

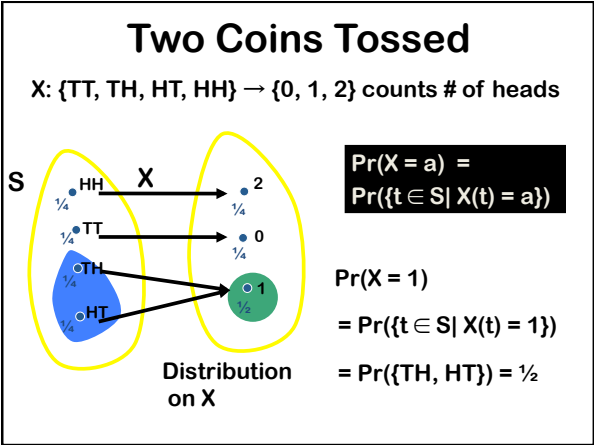
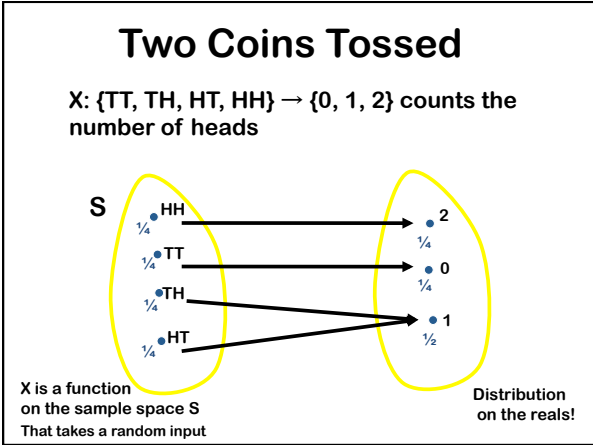
# From 15-251

## Great Theoretical Ideas in Computer Science

CMU (with modifications)

# Random variables

**Discrete random variable:** take a finite or countable number of values



### Definition: Expectation

The expectation, or expected value of a random variable X is written as  $E[X]$ , and is

$$E[X] = \sum_{t \in S} \Pr(t) X(t) = \sum_k k \Pr[X = k]$$

X is a function on the sample space S

X has a prob. distribution on its values

(assuming X takes values in the naturals)

### A Quick Calculation...

What if I flip a coin 3 times? What is the expected number of heads?

$$E[X] = (1/8) \times 0 + (3/8) \times 1 + (3/8) \times 2 + (1/8) \times 3 = 1.5$$

But  $\Pr[X = 1.5] = 0$

Moral: don't always expect the expected.  $\Pr[X = E[X]]$  may be 0!

### Functions of a R.V.

Suppose that  $X$  is a random variable defined on probability space  $(S, p(\cdot))$   
Then  $Y = g(X)$  is also a random variable on the same probability space (if defined everywhere).

E.g.,  $Y = X^2$

$$Z = 3^X + 2$$

### Expectation of a function of a r.v.

$$E(X) = \sum_k k Pr(X = k) = \sum_{\omega \in S} X(\omega) Pr(\omega)$$

$$E(g(X)) = \sum_k g(k) Pr(X = k) \quad k \text{ in range of } X$$

### Operations on R.V.s

You can define any random variable you want on a probability space.

Given a collection of random variables, you can sum them, take differences, or do most other math operations...

$$\text{E.g., } (X + Y)(t) = X(t) + Y(t)$$

$$(X * Y)(t) = X(t) * Y(t)$$

$$(X^Y)(t) = X(t)^{Y(t)}$$

Random variables and expectations allow us to give elegant solutions to problems that seem really really messy...

If I randomly put 100 letters into 100 addressed envelopes, on average how many letters will end up in their correct envelopes?

On average, in class of size  $m$ , how many pairs of people will have the same birthday?



Pretty messy with direct counting...

