### Alex Psomas: Lecture 16.

#### Random Variables

- Regrade requests open.
- Quiz due tomorrow.
- Quiz coming out today.
- Non-technical office hours tomorrow 1-3pm.
- Anonymous questionnaire tonight or tomorrow.

### Random Variables

- 1. Random Variables.
- 2. Distributions.
- 3. Combining random variables.
- 4. Expectation

### Questions about outcomes ...

```
Experiment: roll two dice.
```

Sample Space: 
$$\{(1,1),(1,2),\ldots,(6,6)\}=\{1,\ldots,6\}^2$$

How many dots?

Experiment: flip 100 coins.

Sample Space:  $\{HHH\cdots H, THH\cdots H, \dots, TTT\cdots T\}$ 

How many heads in 100 coin tosses?

Experiment: choose a random student in cs70.

Sample Space:  $\{\textit{Peter}, \textit{Phoebe}, \ldots, \}$ 

What midterm score?

Experiment: hand back assignments to 3 students at random.

Sample Space: {123,132,213,231,312,321}

How many students get back their own assignment?

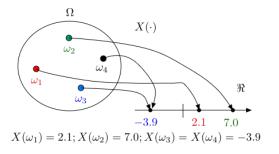
In each scenario, each outcome gives a number.

The number is a (known) function of the outcome.

#### Random Variables.

A **random variable**, X, for an experiment with sample space  $\Omega$  is a function  $X : \Omega \to \Re$ .

Thus,  $X(\cdot)$  assigns a real number  $X(\omega)$  to each  $\omega \in \Omega$ .



The function  $X(\cdot)$  is defined on the outcomes  $\Omega$ .

A random variable X is not random, not a variable!

What varies at random (from experiment to experiment)? The outcome!

# Example 1 of Random Variable

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Experiment: roll two dice. Sample Space: \{(1,1),(1,2),\dots,(6,6)\}=\{1,\dots,6\}^2 Random Variable X: number of pips. X(1,1)=2 X(1,2)=3, \vdots X(6,6)=12, X(a,b)=a+b,(a,b)\in\Omega.
```

## Example 2 of Random Variable

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Experiment: flip three coins

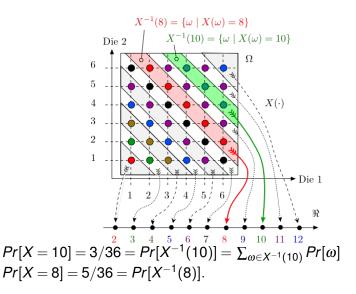
Sample Space: {HHH, THH, HTH, TTH, HHT, THT, HTT, TTT}

Winnings: if win 1 on heads, lose 1 on tails: X

X(HHH) = 3
X(THH) = 1
X(HTH) = 1
X(TTH) = -1
X(HTT) = -1
X(TTT) = -3
```

#### Number of dots in two dice.

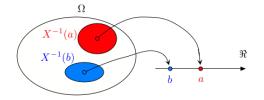
"What is the likelihood of seeing *n* dots?"



#### Distribution

The probability of X taking on a value a.

**Definition:** The **distribution** of a random variable X, is  $\{(a, Pr[X = a]) : a \in \mathcal{A}\}$ , where  $\mathcal{A}$  is the range of X.



$$Pr[X = a] := Pr[X^{-1}(a)] \text{ where } X^{-1}(a) := \{\omega \mid X(\omega) = a\}.$$

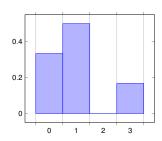
### Handing back assignments

Experiment: hand back assignments to 3 students at random.

Sample Space:  $\Omega = \{123, 132, 213, 231, 312, 321\}$ How many students get back their own assignment? Random Variable: values of  $X(\omega)$ :  $\{3, 1, 1, 0, 0, 1\}$ 

#### Distribution:

$$X = \begin{cases} 0, & \text{w.p. } 1/3 \\ 1, & \text{w.p. } 1/2 \\ 3, & \text{w.p. } 1/6 \end{cases}$$



# Flip three coins

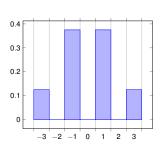
Experiment: flip three coins

Sample Space: {HHH, THH, HTH, TTH, HHT, THT, HTT, TTT}

Winnings: if win 1 on heads, lose 1 on tails. X Random Variable:  $\{3,1,1,-1,1,-1,-1,-3\}$ 

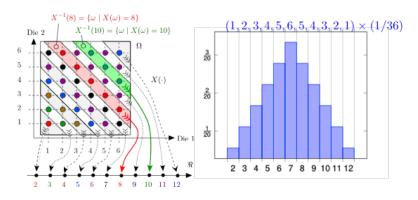
#### Distribution:

$$X = \begin{cases} -3, & \text{w. p. } 1/8 \\ -1, & \text{w. p. } 3/8 \\ 1, & \text{w. p. } 3/8 \\ 3 & \text{w. p. } 1/8 \end{cases}$$



#### Number of dots.

#### Experiment: roll two dice.



#### The Bernoulli distribution

Flip a coin, with heads probability p.

