

# What is Linear Regression?

- Regression allows to model mathematically the relationship between two or more variables (specifically linear relationship with the help of algebra)
- **Dependent Variable (DV)**
- **Independent Variable (IV)**

## Question 1:

Suppose you are an owner of a restaurant and interested to develop a model that will allow you to make a prediction about what amount of tip to expect for any given bill amount?

❖ collected data for six meals

# Data for Meals

Meal (#)	Tip Amount (in Rs.)
1.	7
2.	19
3.	13
4.	10
5.	16
6.	7

7.

?

**Interesting Fact:** You have only tip amount data?

- How to predict tip amount?
- What is DV and IV variable?

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Finding the value of dependent variable

Independent variables

Mathematical model:  $Y = f(X)$

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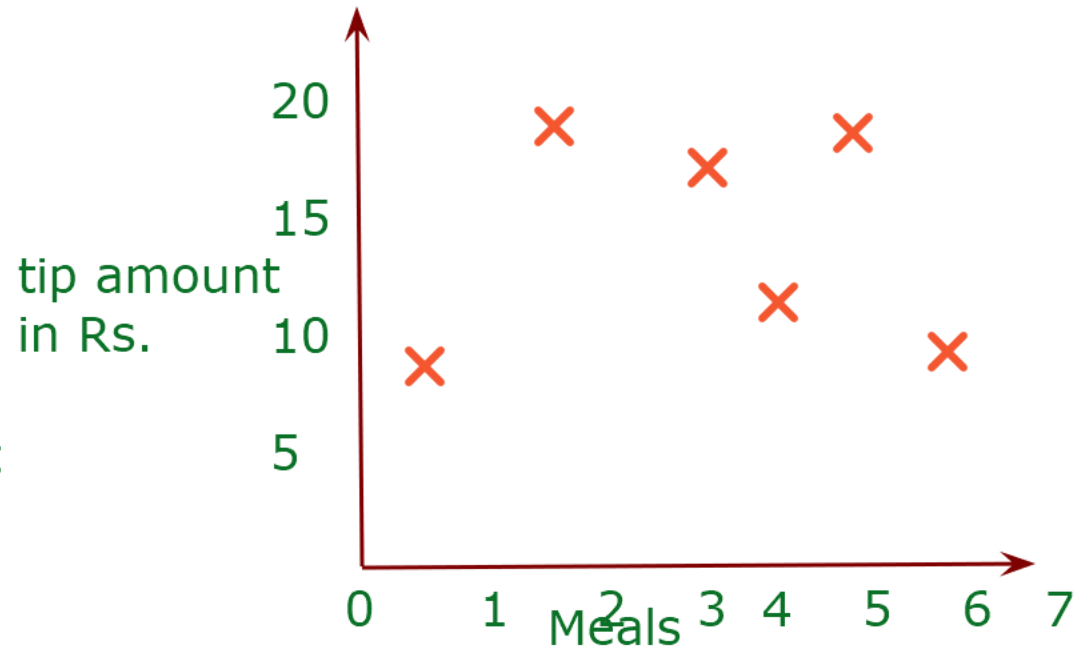
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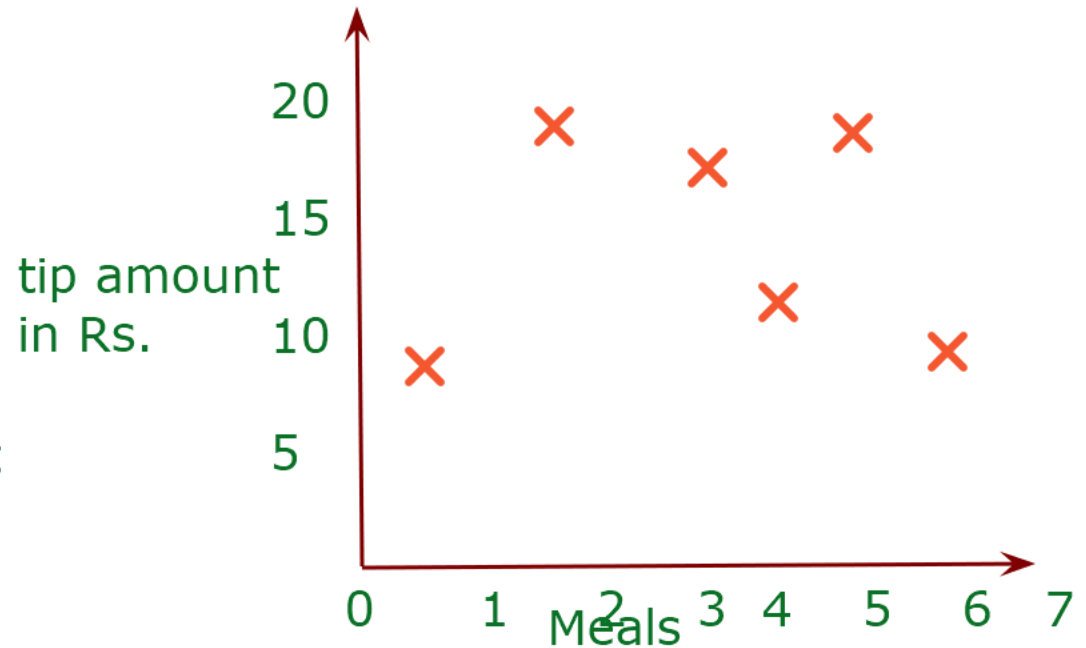
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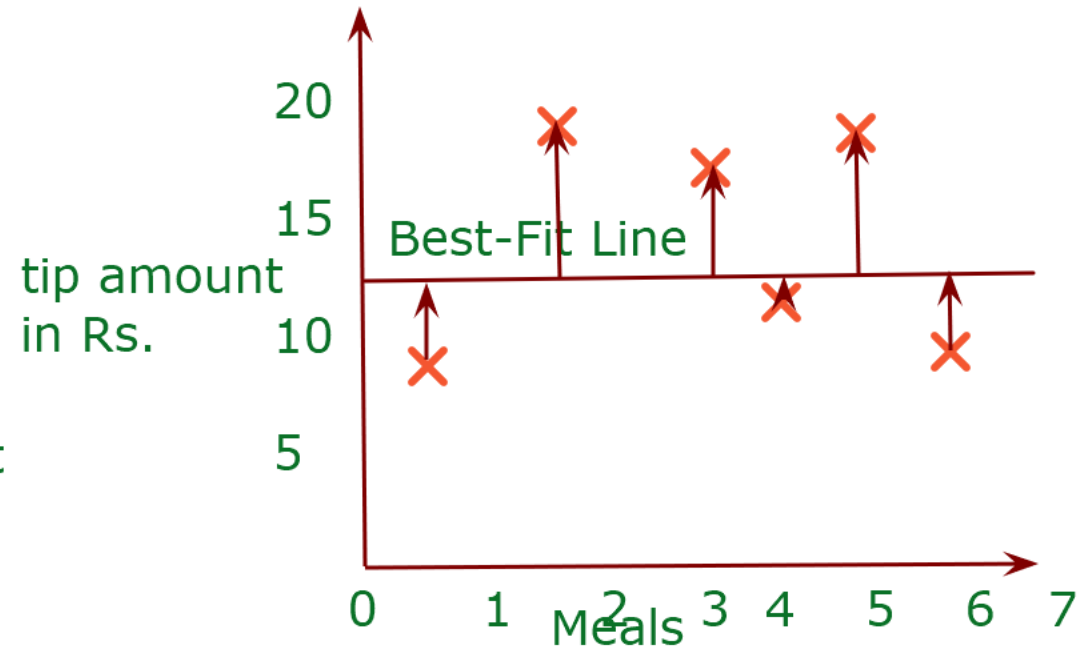
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Mean = 12.16: 12

