

CSE400: Fundamentals of Probability in Computing

Lecture 6: Discrete RVs, Expectation and Problem Solving

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1 Introduction to Random Variables (RVs)

[cite_{start}] A random variable X on a sample space Ω is a function $X : \Omega \rightarrow \mathbb{R}$ that assigns a real number $X(\omega)$ to each sample point $\omega \in \Omega$ [cite: 48].

- **Discrete RVs:** Take values in a range that is finite or countably infinite [cite: 49]. [cite_{start}]
- **Distribution Visualization:** Can be represented as a bar diagram where the x-axis shows values and the height shows $Pr[X = a]$ [cite: 104, 105, 106].

2 Probability Mass Function (PMF)

[cite_{start}] A random variable is discrete if it can take on at most a countable number of values [cite: 204].

Definition: The function $P_X(x_k) = P(X = x_k)$ for $k = 1, 2, 3, \dots$ is the PMF of X [cite: 221, 222]. [cite_{start}]

Property: The sum of all probabilities must equal 1: $\sum_{k=1}^{\infty} P_X(x_k) = 1$ [cite: 233].

3 Bayes' Theorem Recap

$$Pr(B_i|A) = \frac{Pr(A|B_i)Pr(B_i)}{\sum_{i=1}^n Pr(A|B_i)Pr(B_i)}$$

- **A priori probability:** $Pr(B_i)$ - formed from presupposed models [cite: 291]. [cite_{start}]
- **Posteriori probability:** $Pr(B_i|A)$ - calculated after observing event A [cite: 292].

4 Independence

- **Two Events:** A and B are independent if $Pr(A|B) = Pr(A)$ and $Pr(B|A) = Pr(B)$, implying $Pr(A, B) = Pr(A)Pr(B)$ [cite: 448]. [cite_{start}]
- **Three Events:** A, B, and C are mutually independent if [cite: 460]:

- $Pr(A, B) = Pr(A)Pr(B)$, $Pr(B, C) = Pr(B)Pr(C)$, $Pr(A, C) = Pr(A)Pr(C)$ [cite: 462, 463]. [cite_{start}]
- $Pr(A, B, C) = Pr(A)Pr(B)Pr(C)$ [cite: 463].

5 Types of Discrete Random Variables

5.1 Bernoulli Random Variable

[cite_{start}] Models an experiment with only two outcomes : Success(1) or Failure(0)[cite : 535, 544].

PMF: $P_X(1) = p$ and $P_X(0) = 1 - p$, where $p \in (0, 1)$ [cite: 556, 557]. [cite_{start}]

Examples: Single coin toss, email spam classification[cite: 572, 574].

5.2 Binomial Random Variable $B(n, p)$

[cite_{start}] Models the number of successes in n independent trials, each with success probability[cite : 597, 599].

PMF: $p(i) = \binom{n}{i} p^i (1-p)^{n-i}$ for $i = 0, 1, \dots, n$ [cite: 612]. [cite_{start}]

Examples: Correct answers on a test, number of defective items in a sample[cite: 626, 627].

5.3 Geometric Random Variable

[cite_{start}] Models the number of independent trials required until the first success occurs[cite : 644, 645].

PMF: $P_X(X = n) = (1 - p)^{n-1} \times p$ for $n = 1, 2, \dots$ [cite: 659]. [cite_{start}]

Examples: Tosses until first head, attempts until first sale[cite: 669, 670].

5.4 Poisson Random Variable

[cite_{start}] A discrete RV with parameter $\lambda > 0$ [cite: 695].

- **PMF:** $p(i) = P\{X = i\} = e^{-\lambda} \frac{\lambda^i}{i!}$ for $i = 0, 1, 2, \dots$ [cite: 702, 703]. [cite_{start}]
- **Approximation:** Used for $B(n, p)$ when n is large and p is small such that $n \times p$ is moderate[cite: 725, 726]. [cite_{start}]
- **Examples:** Misprints on a page, customers entering a post office daily[cite: 740, 744].