Random variable X: 1 is heads, 0 if not heads.

X has the Bernoulli distribution.

We will also call this an **indicator random variable**. It indicates whether the event happened.

#### Distribution:

$$X = \begin{cases} 1 & \text{w.p. } p \\ 0 & \text{w.p. } 1 - p \end{cases}$$

#### The binomial distribution.

Flip n coins with heads probability p.

Random variable: number of heads.

Binomial Distribution: Pr[X = i], for each i.

How many sample points in event "X = i"? i heads out of n coin flips  $\implies \binom{n}{i}$ 

Sample space:  $\Omega = \{HHH...HH, HHH...HT, ...\}$ What is the probability of  $\omega$  if  $\omega$  has i heads?

Probability of heads in any position is p.

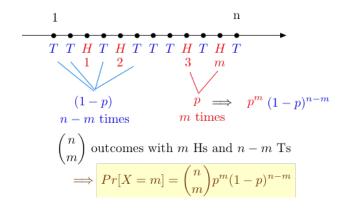
Probability of tails in any position is (1 - p).

So, we get  $Pr[\omega] = p^{i}(1-p)^{n-i}$ .

Probability of "X = i" is sum of  $Pr[\omega]$ ,  $\omega \in "X = i$ ".

$$Pr[X = i] = \binom{n}{i} p^i (1-p)^{n-i}, i = 0, 1, \dots, n : B(n, p)$$
 distribution

### The binomial distribution.



# Combining Random Variables.

Let *X* and *Y* be two RV on the same probability space.

That is,  $X : \Omega \to \Re$  assigns the value  $X(\omega)$  to  $\omega$ . Also,  $Y : \Omega \to \Re$  assigns the value  $Y(\omega)$  to  $\omega$ .

Then Z = X + Y is a random variable: It assigns the value

$$Z(\omega) = X(\omega) + Y(\omega)$$

to outcome  $\omega$ .

Experiment: Roll two dice. X = outcome of first die, Y = outcome of second die.

$$X(a,b) = a \text{ and } Y(a,b) = b \text{ for } (a,b) \in \Omega = \{1,\ldots,6\}^2.$$

Then Z = X + Y = sum of two dice is defined by

$$Z(a,b) = X(a,b) + Y(a,b) = a + b.$$

# Combining Random Variables

#### Other random variables:

- ►  $X^k : \Omega \to \Re$  is defined by  $X^k(\omega) = [X(\omega)]^k$ . In the dice example,  $X^3(a,b) = a^3$ .
- $(X-2)^2 + 4XY$  assigns the value  $(X(\omega)-2)^2 + 4X(\omega)Y(\omega)$  to  $\omega$ .
- ▶ g(X, Y, Z) assigned the value  $g(X(\omega), Y(\omega), Z(\omega))$  to  $\omega$ .

### Expectation.

How did people do on the midterm?

Distribution.

Summary of distribution?

Average!



### **Expectation - Intuition**

Flip a loaded coin with Pr[H] = p a large number N of times.

We expect heads to come up a fraction p of the times and tails a fraction 1 - p.

Say that you get 5 for every H and 3 for every T.

If there are  $N_H$  outcomes equal to H and  $N_T$  outcomes equal to T, you collect

$$5 \times N_H + 3 \times N_T$$
.

Your average gain per experiment is

$$\frac{5N_H+3N_T}{N}.$$

Since  $\frac{N_H}{N} \approx p = Pr[X = 5]$  and  $\frac{N_T}{N} \approx 1 - p = Pr[X = 3]$ , we find that the average gain per outcome is approximately equal to

$$5Pr[X = 5] + 3Pr[X = 3].$$

We use this frequentist interpretation as a definition.

## **Expectation - Definition**

**Definition:** The **expected value** of a random variable *X* is

$$E[X] = \sum_{a} a \times Pr[X = a].$$

a in the range of X.

The expected value is also called the mean.

According to our intuition, we expect that if we repeat an experiment a large number N of times and if  $X_1, \ldots, X_N$  are the successive values of the random variable, then

$$\frac{X_1+\cdots+X_N}{N}\approx E[X].$$

That is indeed the case, in the same way that the fraction of times that X = x approaches Pr[X = x].

