Understanding Distributions in Statistics

- 1. What is a Distribution?
- A distribution in statistics describes how the values of a variable are spread out or distributed.
- It tells us what values are possible and how often (or how likely) each value occurs.
- Distributions help us understand the behavior of data, perform statistical analysis, and make predictions.
- 2. Two Definitions or Views of Distribution:
- a) Mathematical View (Theoretical Distribution):
- A distribution can be defined as a mathematical function (like a PDF or PMF).
- It maps each possible value (x) to a probability (discrete) or density (continuous).
- Examples include Normal distribution, Binomial distribution, Poisson distribution, etc.
- b) Data-Based View (Empirical Distribution):
- Based on actual observed data.
- Constructed by calculating the frequency or relative frequency of values in the dataset.
- Represented using a histogram, bar chart, or empirical cumulative distribution function (ECDF).
- 3. Why Do We Fit a Theoretical Distribution to Data?
- Real-world data is just a sample. It may contain noise, variability, and imperfections.
- Fitting a known distribution (like Normal) helps generalize our understanding to the entire population.
- Theoretical distributions allow us to compute probabilities, make predictions, run simulations, and apply statistical tests.
- Example: If data looks bell-shaped, we may fit a Normal distribution using its formula with estimated mean and standard deviation.

4. Understanding Histogram vs PDF (Probability Density Function): Histogram: - Built from data by dividing the range into bins and counting values in each bin. - Shows relative frequency (proportion of data in each bin). - Gives approximate probability by: relative frequency = count in bin / total count. PDF (for continuous distributions): - A smooth curve that defines the density of data. - PDF value at a point is not a probability. To get probability, we must integrate the PDF over a range. - For example, P(55 < X < 65) =Area under PDF curve from 55 to 65. 5. Key Difference Between Relative Frequency and Density: - Relative Frequency: Computed directly from data. It is the proportion of observations in a bin. - Density: Found in a histogram when using 'density=True'. It is relative frequency divided by bin width. - To get probability from density: Probability approximately Density x Bin Width - In a fitted PDF, probability = Area under the curve = Integral of f(x) from a to b. 6. Summary Table: | Feature | Empirical Distribution | Theoretical Distribution |------| Mathematical model (PDF/PMF) Source Observed data

| Relative frequency or count | Density (continuous) or probability (discrete) |

| Generalized, population-level

| Integrate (continuous) or lookup (discrete) |

| Y-axis shows

| Nature

| Use of probability | Count or proportion

Data-specific, finite

- 7. Clarifying: Relative Frequency vs Density
- Relative Frequency: Proportion of data in a bin. It is an estimate of probability for that range.
- Density: Not a probability. It shows how concentrated the data is. Found in histograms when using density=True.

The Relationship:

Density = Relative Frequency / Bin Width

- -> Relative Frequency = Density * Bin Width
- In a regular histogram (density=False), bar height = relative frequency.
- In a density histogram (density=True), bar height = density. You need to multiply it by bin width to estimate probability.

Final Summary:

- Relative frequency is a direct estimate of probability from data.
- Density (PDF) is the rate of accumulation of probability per unit on the X-axis.
- To get true probability from PDF, integrate over a range.

8. Visual Explanation:

- The following chart shows how a histogram (empirical distribution) compares to a fitted Normal distribution (PDF).
- The shaded orange area between 55 and 65 represents the probability under the fitted curve for that range.

