

# More Random Variables Expectation, Independence

CSE 312 Winter 26  
Lecture 10

# Two descriptions

$$P(X=x)$$

## PROBABILITY MASS FUNCTION

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

Usually has “0 otherwise” as an extra case.

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

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

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

$$P(X \leq k)$$

## CUMULATIVE DISTRIBUTION FUNCTION

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

Often has “0 otherwise” and 1 otherwise” extra cases

Non-decreasing function

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

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

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

# Expectation

$$E[X]$$

## 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.

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

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

## Example 2

You roll a biased die.

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

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

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

$\mathbb{E}[X]$  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  $\mathbb{E}[Y]$ ?

# 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 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.

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

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# 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?

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


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


What does  $XY$  mean? I rolled two dice, let  $X$  be the red die,  $Y$  the blue die.  $XY$  is the random variable that tells you the product of the two dice.

That's a function that takes in an outcome and gives you a number back...so a random variable!! (Same for  $X + Y$ ).

# Functions of a random variable

Let  $X, Y$  be random variables defined on the same sample space.

Functions of  $X$  and/or  $Y$  like

$$\underline{X + Y}$$

$$\underline{X^2}$$

$$\underline{2X + 3}$$

Etc.

$$\mathbb{E} [2X + 3]$$

**Are** random variables! (Say what the outcome is, and these functions give you a number. They're functions from  $\Omega \rightarrow \mathbb{R}$ . That's the definition of a random variable!)

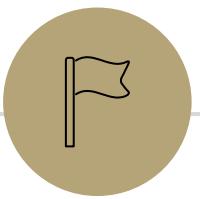
# Expectations of functions of random variables

Let's say we have a random variable  $X$  and a function  $g$ . What is  $\mathbb{E}[g(X)]$ ?

$$\mathbb{E}[g(X)] = \sum_{\omega \in \Omega} g(X(\omega)) \cdot \mathbb{P}(\omega)$$

$$\text{Equivalently: } \mathbb{E}[g(X)] = \sum_{k \in \Omega_{g(X)}} k \cdot \mathbb{P}(g(X) = k)$$

Notice that  $\mathbb{E}[g(X)]$  might not be  $g(\mathbb{E}[X])$ .



**Variance**

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# Where are we?

A random variable is a way to summarize what outcome you saw.

The Expectation of a random variable is its average value.

A way to summarize a random variable

# Variance

Another one number summary of a random variable.

But wait, we already have expectation, what's this for?

# Consider these two games

Would you be willing to play these games?

Game 1: I will flip a fair coin; if it's heads, I pay you \$1. If it's tails, you pay me \$1. Let  $\underline{X_1}$  be your profit if you play game 1

Game 2: I will flip a fair coin; if it's heads, I pay you \$10,000. If it's tails, you pay me \$10,000. Let  $X_2$  be your profit if you play game 2.

Both games are "fair" ( $\mathbb{E}[X_1] = \mathbb{E}[X_2] = 0$ )

# What's the difference

Expectation tells you what the average will be...

But it doesn't tell you how "extreme" your results could be.

Nor how likely those extreme results are.

Game 2 has many (well, only) very extreme results.

In expectation they "cancel out" but if you can only play once...

...it would be nice to measure that.

# Designing a Measure – Try 1

Well let's measure how far all the events are away from the center, and how likely they are

$$\sum_{\omega} \underline{\mathbb{P}(\omega)} \cdot (X(\omega) - \mathbb{E}[X])$$

What happens with Game 1?

$$\frac{1}{2} \cdot (1 - 0) + \frac{1}{2} \cdot (-1 - 0) \\ \frac{1}{2} - \frac{1}{2} = 0$$

What happens with Game 2?

$$\frac{1}{2} \cdot (100000 - 0) + \frac{1}{2} \cdot (-100000 - 0) \\ 5000 - 5000 = 0$$

# Designing a Measure – Try 2

How do we prevent cancelling? Squaring makes everything positive.

$$\sum_{\omega} \mathbb{P}(\omega) \cdot (X(\omega) - \mathbb{E}[X])^2$$


What happens with Game 1?

$$\frac{1}{2} \cdot (1 - 0)^2 + \frac{1}{2} \cdot (-1 - 0)^2$$
$$\frac{1}{2} + \frac{1}{2} = 1$$


What happens with Game 2?

$$\frac{1}{2} \cdot (100000 - 0)^2 + \frac{1}{2} \cdot (-100000 - 0)^2$$
$$5,000,000,000 + 5,000,000,000 = 10^{10}$$


# Why Squaring

Why not absolute value? Or Fourth power?