This (nontrivial) result is called the Law of Large Numbers.

# **Expectation: A Useful Fact**

#### Theorem:

$$E[X] = \sum_{\alpha} X(\omega) \times Pr[\omega].$$

#### **Proof:**

$$E[X] = \sum_{a} a \times Pr[X = a]$$

$$= \sum_{a} a \times \sum_{\omega: X(\omega) = a} Pr[\omega]$$

$$= \sum_{a} \sum_{\omega: X(\omega) = a} a \times Pr[\omega]$$

$$= \sum_{a} \sum_{\omega: X(\omega) = a} X(\omega) Pr[\omega]$$

$$= \sum_{\omega: X(\omega) = a} X(\omega) Pr[\omega]$$

### An Example

Flip a fair coin three times.

$$\Omega = \{HHH, HHT, HTH, THH, HTT, THT, TTH, TTT\}.$$

X = number of H's:  $\{3, 2, 2, 2, 1, 1, 1, 0\}$ .

Thus,

$$\sum_{\omega} X(\omega) Pr[\omega] = \{3+2+2+2+1+1+1+0\} \times \frac{1}{8}.$$

Also,

$$\sum_{a} a \times Pr[X = a] = 3 \times \frac{1}{8} + 2 \times \frac{3}{8} + 1 \times \frac{3}{8} + 0 \times \frac{1}{8}.$$

# Expectation and Average.

There are *n* students in the class;

$$X(m)$$
 = score of student  $m$ , for  $m = 1, 2, ..., n$ .

"Average score" of the n students: add scores and divide by n:

Average = 
$$\frac{X(1) + X(1) + \cdots + X(n)}{n}.$$

Experiment: choose a student uniformly at random. Uniform sample space:  $\Omega = \{1, 2, \dots, n\}, Pr[\omega] = 1/n$ , for all  $\omega$ .

Random Variable: midterm score:  $X(\omega)$ .

Expectation:

$$E(X) = \sum_{\omega} X(\omega) Pr[\omega] = \sum_{\omega} X(\omega) \frac{1}{n}.$$

Hence,

Average 
$$= E(X)$$
.

Our intuition matches the math.

# Handing back assignments

We give back assignments randomly to three students. What is the expected number of students that get their own assignment back?

The expected number of **fixed points** in a random permutation.

Expected value of a random variable:

$$E[X] = \sum_{a} a \times Pr[X = a].$$

For 3 students (permutations of 3 elements):

$$Pr[X = 3] = 1/6, Pr[X = 1] = 3/6, Pr[X = 0] = 2/6.$$
  
$$E[X] = 3 \times \frac{1}{6} + 1 \times \frac{3}{6} + 0 \times \frac{2}{6} = 1.$$

#### Win or Lose.

Expected winnings for heads/tails games, with 3 flips? Every time it's H, I get 1,. Every time it's T, I lose 1.

$$E[X] = 3 \times \frac{1}{8} + 1 \times \frac{3}{8} - 1 \times \frac{3}{8} - 3 \times \frac{1}{8} = 0.$$

Can you ever win 0?

Apparently: expected value is not a common value, by any means.

### Expectation

Recall:  $X : \Omega \to \Re; Pr[X = a]; = Pr[X^{-1}(a)];$ 

**Definition:** The **expectation** of a random variable *X* is

$$E[X] = \sum_{a} a \times Pr[X = a].$$

#### Indicator:

Let A be an event. The random variable X defined by

$$X(\omega) = \begin{cases} 1, & \text{if } \omega \in A \\ 0, & \text{if } \omega \notin A \end{cases}$$

is called the indicator of the event A.

Note that Pr[X = 1] = Pr[A] and Pr[X = 0] = 1 - Pr[A]. Hence.

$$E[X] = 1 \times Pr[X = 1] + 0 \times Pr[X = 0] = Pr[A].$$

The random variable X is sometimes written as

$$1\{\omega \in A\}$$
 or  $1_A(\omega)$ .

### Linearity of Expectation

**Theorem:** 

$$E[X] = \sum_{\omega} X(\omega) \times Pr[\omega].$$

**Theorem:** Expectation is linear

$$E[a_1X_1 + \cdots + a_nX_n] = a_1E[X_1] + \cdots + a_nE[X_n].$$

#### **Proof:**

$$E[a_1X_1 + \dots + a_nX_n]$$

$$= \sum_{\omega} (a_1X_1 + \dots + a_nX_n)(\omega)Pr[\omega]$$

$$= \sum_{\omega} (a_1X_1(\omega) + \dots + a_nX_n(\omega))Pr[\omega]$$

$$= a_1\sum_{\omega} X_1(\omega)Pr[\omega] + \dots + a_n\sum_{\omega} X_n(\omega)Pr[\omega]$$

$$= a_1E[X_1] + \dots + a_nE[X_n].$$

## Using Linearity - 1: Dots on dice

Roll a die n times.

