

# CS 112: Computer System Modeling Fundamentals

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Lecture 4

# Quiz #1!

# Reminders & Announcements

- Homework 1 is due **in class on Tuesday**
- There was initially a small typo in problem 6, so get the revised version if you haven't already

# Today

- Random variables
- Expectation
- Examples of discrete random variables
  - Bernoulli random variables
  - Binomial random variables
  - Geometric random variables
  - Poisson random variables
  - Power law distributions (didn't get to this.. next time!)

# Review of Probability Definitions

- An **experiment** is an activity that results in exactly one of several possible outcomes.
- The set of all possible outcomes of an experiment is called the **sample space**, typically denoted  $\Omega$ .
- An **event** is a collection of outcomes – that is, a **subset** of the sample space.
- The **atomic events** are the basic, fundamental elements of the sample space. They are always **mutually exclusive** and **exhaustive**.

# Random Variables

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Example: Suppose we roll two six-sided dice.

- Sample space  $\Omega = \{11, 12, 13, \dots, 66\}$
- Random variables: the sum of the two rolls, the number of sixes rolled, the product of the two rolls, the value of the first roll, the number of even rolls, etc.

# Random Variables

Example: Suppose we have four disks, each of which fails with probability  $p$ .

- Sample space  $\Omega = \{0000, 0001, 0010, \dots, 1111\}$
- Random variables: the number of disks that failed, a Boolean value indicating if one of the first two disks failed, a Boolean value indicating if at least one disk failed, etc.

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In this class, we will use upper case letters for random variables and lower case letters for their values.

# Probability Mass Functions

The **probability mass function (PMF)** of a random variable  $X$  is denoted  $p_X$ . For any value  $k$  that  $X$  can take on,  $p_X(k)$  is the probability of the event  $\{X=k\}$ , so

$$p_X(k) = P(X=k) = P(\{X=k\})$$

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Example: Suppose we flip a fair coin twice. Let  $X$  be the number of heads. What is  $p_X(0)$ ? What is  $p_X(1)$ ?

# Probability Mass Function

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- Example: Suppose we flip a fair coin twice. Let  $X$  be the number of heads. What is  $P(X \geq 1)$ ?
- The same reasoning can be applied to show that

$$\sum_{\text{all values } k} p_X(k) = 1$$

# Bernoulli Random Variables

- Consider a biased coin that lands on heads with probability  $p$ . Let  $X$  be a random variable that takes on the value 1 if the coin is heads and 0 if the coin is tails. Such a variable is called a **Bernoulli random variable**.

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- Examples: A server can be online or not. A new email that you receive can be spam or not. A web user can click or not click on an ad that is displayed.
- The probability mass function is defined as follows:

$$p_X(1) = p$$

$$p_X(0) = 1-p$$

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- How can we calculate  $p_X(k)$  for a particular value of  $k$ ?

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# Expectation

Suppose you take a trip to Las Vegas and play a slot machine  $n$  times. Each time you play, you receive a random payout from a set of possible payouts  $\{r_1, r_2, \dots, r_m\}$ , where for each  $i$ , the probability of receiving  $r_i$  is  $p_i$ .

Let  $X$  be a random variable denoting your average payout (that is, total payout divided by  $n$ ).

How much of a payout should you “expect” to get on average each time you play?

# Expectation

- The **expected value** of a random variable  $X$  with probability mass function  $p_X$  is

$$E[X] = \sum_x xp_X(x)$$

- We sometimes call this the **expectation** or **mean** of  $X$ .

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- This is simply the value we'd expect to get for  $X$  “on average” if we repeated the same experiment many times.

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- Let  $X$  and  $Y$  be random variables. Define a new random variable  $Z = X+Y$ . Then  $E[Z] = E[X]+E[Y]$ .
- What is the expected value of a binomial random variable with parameters  $p$  and  $n$ ?

# Poisson Random Variables

- Consider a book containing  $n$  words. Suppose that the probability of any particular word being misspelled is  $p$ . Let  $X$  be the number of words that are misspelled.

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- We could model  $X$  as a binomial random variable, but when  $n$  is large and  $p$  is small, it is often more convenient to model it as a **Poisson random variable**.

# Poisson Random Variables

- A **Poisson random variable** with parameter  $\lambda$  takes on values in  $\{0, 1, 2, \dots\}$  and has the PMF

$$p_X(k) = \frac{e^{-\lambda} \lambda^k}{k!}$$

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- The expected value is simply  $\lambda$
- If we want our random variable to represent the number of misspelled words in a book with  $n$  words, each misspelled with probability  $p$ , how should we set  $\lambda$ ?

# Poisson vs. Binomial

Black dots:

Poisson,  $\lambda=5$

Red curve:

Bin,  $n=10, p=0.5$

Blue curve:

Bin,  $n=20, p=0.25$

Green curve:

Bin,  $n=1000,$

$p=0.005$

(All have mean 5)