Squaring is nicer algebraically.

Our goal with variance was to talk about the spread of results. Squaring makes extreme results even more extreme.

Fourth power over-emphasizes the extreme results (for our purposes).

# Variance

## Variance

The variance of a random variable  $X$  is

$$\text{Var}(X) = \sum_{\omega} \mathbb{P}(\omega) \cdot (X(\omega) - \mathbb{E}[X])^2 = \mathbb{E}[(X - \mathbb{E}[X])^2] = \mathbb{E}[X^2] - \mathbb{E}[X]^2$$

The first two forms are the definition. The last one is an algebra trick.

# Variance of a die

Let  $X$  be the result of rolling a fair die.

$$\begin{aligned}\text{Var}(X) &= \mathbb{E}[(X - \mathbb{E}[X])^2] = \mathbb{E}[(X - 3.5)^2] \\ &= \frac{1}{6}(1 - 3.5)^2 + \frac{1}{6}(2 - 3.5)^2 + \frac{1}{6}(3 - 3.5)^2 + \frac{1}{6}(4 - 3.5)^2 + \frac{1}{6}(5 - 3.5)^2 + \frac{1}{6}(6 - 3.5)^2 \\ &= \frac{35}{12} \approx 2.92.\end{aligned}$$

$$\text{Or } \mathbb{E}[X^2] - (\mathbb{E}[X])^2 = \sum_{k=1}^6 \frac{1}{6} \cdot k^2 - 3.5^2 = \frac{91}{6} - 3.5^2 \approx 2.92$$

# Variance of $n$ Coin Flips

Flip a coin  $n$  times, where it comes up heads with probability  $p$  each time (independently). Let  $X$  be the total number of heads.

We'll see next time  $\mathbb{E}[X] = np$ .

Also define:  $X_i = \begin{cases} 1 & \text{if flip } i \text{ is heads} \\ 0 & \text{otherwise} \end{cases}$

# Variance of $n$ Coin Flips

Flip a coin  $n$  times, where it comes up heads with probability  $p$  each time (independently). Let  $X$  be the total number of heads.

What about  $\text{Var}(X)$

$$\begin{aligned}\mathbb{E}[(X - \mathbb{E}[X])^2] &= \sum_{\omega} \mathbb{P}(\omega)(X(\omega) - np)^2 \\ &= \sum_{k=0}^n \binom{n}{k} \cdot p^k (1-p)^{n-k} \cdot (k - np)^2\end{aligned}$$

Algebra time?

# Variance

If  $X$  and  $Y$  are independent then  
 $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$

Are the  $X_i$  independent? Yes!

In this problem  $X_i$  is independent of  $X_j$  for  $i \neq j$  where

$$X_i = \begin{cases} 1 & \text{if flip } i \text{ was heads} \\ 0 & \text{otherwise} \end{cases}$$

# Variance

$$\text{Var}(X) = \text{Var}(\sum_{i=1}^n X_i) = \sum_{i=1}^n \text{Var}(X_i)$$

What's the  $\text{Var}(X_i)$ ?

$$\mathbb{E}[(X_i - \mathbb{E}[X_i])^2]$$

$$= \mathbb{E}[(X_i - p)^2]$$

$$= p(1-p)^2 + (1-p)(0-p)^2$$

$$= p(1-p)[(1-p) + p] = p(1-p).$$

$$\text{OR } \text{Var}(X_i) = \mathbb{E}[X_i^2] - \mathbb{E}[X_i]^2 = \mathbb{E}[X_i] - p^2 = p - p^2 = p(1-p).$$

# Plugging In

$$\text{Var}(X) = \text{Var}(\sum_{i=1}^n X_i) = \sum_{i=1}^n \text{Var}(X_i)$$

What's the  $\text{Var}(X_i)$ ?

