

Question 9 1 Pts Find The Best Predicted Value Of Y Corresponding To The Given Value Of X. Six Pairs

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Understanding how to predict the value of Y based on a given X is a fundamental aspect of data analysis and statistical modeling. When working with six pairs of data points, the goal is often to determine the most accurate predicted value of Y for a specified X, utilizing techniques such as linear regression or other predictive models. This article provides a comprehensive guide to understanding, calculating, and interpreting the best predicted Y value corresponding to a given X, especially in the context of six data pairs.

Introduction to Predictive Modeling with Data Pairs

Predictive modeling involves using known data points to estimate unknown values. Given six pairs of data points, each consisting of an X and a Y value, the aim is to develop a model that captures the relationship between these variables. Once the model is established, it can be used to predict Y for any given X within or outside the range of observed data.

Understanding the Six Data Pairs

What Are Data Pairs?

Data pairs are ordered pairs (X, Y) that represent observed values. For example:

- (X_1, Y_1)
- (X_2, Y_2)
- (X_3, Y_3)
- (X_4, Y_4)

- (X_5, Y_5)
- (X_6, Y_6)

These pairs form the basis for analyzing the relationship between X and Y.

The Significance of Six Data Pairs

While a larger dataset often provides more reliable models, six pairs are sufficient for simple linear regression analysis, particularly when the relationship appears linear. This small sample size requires careful analysis to avoid overfitting or erroneous predictions.

Methods for Finding the Best Predicted Y Value

There are several approaches to predict Y given X, but the most common is linear regression. Other methods include polynomial regression, exponential models, or non-parametric approaches, depending on the data pattern.

Linear Regression Overview

Linear regression models the relationship between X and Y as a straight line:

$$Y = a + bX + \epsilon$$

where:

- a is the intercept,
- b is the slope,
- ϵ is the error term.

The goal is to find the line that minimizes the sum of squared residuals (differences between observed and predicted Y values).

Calculating the Regression Line

To determine the best-fit line:

1. Compute the mean values of X and Y:

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

$$\bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i$$

2. Calculate the slope (b):

$$b = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (X_i - \bar{X})^2}$$

3. Calculate the intercept (a):

$$a = \bar{Y} - b \bar{X}$$

Once a and b are determined, the regression line is established.

Predicting Y for a Given X

With the regression line $(Y = a + bX)$, predicting the Y value for a specific X (say, X_{new}) is straightforward:

$$Y_{\text{predicted}} = a + b X_{\text{new}}$$

This predicted Y is considered the best estimate based on the data and the model assumptions.

Handling Six Data Pairs: Step-by-Step Example

Suppose the six pairs are:

X	Y
1	2
2	3
3	5
4	4
5	6
6	8

And you are asked to predict Y when $X = 7$.

Step 1: Calculate Means

$$\bar{X} = \frac{1 + 2 + 3 + 4 + 5 + 6}{6} = \frac{21}{6} = 3.5$$

$$\bar{Y} = \frac{2 + 3 + 5 + 4 + 6 + 8}{6} = \frac{28}{6} \approx 4.67$$

Step 2: Calculate the Slope (b)

Compute numerator:

$$\sum (X_i - \bar{X})(Y_i - \bar{Y}) = (1 - 3.5)(2 - 4.67) + (2 - 3.5)(3 - 4.67) + \dots + (6 - 3.5)(8 - 4.67)$$

Calculations:

$$\begin{aligned} - (1 - 3.5)(2 - 4.67) &= (-2.5)(-2.67) \approx 6.68 \\ - (2 - 3.5)(3 - 4.67) &= (-1.5)(-1.67) \approx 2.50 \\ - (3 - 3.5)(5 - 4.67) &= (-0.5)(0.33) \approx -0.17 \\ - (4 - 3.5)(4 - 4.67) &= (0.5)(-0.67) \approx -0.33 \\ - (5 - 3.5)(6 - 4.67) &= (1.5)(1.33) \approx 1.99 \\ - (6 - 3.5)(8 - 4.67) &= (2.5)(3.33) \approx 8.33 \end{aligned}$$

Sum:

$$6.68 + 2.50 - 0.17 - 0.33 + 1.99 + 8.33 = 19.00$$

Calculate denominator:

$$\sum (X_i - \bar{X})^2 = (-2.5)^2 + (-1.5)^2 + (-0.5)^2 + 0.5^2 + 1.5^2 + 2.5^2 = 6.25 + 2.25 + 0.25 + 0.25 + 2.25 + 6.25 = 17.5$$

Calculate slope:

$$b = \frac{19.00}{17.5} \approx 1.086$$

Step 3: Calculate Intercept (a)

$$\begin{aligned} a &= \bar{Y} - b \bar{X} \approx 4.67 - (1.086)(3.5) \approx 4.67 - 3.80 = \\ &0.87 \end{aligned}$$

Step 4: Predict Y at X = 7

$$\begin{aligned} Y_{\text{predicted}} &= a + b \times 7 \approx 0.87 + 1.086 \times 7 \approx 0.87 + \\ &7.602 = 8.472 \end{aligned}$$

Thus, the best predicted Y value when X = 7 is approximately 8.47.

Considerations and Limitations

Model Assumptions

Linear regression assumes a linear relationship between X and Y, homoscedasticity (constant variance of errors), independence of errors, and normally distributed residuals. Violating these assumptions can lead to inaccurate predictions.

Sample Size Constraints

With only six data pairs, the model may not capture complex relationships or outliers. Small datasets are more sensitive to anomalies, which can skew the model.

Extrapolation Risks

Predicting Y for X values outside the observed range (extrapolation) can be unreliable because the relationship might not hold beyond the data scope.

Advanced Techniques for Better Predictions

While linear regression is often sufficient for simple datasets, other methods can improve accuracy:

- **Polynomial Regression:** Captures non-linear relationships by including polynomial terms.
- **Logarithmic or Exponential Models:** Appropriate when data follows exponential growth or decay patterns.
- **Non-parametric Methods:** Techniques like k-nearest neighbors (k-NN) or kernel regression that do not assume a specific functional form.

These approaches can be employed when data analysis indicates non-linearity or complex relationships.

Practical Tips for Accurate Predictions

- **Plot the Data:**