

Random Variables

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Usage Rights

These materials were created for educational purposes and to be completed as part of material covered in the CE305: Probability, Statistics, and Data Analysis undergraduate course. Students will have a similar notebook to turn in as part of the weekly assignment, which will be graded for completion and not correctness towards the participation category of their final grade.

Learning Objectives

By the end of this notebook, students will be able to:

- **Part I:**
 - Define a random variable for an experiment;
 - Differentiate between discrete and continuous random variables;
 - Compute and interpret descriptive statistics of random variables;
 - Differentiate between probability mass functions and probability density functions;
 - Construct the appropriate probability distribution for a discrete and a continuous random variable;
 - Use the definition of cumulative distribution functions (CDFs) to compute event probabilities;
- **Part II**
 - Understand linear functions and combinations of random variables;
 - Interpret the impact of sample size on the variability of statistical measures.
 - Determine whether a continuous random variable is appropriate to describe data trends for a discrete random variable.

Recap Part I (prev. lecture)

Random Variables (RVs)

- A random variable assigns a numerical value to each outcome in a sample space. Two types of random variables are: discrete and continuous.
- A DRV can take distinct (separate), countable values, whereas a CRV can take on any value over the range it is defined.

Probability Distributions

- A probability distribution is the list of probabilities associated with the value(s) a RV takes. It is referred to as a PMF for DRV, and as a PDF for CRV.
- The CDF of a random variable, denoted by, gives the probability that a RV takes on all values less than or equal to a value of interest in its range.
- The median of a probability distribution is the value of the RV, x_m , at which the data is divided in half
- The p-th percentile of a distribution is the value of the RV, x_p , below which p% of the distribution is found.
- The definition of the CDF can be used to determine the RV median, x_m , and p-th percentiles of interest,

$$x_p$$

Detailed discussion and explanations are found in the Textbook, Ch. 2.4 Random Variables. (Navidi Textbook, 6th Ed)

Part II

Linear Functions of Random Variables

Operations with a Constant

If X is a random variable, and a and b are constants, then:

- $\mu_{X+b} = \mu_X + b$, $\sigma_{X+b}^2 = \sigma_X^2$
- $\mu_{aX} = a\mu_X$, $\sigma_{aX}^2 = a^2\sigma_X^2$, $\sigma_{aX} = |a|\sigma_X$
- $\mu_{aX+b} = a\mu_X + b$, $\sigma_{aX+b}^2 = a^2\sigma_X^2$, $\sigma_{aX+b} = |a|\sigma_X$

Linear Combinations of Random Variables

If X_1, \dots, X_n are random variables and c_1, \dots, c_n are constants, we can get the linear combination of X_1, \dots, X_n : $c_1X_1 + \dots + c_nX_n$, then:

- $\mu_{c_1X_1 + \dots + c_nX_n} = c_1\mu_{X_1} + \dots + c_n\mu_{X_n}$

Linear Combinations of Independent Random Variables

If X_1, \dots, X_n are independent random variables, c_1, \dots, c_n are constants, and S_1, \dots, S_n are sets, then:

- $P(X_1 \in S_1 \text{ and } \dots \text{ and } X_n \in S_n) = P(X_1 \in S_1) \dots P(X_n \in S_n)$

- $\sigma_{c_1X_1 + \dots + c_nX_n}^2 = c_1^2\sigma_{X_1}^2 + \dots + c_n^2\sigma_{X_n}^2$

Independence and Simple Random Samples

Independent and identically distributed (i.i.d.) sample: X_1, \dots, X_n are independent random variables and all have the same distribution.

If X_1, \dots, X_n is a simple random sample from a population with mean μ and variance σ^2 , then:

- The sample mean \bar{X} will be a random variable with $\mu_{\bar{X}} = \mu$ and $\sigma_{\bar{X}}^2 = \frac{\sigma^2}{n}$
- The standard deviation of \bar{X} is: $\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$

More details can be found in Ch. 2.5 Linear Functions of Random Variables (Navidi Textbook, 6th Ed)

Example 1

A GPS app tracks a cyclist's average travel time per mile during a ride and provides the following data.

- Mean travel time: $\mu_X = 4$ minutes.
- Standard deviation: $\sigma_X = 0.5$ minutes.
- Question 1: Define the experiment.
- Question 2: Define the random variable in words.

Now, the app wants to convert this to speed in miles per hour. The speed in mph is calculated as: $Y = \frac{60}{X}$. Since this transformation is not linear, the app also considers a linear approximation near $X = 4$, using the function:

- $Y_{\text{approx}} = -3X + 20$
- Question 3: Find the expected speed $\mu_{Y_{\text{approx}}}$.
- Question 4: Find the standard deviation $\sigma_{Y_{\text{approx}}}$ of the approximate speed.

Example 2

At a popular coffee shop, the average waiting time for a customer is known to be $\mu = 6$ minutes, with a standard deviation of $\sigma = 3$ minutes. A manager wants to estimate the average wait time on a given day by randomly sampling the wait times of several customers.

Let X_1, \dots, X_n be the set of RVs for the waiting time of a random sample of n customers. Define \bar{X} as the sample mean wait time.

- Question 1: Define the experiment
- Question 2: Define, in words, the random variable X_n
- Question 3: If the manager samples $n=36$ customers, what is the mean and standard deviation of \bar{X} ?
- Question 4: Explain why the sample mean becomes more stable (less variable) as n increases.
- Question 5: What sample size n is needed to reduce the standard deviation of \bar{X} to 0.2 minute?