$$p(1 - p).$$

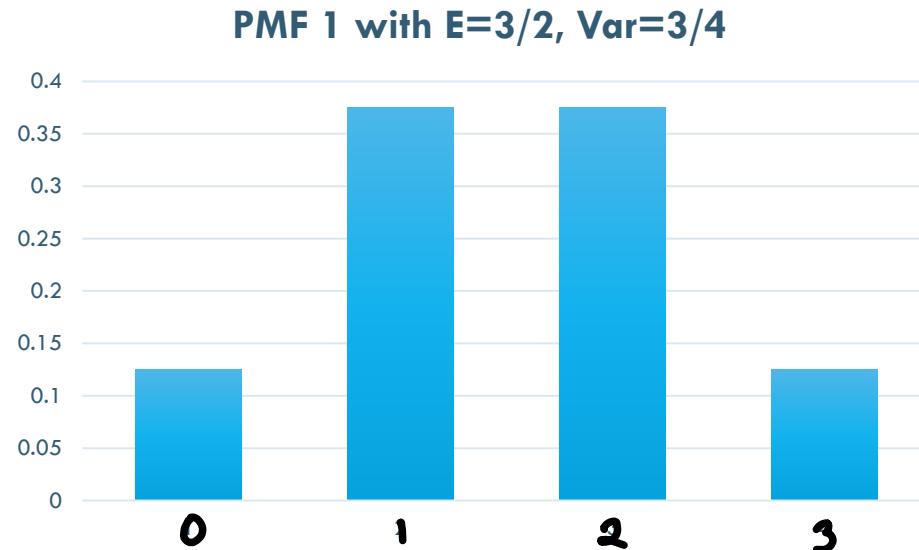
$$\text{Var}(X) = \sum_{i=1}^n p(1 - p) = np(1 - p).$$

# Expectation and Variance aren't everything

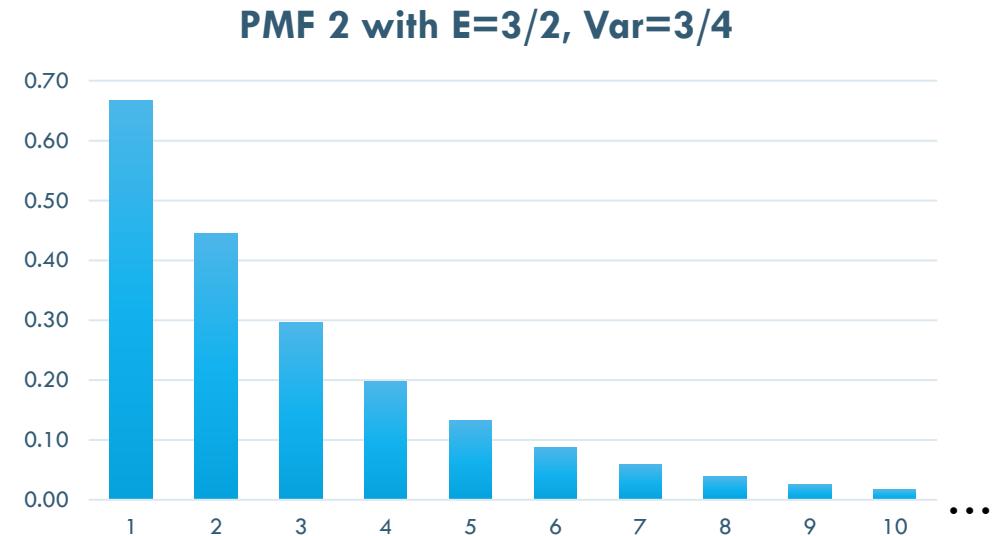
Alright, so expectation and variance is everything right?

No!

Flip a fair coin 3 times indep. Count heads.



Flip a biased coin (prob heads=2/3) until heads. Count flips.

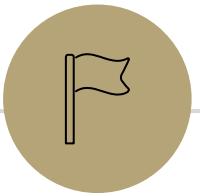


A PMF or CDF *\*does\** fully describe a random variable.

