

CSE400 – Fundamentals of Probability in Computing

Lecture 6: Discrete RVs, Expectation and Problem Solving

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Outline

- Previous Lecture Recap: Random Variables (RVs)
 - Independent Events / Examples
- Definition and Example
- Types of Discrete Random Variables
 - Bernoulli RV
 - Binomial RV
 - Geometric RV
 - Poisson RV
- Expectation of RVs
 - Definition and Example
 - Expectation of a Function of RV
 - Linear Operation with Expectation
- Moments and Central Moments of RVs
 - Variance, Skewness and Kurtosis

- The Cumulative Density Function (CDF)
 - Definition, Properties and Examples
- The Probability Density Function (PDF)
 - Definition, Properties and Examples

Random Variables

Motivation and Concept

A random variable (X) on a sample space (Ω) is a function

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

that assigns to each sample point ($\omega \in \Omega$) a real number ($X(\omega)$).

Until further notice, we will restrict our attention to random variables that are **discrete**, i.e., they take values in a range that is **finite or countably infinite**. This means even though we define (X) to map (Ω) to (\mathbb{R}), the actual set of values

$$\{X(\omega) : \omega \in \Omega\}$$

that (X) takes is a **discrete subset of (\mathbb{R})**.

Sample space

Sample point (s)

Sample points mapped by the discrete random variable ($X(s)$) into numbers on the real line.

Sample space of all permutations

Random Variables

Motivation and Concept

The distribution of a random variable can be visualized as a **bar diagram**:

$$\Pr[X = a]$$

The x-axis represents the values that a random variable can take on.

The height of the bar at a value (a) is the probability ($\Pr[X = a]$).

Each of these probabilities can be computed by looking at the probability of the corresponding event in the sample space.

Guide to Selecting a Probability Distribution

Random Variables

Discrete variable

- Countable support
- Probability mass function
- Probabilities assigned to single values
- Each possible value has strictly positive probability

Bernoulli Random Variable

A Bernoulli random variable (X) takes values in

$$\{0, 1\}$$

$$\Pr[X = 1] = p$$

$$\Pr[X = 0] = 1 - p$$

Binomial Random Variable

A binomial random variable counts the number of successes in (n) independent Bernoulli trials.

$$X \sim \text{Binomial}(n, p)$$

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

Geometric Random Variable

A geometric random variable models the number of trials until the first success.

$$\Pr[X = k] = (1 - p)^{k-1} p$$

Poisson Random Variable

A Poisson random variable models the number of occurrences in a fixed interval.

$$\Pr[X = k] = \frac{\lambda^k e^{-\lambda}}{k!}$$

Expectation of Random Variables

Definition

The expectation of a discrete random variable (X) is defined as

$$E[X] = \sum_x x \Pr[X = x]$$

Expectation

Example

(Example shown on slide – no additional steps provided)

Expectation of a Function of a Random Variable

Let ($Y = g(X)$)

$$E[Y] = \sum_x g(x) \Pr[X = x]$$

Linear Operation with Expectation

$$E[aX + b] = aE[X] + b$$

Moments and Central Moments of RVs

- Mean
- Variance
- Skewness
- Kurtosis

Variance

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

Cumulative Density Function (CDF)

Definition

$$F_X(x) = \Pr[X \leq x]$$

Properties of CDF

- Non-decreasing
- Right-continuous
- $\lim_{x \rightarrow -\infty} F_X(x) = 0$
- $\lim_{x \rightarrow \infty} F_X(x) = 1$

Probability Density Function (PDF)

Definition

$$f_X(x) = \Pr[X = x]$$

Properties of PDF

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

$$f_X(x) \geq 0$$

End of Lecture 6

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