac{2(5+4+3+2+1)}{15} = 2 + \frac{30}{15} = 4 \end{aligned}$$

$\mathbb{E}[X]$  is not just the most likely outcome!

# Try it yourself

Let  $X$  be the result shown on a fair die. What is  $\mathbb{E}[X]$ ?

Let  $Y$  be the sum of two (independent) fair die rolls. What is  $\mathbb{E}[Y]$ ?

~~Fill out the poll everywhere so Robbie  
knows how long to explain  
Go to [pollev.com/robbie](https://pollev.com/robbie)~~

# Try it yourself

Let  $X$  be the result shown on a fair die. What is  $\mathbb{E}[X]$

$$6 \cdot \frac{1}{6} + 5 \cdot \frac{1}{6} + 4 \cdot \frac{1}{6} + 3 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + 1 \cdot \frac{1}{6}$$
$$= \frac{21}{6} = 3.5$$

$\mathbb{E}[X]$  is not necessarily a possible outcome!

That's ok, it's an average!



# Try it yourself

$$\mathbb{E}[Y] =$$

$$\frac{1}{36} \cdot 2 + \frac{2}{36} \cdot 3 + \frac{3}{36} \cdot 4 + \frac{4}{36} \cdot 5 + \frac{5}{36} \cdot 6 + \frac{6}{36} \cdot 7 + \frac{5}{36} \cdot 8 + \frac{4}{36} \cdot 9 + \frac{3}{36} \cdot 10 + \frac{2}{36} \cdot 11 + \frac{1}{36} \cdot 12$$

$$= 7$$

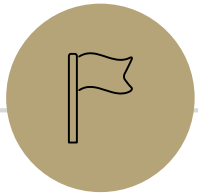
$\mathbb{E}[Y] = 2\mathbb{E}[X]$ . That's not a coincidence...we'll talk about why on Friday.

# Subtle but Important

$X$  is random. You don't know what it is (at least until you run the experiment).

$\mathbb{E}[X]$  is not random. It's a number.

You don't need to run the experiment to know what it is.



**More Independence**

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# Independence of events

Recall the definition of independence of events:

## Independence

Two events  $A, B$  are independent if

$$\mathbb{P}(A \cap B) = \mathbb{P}(A) \cdot \mathbb{P}(B)$$

# Independence for 3 or more events

For three or more events, we need two kinds of independence

## Pairwise Independence

Events  $A_1, A_2, \dots, A_n$  are pairwise independent if

$$\mathbb{P}(A_i \cap A_j) = \mathbb{P}(A_i) \cdot \mathbb{P}(A_j) \text{ for all } i, j$$

## Mutual Independence

Events  $A_1, A_2, \dots, A_n$  are mutually independent if

$$\mathbb{P}(A_{i_1} \cap A_{i_2} \cap \dots \cap A_{i_k}) = \mathbb{P}(A_{i_1}) \cdot \mathbb{P}(A_{i_2}) \cdots \mathbb{P}(A_{i_k})$$

for every subset  $\{i_1, i_2, \dots, i_k\}$  of  $\{1, 2, \dots, n\}$ .

# Pairwise Independence vs. Mutual Independence

Roll two fair dice (one red one blue) independently

$R$  = "red die is 3"

$B$  = "blue die is 5"

$S$  = "sum is 7"

How should we describe these events?

# Pairwise Independence

$R, B, S$  are pairwise independent

$$\mathbb{P}(R \cap B) \stackrel{?}{=} \mathbb{P}(R)\mathbb{P}(B)$$

$$\frac{1}{36} = \frac{1}{6} \cdot \frac{1}{6} \text{ Yes! (These are also independent by the problem statement)}$$

$$\mathbb{P}(R \cap S) \stackrel{?}{=} \mathbb{P}(R)\mathbb{P}(S)$$

$$\frac{1}{36} \stackrel{?}{=} \frac{1}{6} \cdot \frac{1}{6} \text{ Yes!}$$

$$\mathbb{P}(B \cap S) \stackrel{?}{=} \mathbb{P}(B)\mathbb{P}(S)$$

$$\frac{1}{36} \stackrel{?}{=} \frac{1}{6} \cdot \frac{1}{6} \text{ Yes!}$$

Since all three pairs are independent, we say the random variables are pairwise independent.

# Mutual Independence

$R, B, S$  are not mutually independent.

$\mathbb{P}(R \cap B \cap S) = 0$ ; if the red die is 3, and blue die is 5 then the sum is 8 (so it can't be 7)

$$\mathbb{P}(R)\mathbb{P}(B)\mathbb{P}(S) = \left(\frac{1}{6}\right)^3 = \frac{1}{216} \neq 0$$



# Checking Mutual Independence

It's not enough to check just  $\mathbb{P}(A \cap B \cap C)$  either.