# Proof of Calculation Trick

$$\begin{aligned}\mathbb{E}[(X - \mathbb{E}[X])^2] &= \mathbb{E}[X^2 - 2X\mathbb{E}[X] + (\mathbb{E}[X])^2] \text{ expanding the square} \\ &= \mathbb{E}[X^2] - \mathbb{E}[2X\mathbb{E}[X]] + \mathbb{E}[(\mathbb{E}[X])^2] \text{ linearity of expectation.} \\ &= \mathbb{E}[X^2] - 2\mathbb{E}[X]\mathbb{E}[X] + \mathbb{E}[(\mathbb{E}[X])^2] \text{ linearity of expectation.} \\ &= \mathbb{E}[X^2] - 2\mathbb{E}[X]\mathbb{E}[X] + (\mathbb{E}[X])^2 \text{ expectation of a constant is the constant} \\ &= \mathbb{E}[X^2] - 2(\mathbb{E}[X])^2 + (\mathbb{E}[X])^2 \\ &= \mathbb{E}[X^2] - (\mathbb{E}[X])^2\end{aligned}$$

$$\text{So } \text{Var}(X) = \mathbb{E}[X^2] - (\mathbb{E}[X])^2.$$



## Extra Practice

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# More Practice

Suppose you flip a coin until you see a heads for the first time.

Let  $X$  be the number of trials (including the heads)

What is the pmf of  $X$ ?

The cdf of  $X$ ?

$\mathbb{E}[X]$ ?

# More Practice

Suppose you flip a coin until you see a heads for the first time.

Let  $X$  be the number of trials (including the heads)

What is the pmf of  $X$ ?  $f_X(x) = 1/2^x$  for  $x \in \mathbb{Z}^+$ , 0 otherwise

The cdf of  $X$ ?  $F_X(x) = 1 - 1/2^{\lfloor x \rfloor}$  for  $x \geq 0$ , 0 for  $x < 0$ .

$$\mathbb{E}[X]?$$
  $\sum_{i=1}^{\infty} \frac{i}{2^i} = 2$

# More Random Variable Practice

Roll a fair die  $n$  times. Let  $X$  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.

What is the expectation?

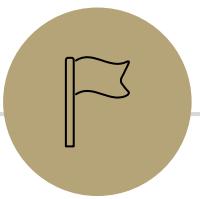
# More Random Variable Practice

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

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

Expectation formula is a mess. If you plug it into a calculator you'll get a nice, clean simplification:  $n/3$ .



**More Practice: Infinite  
sequential processes**

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# Infinite sequential process