 $X_m$  = number of dots on roll m.

$$X = X_1 + \cdots + X_n$$
 = total number of dots in  $n$  rolls.

$$E[X] = E[X_1 + \cdots + X_n]$$
  
=  $E[X_1] + \cdots + E[X_n]$ , by linearity  
=  $nE[X_1]$ , because the  $X_m$  have the same distribution

Now,

$$E[X_1] = 1 \times \frac{1}{6} + \dots + 6 \times \frac{1}{6} = \frac{6 \times 7}{2} \times \frac{1}{6} = \frac{7}{2}.$$

Hence,

$$E[X] = \frac{7n}{2}$$
.

# Using Linearity - 2: Fixed point.

Hand out assignments at random to *n* students.

X = number of students that get their own assignment back.

$$X = X_1 + \cdots + X_n$$
 where

 $X_m = 1$ {student m gets his/her own assignment back}.

One has

$$E[X] = E[X_1 + \cdots + X_n]$$
  
 $= E[X_1] + \cdots + E[X_n]$ , by linearity  
 $= nE[X_1]$ , because all the  $X_m$  have the same distribution  
 $= nPr[X_1 = 1]$ , because  $X_1$  is an indicator  
 $= n(1/n)$ , because student 1 is equally likely  
to get any one of the  $n$  assignments  
 $= 1$ .

Note that linearity holds even though the  $X_m$  are not independent (whatever that means).

## Using Linearity - 3: Binomial Distribution.

Flip n coins with heads probability p. X - number of heads

Binomial Distibution: Pr[X = i], for each i.

$$Pr[X=i] = \binom{n}{i} p^{i} (1-p)^{n-i}.$$

$$E[X] = \sum_{i} i \times Pr[X = i] = \sum_{i} i \times \binom{n}{i} p^{i} (1 - p)^{n - i}.$$

No no no no no. NO ... Or... a better approach: Let

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

$$E[X_i] = 1 \times Pr["heads"] + 0 \times Pr["tails"] = p.$$

Moreover  $X = X_1 + \cdots X_n$  and

$$E[X] = E[X_1] + E[X_2] + \cdots + E[X_n] = n \times E[X_i] = np.$$

## Today's gig: St. Petersburg paradox

I offer the following game:

We start with a pot of 2 dollars.

Flip a fair coin. If it's tails, you take the pot. If it's heads, I double the pot.

So, if the sequence is *HHT*, you make 8 dollars.

How much would you we willing to pay?

# Today's gig: St. Petersburg paradox

Well, how much money should you expect to make?

Let *X* be the random variable indicating how much money you make for each outcome:

$$X = 2$$
 with probability  $\frac{1}{2}$ 

$$X = 4$$
 with probability  $\frac{1}{4}$ 

$$X = 8$$
 with probability  $\frac{1}{8}$ 

$$E[X] = 2\frac{1}{2} + 4\frac{1}{4} + 8\frac{1}{8} + \dots$$
$$= 1 + 1 + 1 + \dots = \infty$$

So, if you were rational you would be willing to pay anything! Is there a trick here?

# Today's gig: St. Petersburg paradox

#### What if I didn't have infinite money?

Banker	Bankroll	Expected value of lottery
Friendly game	\$100	\$7.56
Millionaire	\$1,000,000	\$20.91
Billionaire	\$1,000,000,000	\$30.86
Bill Gates (2015)	\$79,200,000,000 <sup>[5]</sup>	\$37.15
U.S. GDP (2007)	\$13.8 trillion <sup>[6]</sup>	\$44.57
World GDP (2007)	\$54.3 trillion <sup>[6]</sup>	\$46.54
Googolaire	\$10 <sup>100</sup>	\$333.14

Random Variables

- ▶ A random variable X is a function  $X : \Omega \to \Re$ .
- ►  $Pr[X = a] := Pr[X^{-1}(a)] = Pr[\{\omega \mid X(\omega) = a\}].$
- ▶  $Pr[X \in A] := Pr[X^{-1}(A)].$
- ▶ The distribution of X is the list of possible values and their probability:  $\{(a, Pr[X = a]), a \in \mathcal{A}\}.$
- g(X, Y, Z) assigns the value .....
- $\blacktriangleright E[X] := \sum_a aPr[X = a].$
- Expectation is Linear.
- ► *B*(*n*,*p*).