Roll a fair 8-sided die.

Let  $A$  be  $\{1,2,3,4\}$

$B$  be  $\{2,4,6,8\}$

$C$  be  $\{2,3,5,7\}$

$$\mathbb{P}(A \cap B \cap C) = \mathbb{P}(\{2\}) = \frac{1}{8}$$

$$\mathbb{P}(A)\mathbb{P}(B)\mathbb{P}(C) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{8}$$

# Checking Mutual Independence

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$C$  be  $\{2,3,5,7\}$

$$\mathbb{P}(A \cap B \cap C) = \mathbb{P}(\{2\}) = \frac{1}{8}$$

$$\mathbb{P}(A)\mathbb{P}(B)\mathbb{P}(C) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{8}$$

But  $B$  and  $C$  aren't independent. Because there's a subset that's not independent,  $A, B, C$  are not mutually independent.

# Checking Mutual Independence

To check mutual independence of events:

Check **every** subset.

To check pairwise independence of events:

Check **every** subset of size two.

# Independence of Random Variables

That's for events...what about random variables?

## Independence (of random variables)

$X$  and  $Y$  are independent if for all  $k, \ell$

$$\mathbb{P}(X = k, Y = \ell) = \mathbb{P}(X = k)\mathbb{P}(Y = \ell)$$

We'll often use commas instead of  $\cap$  symbol.

# Independence of Random Variables

The “for all values” is important.

We say that the event “the sum is 7” is independent of “the red die is 5”  
What about  $S$  = “the sum of two dice” and  $R$  = “the value of the red die”

# Independence of Random Variables

The “for all values” is important.

We say that the event “the sum is 7” is independent of “the red die is 5”  
What about  $S$  = “the sum of two dice” and  $R$  = “the value of the red die”

NOT independent.

$$\mathbb{P}(S = 2, R = 5) \neq \mathbb{P}(S = 2)\mathbb{P}(R = 5) \text{ (for example)}$$

# Independence of Random Variables

Flip a coin independently  $2n$  times.

Let  $X$  be “the number of heads in the first  $n$  flips.”

Let  $Y$  be “the number of heads in the last  $n$  flips.”

$X$  and  $Y$  are independent.

# Mutual Independence for RVs

A little simpler to write down than for events

## Mutual Independence (of random variables)

$X_1, X_2, \dots, X_n$  are mutually independent if for all  $x_1, x_2, \dots, x_n$

$$\mathbb{P}(X_1 = x_1, X_2 = x_2, \dots, X_n = x_n) = \mathbb{P}(X_1 = x_1)\mathbb{P}(X_2 = x_2) \cdots \mathbb{P}(X_n = x_n)$$

DON'T need to check all subsets for random variables...

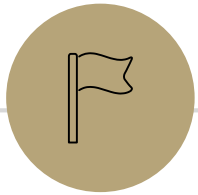
But you do need to check all values (all possible  $x_i$ ) still.



# What does Independence give you?

$$\mathbb{E}[XY] = \mathbb{E}[X] \cdot \mathbb{E}[Y]$$

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$$



## More Practice: Infinite sequential processes

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# Infinite sequential process

In volleyball, sets are played first team to

- Score 25 points
- Lead by at least 2

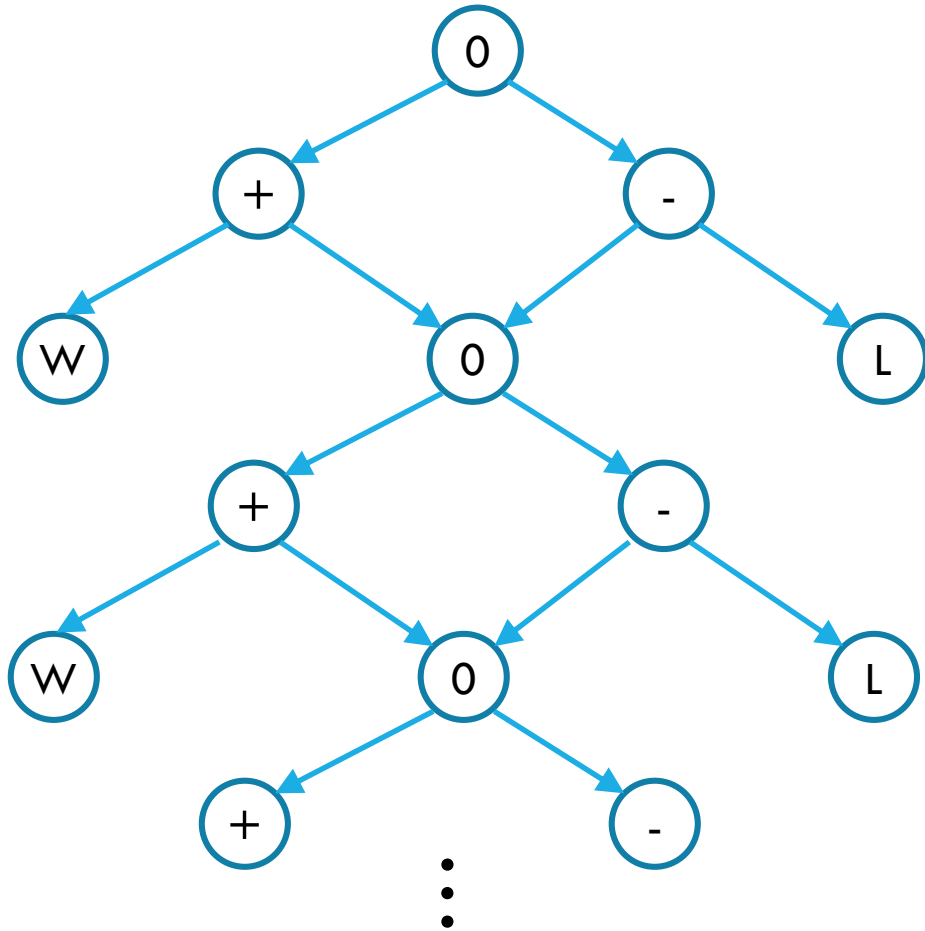
At the same time wins a set.