In volleyball, sets are played first team to

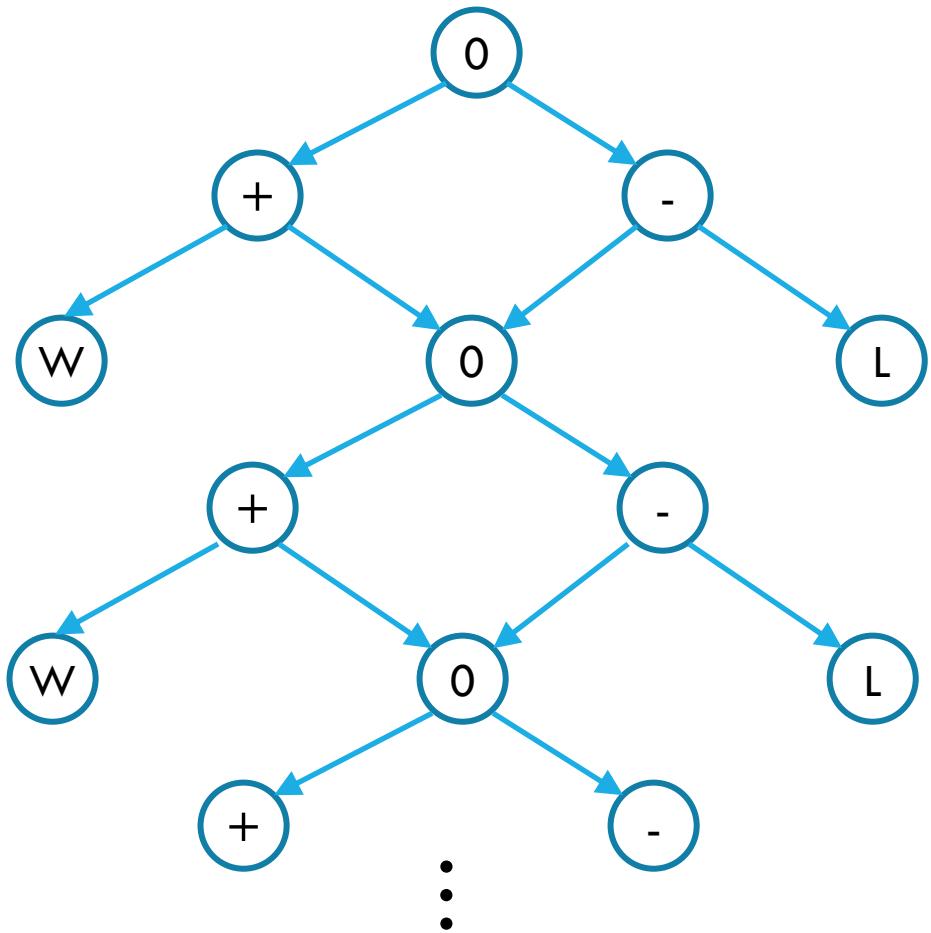
- Score 25 points
- Lead by at least 2

At the same time wins a set.

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

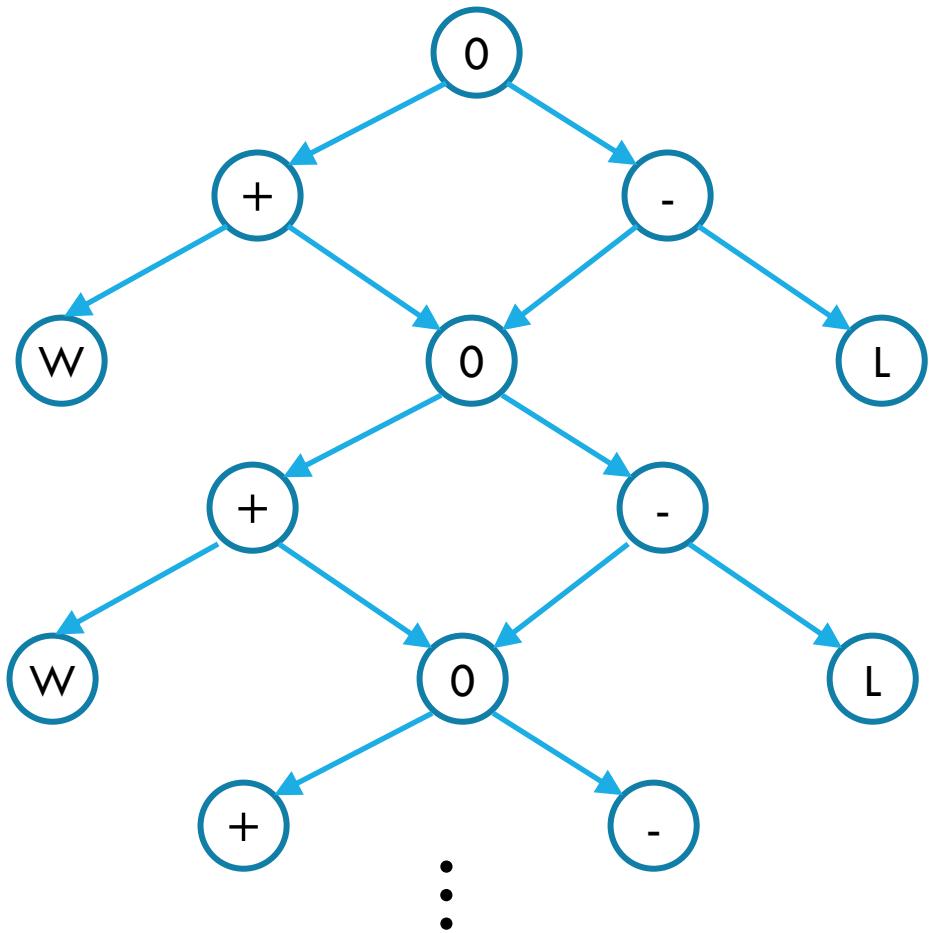
What is the probability your team wins the set?

# Sequential Process



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

# Sequential Process



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

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

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

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