

# Random Variables

CSE 312 Winter 26  
Lecture 9

# 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.

# The sum of two dice

## EVENTS

We could define

$E_2$  = "sum is 2"

$E_3$  = "sum is 3"

...

$E_{12}$  = "sum is 12"

And ask "which event occurs"?

## RANDOM VARIABLE

$X: \Omega \rightarrow \mathbb{R}$

$X$  is the sum of the two dice.

# More random variables

From one sample space, you can define many random variables.

Roll a fair red die and a fair blue die

Let  $D$  be the value of the red die minus the blue die  $D(4,2) = 2$

Let  $S$  be the sum of the values of the dice  $S(4,2) = 6$

Let  $M$  be the maximum of the values  $M(4,2) = 4$

...

# Notational Notes

We will always use capital letters for random variables.

It's common to use lower-case letters for the values they could take on.

**Formally** random variables are functions, so you'd think we'd write

$$X(H, H, T) = 2$$

But we nearly never do. We just write  $X = 2$

# Support ( $\Omega_X$ )

The “support” (aka “the range”) is the set of values  $X$  can actually take.

We called this the “image” in 311.

$D$  (difference of red and blue) has support  $\{-5, -4, -3, \dots, 4, 5\}$

$S$  (sum) has support  $\{2, 3, \dots, 12\}$

What is the support of  $M$  (max of the two dice)

# Probability Mass Function

Often, we're interested in the event  $\{\omega: X(\omega) = x\}$

Which is the event...that  $X = x$ .

We'll write  $\mathbb{P}(X = x)$  to describe the probability of that event

So  $\mathbb{P}(S = 2) = \frac{1}{36}$ ,  $\mathbb{P}(S = 7) = \frac{1}{6}$

The function that tells you  $\mathbb{P}(X = x)$  is the “probability mass function”

We'll often write  $p_X(x)$  for the pmf.

# Partition

A random variable partitions  $\Omega$ .

Let  $T$  be the number of twos in rolling a (fair) red and blue die.

$$p_T(0) = 25/36$$

$$p_T(1) = 10/36$$

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

	D2=1	D2=2	D2=3	D2=4	D2=5	D2=6
D1=1	(1,1)	(1,2)	(1,3)	(1,4)	(1,5)	(1,6)
D1=2	(2,1)	(2,2)	(2,3)	(2,4)	(2,5)	(2,6)
D1=3	(3,1)	(3,2)	(3,3)	(3,4)	(3,5)	(3,6)
D1=4	(4,1)	(4,2)	(4,3)	(4,4)	(4,5)	(4,6)
D1=5	(5,1)	(5,2)	(5,3)	(5,4)	(5,5)	(5,6)
D1=6	(6,1)	(6,2)	(6,3)	(6,4)	(6,5)	(6,6)



# Describing a Random Variable (differently)

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

There are 20 balls, numbered  $1, 2, \dots, 20$  in an urn.

You'll draw out a size-three subset. (i.e. without replacement)

$\Omega = \{\text{size three subsets of } \{1, \dots, 20\}\}$ ,  $\mathbb{P}()$  is uniform measure.

Let  $X$  be the largest value among the three balls.

If outcome is  $\{4, 2, 10\}$  then  $X = 10$ .

Write down the PMF of  $X$ ; Write down the CDF of  $X$ .

# Try It Yourself

There are 20 balls, numbered 1,2,...,20 in an urn.

You'll draw out a size-three subset. (i.e. without replacement)

Let  $X$  be the largest value among the three balls.

$$p_X(x) = \begin{cases} \binom{x-1}{2} / \binom{20}{3} & \text{if } x \in \mathbb{N}, 3 \leq x \leq 20 \\ 0 & \text{otherwise} \end{cases}$$

Good check: if you sum up  $p_X(x)$  do you get 1?

Good check: is  $p_X(x) \geq 0$  for all  $x$ ? Is it defined for all  $x$ ?

# 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.

# 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.

$\Omega = \{TT, TH, HT, HH\}$ ,  $\mathbb{P}()$  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  $\mathbb{E}[Y]$ ?

# Try it yourself

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

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

$\mathbb{E}[X]$  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.

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

# 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?

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

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

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

$$X + Y$$

$$X^2$$

$$2X + 3$$

Etc.

**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  $g(\mathbb{E}[X])$ .