Suppose a set is 23-23. Your team wins each point independently with probability  $p$ .

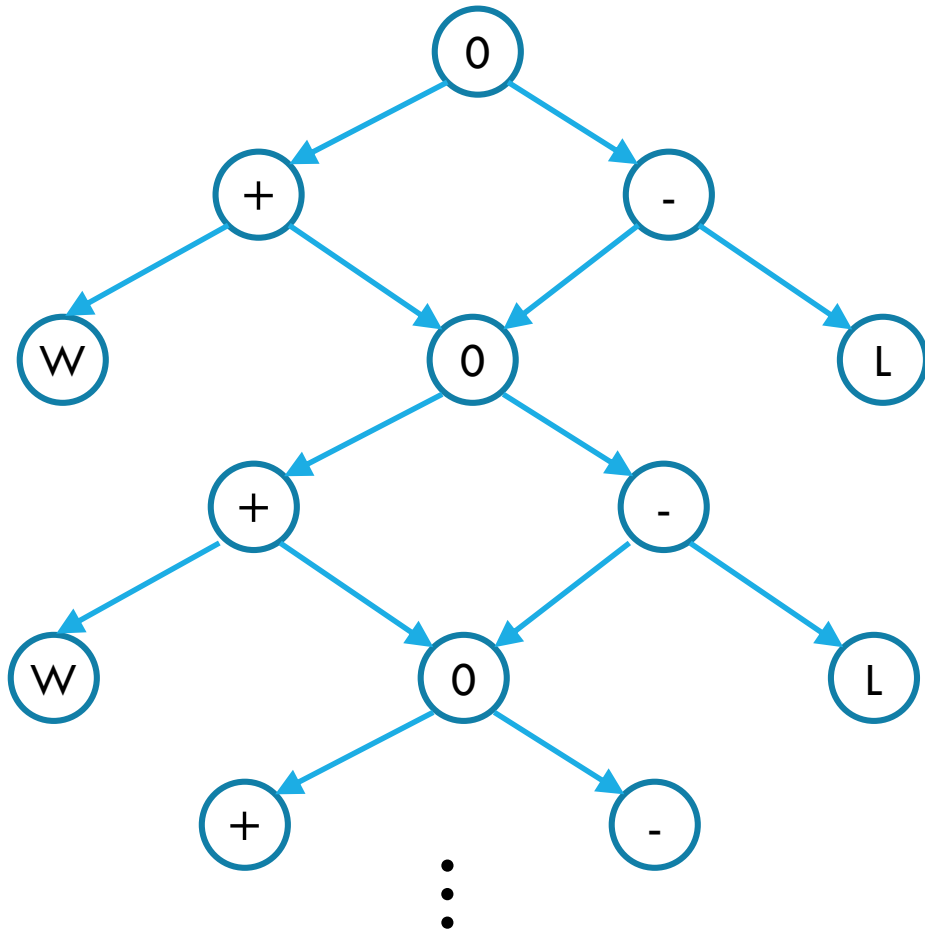
What is the probability your team wins the set?

# Sequential Process

$$\mathbb{P}(\text{win from even}) = p^2 + 2p(1-p)\mathbb{P}(\text{win from even})$$



# Sequential Process



$$\mathbb{P}(\text{win from even}) = p^2 + 2p(1 - p)\mathbb{P}(\text{win from even})$$

$$x - x[2p - p^2] = p^2$$

$$x[1 - 2p + p^2] = p^2$$

$$x = \frac{p^2}{p^2 - 2p + 1}$$