Training Day-112 Report:

Z-Score:

What is Z-Score?

The **Z-Score** (also called the **Standard Score**) measures how many standard deviations a data point is from the mean of a distribution. It allows comparison of data points from different distributions by standardizing them.

A Z-score helps to determine whether a value is typical or unusual in a given dataset.

Formula for Z-Score

 $Z=X-\mu\sigma Z = \frac{X - \mu\sigma Z}{\sin a}$

Where:

- ZZ = Z-score
- XX = Individual data point
- μ \mu = Mean of the distribution
- $\sigma \setminus sigma = Standard deviation of the distribution$

Interpretation of Z-Score

- 1. Positive Z-Score (Z>0Z>0):
 - o Indicates that the data point is above the mean.
- 2. Negative Z-Score (Z<0Z<0):
 - o Indicates that the data point is below the mean.
- 3. **Z-Score of 0:**
 - o Indicates that the data point is exactly at the mean.
- 4. Magnitude of Z-Score:
 - o The further the Z-score is from 0, the more unusual the data point is.

Example of Z-Score Calculation

Example 1: Exam Scores

Suppose exam scores are normally distributed with a mean (μ \mu) of 75 and a standard deviation (σ \sigma) of 10. A student scores 85 on the exam.

To calculate the Z-score:

$$Z=85-7510=1010=1Z = \frac{85-75}{10} = \frac{10}{10} = 1$$

The Z-score is 1, meaning the student scored 1 standard deviation above the mean.

Applications of Z-Score

1. Comparing Data Points:

- o Standardizes values from different distributions for comparison.
- o Example: Comparing test scores from two tests with different scales.

2. Outlier Detection:

 Data points with Z-scores greater than 3 or less than -3 are often considered outliers.

3. Probability Calculations:

 Z-scores are used with standard normal tables to calculate probabilities for normal distributions.

4. Standardizing Data in Machine Learning:

 Z-scores are used to normalize features to have a mean of 0 and standard deviation of 1.

Z-Score and the Standard Normal Table

The **Standard Normal Table** (Z-Table) provides the cumulative probability for a given Z-score.

 Example: A Z-score of 1.25 corresponds to a cumulative probability of approximately 0.8944.

This means that 89.44% of the data lies below this Z-score.

Examples with Probability

Example 1: Finding Probabilities

A dataset of weights is normally distributed with a mean of 70 kg and a standard deviation of 5 kg. What is the probability that a randomly chosen person weighs less than 75 kg?

1. Calculate the Z-score:

$$Z=75-705=1Z = \frac{75 - 70}{5} = 1$$

- 2. Use the Z-Table:
 - o A Z-score of 1 corresponds to a cumulative probability of **0.8413**.
 - o Therefore, there is an **84.13%** chance that a person weighs less than 75 kg.

Example 2: Finding Intervals

What is the probability of a person weighing between 65 kg and 75 kg?

- 1. Calculate the Z-scores:
 - o For 65 kg: $Z=65-705=-1Z = \frac{65 70}{5} = -1$.
 - o For 75 kg: $Z=75-705=1Z = \frac{75 70}{5} = 1$.

2. Use the Z-Table:

- o Cumulative probability for Z=-1Z=-1: **0.1587**.
- o Cumulative probability for Z=1Z=1: **0.8413**.
- 3. Subtract probabilities:

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 $_{\odot}$ Therefore, there is a **68.26%** chance that a person weighs between 65 kg and 75 kg.