The new tool is called  
"Linearity of Expectation"

## Linearity of Expectation



If  $Z = X+Y$ , then  
 $E[Z] = E[X] + E[Y]$

Without any conditions on X and Y

## By Induction

$$E[X_1 + X_2 + \dots + X_n] = E[X_1] + E[X_2] + \dots + E[X_n]$$



The expectation of the sum  
 =  
 The sum of the expectations

Let's see why  
 Linearity of Expectation  
 is so useful...

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 into 100 addressed  
 envelopes, on average how  
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Hmm...

$$\sum_k k \Pr(\text{exactly } k \text{ letters end up in correct envelopes})$$

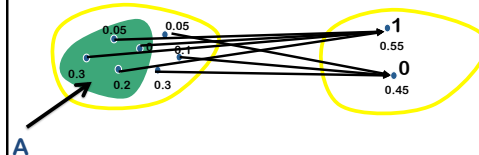
$$= \sum_k k (\dots \text{aargh!!} \dots)$$



## Indicator Random Variables

For any event  $A$ , can define the  
 "indicator random variable" for event  $A$ :

$$X_A(t) = \begin{cases} 1 & \text{if } t \in A \\ 0 & \text{if } t \notin A \end{cases} \quad E[X_A] = 1 \times \Pr(X_A = 1) = \Pr(A)$$



### Use Linearity of Expectation

Let  $A_i$  be the event the  $i^{\text{th}}$  letter ends up in its correct envelope

Let  $X_i$  be the "indicator" R.V. for  $A_i$

$$X_i = \begin{cases} 1 & \text{if } A_i \text{ occurs} \\ 0 & \text{otherwise} \end{cases}$$

Let  $Z = X_1 + \dots + X_{100}$

We are asking for  $E[Z]$

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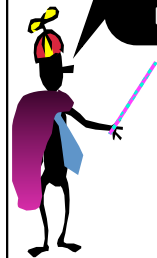


$$E[X_i] = \Pr(A_i) = 1/100$$

$$\text{So } E[Z] = 1$$

So, in expectation, 1 letter will be in the same correct envelope

Pretty neat: it doesn't depend on how many letters!



### Use Linearity of Expectation

General approach:

View thing you care about as expected value of some RV

Write this RV as sum of simpler RVs

Solve for their expectations and add them up!



### Ex. #2

We flip  $n$  coins of bias  $p$ . What is the expected number of heads?

We could do this by summing

$$\sum_k k \Pr(X = k) = \sum_k k \binom{n}{k} p^k (1-p)^{n-k}$$

But now we know a better way!



### Linearity of Expectation!

Let  $X$  = number of heads when  $n$  independent coins of bias  $p$  are flipped

Break  $X$  into  $n$  simpler RVs:

$$X_i = \begin{cases} 1 & \text{if the } i^{\text{th}} \text{ coin is heads} \\ 0 & \text{if the } i^{\text{th}} \text{ coin is tails} \end{cases}$$

$$E[X] = E[\sum_i X_i] = \sum_i E[X_i] = \sum_i p = np$$

On average, in class of size  $m$ , how many pairs of people will have the same birthday?

$\sum_k k \Pr(\text{exactly } k \text{ pairs})$   
 $= \sum_k k (\dots \text{aargh!!!!} \dots)$

Use linearity of expectation

Suppose we have  $m$  people each with a uniformly chosen birthday from 1 to 365

$X$  = number of pairs of people with the same birthday

$E[X] = ?$

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Use  $m(m-1)/2$  indicator variables, one for each pair of people

$X_{jk} = 1$  if person  $j$  and person  $k$  have the same birthday; else 0

$E[X_{jk}] = (1/365) \cdot 1 + (1 - 1/365) \cdot 0 = 1/365$

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$E[X] = E[\sum_{1 \leq j < k \leq m} X_{jk}]$   
 $= \sum_{1 \leq j < k \leq m} E[X_{jk}]$   
 $= \binom{m}{2} \frac{1}{365}$

### Type Checking

$P(B)$      **B** must be an **event**

$E(X)$      **X** must be a **R.V.**

**cannot do  $P(\text{R.V.})$  or  $E(\text{event})$**