Errors



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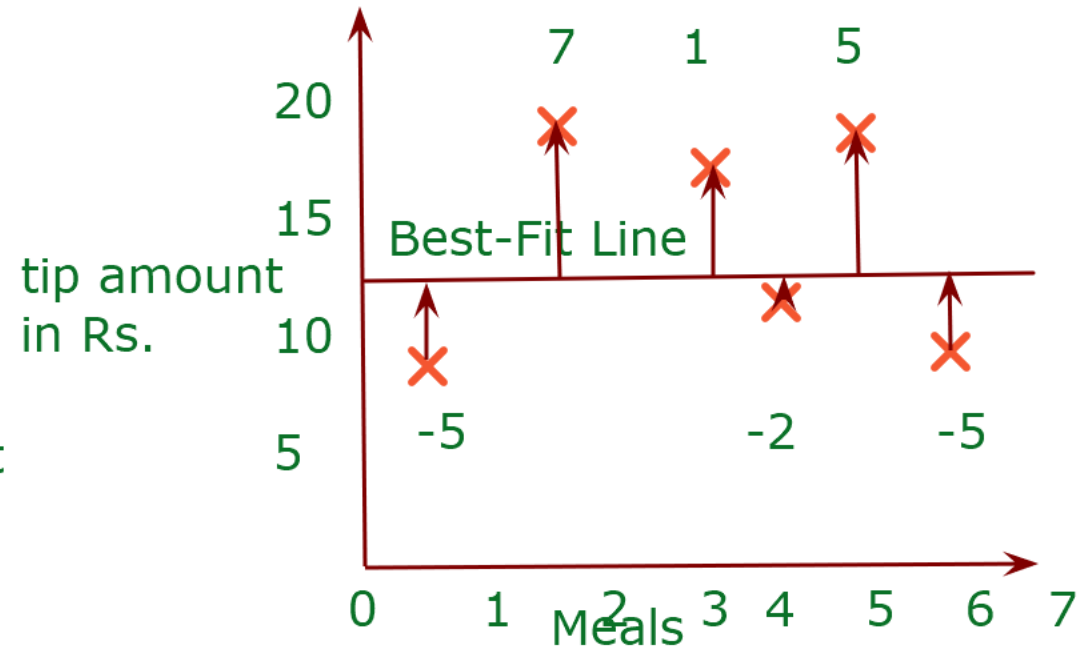
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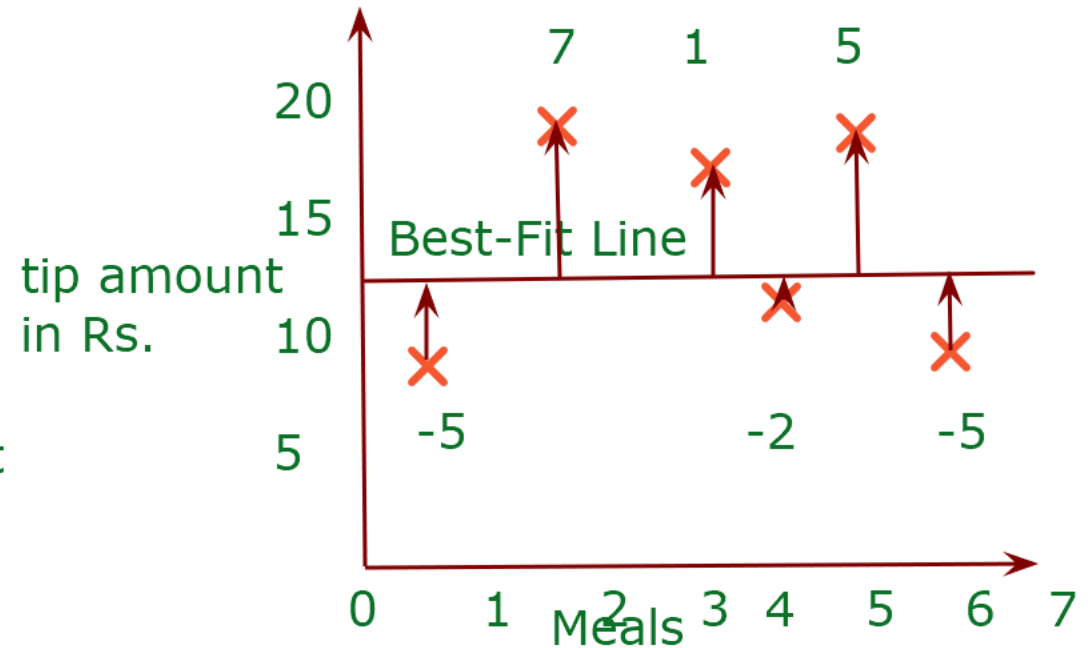
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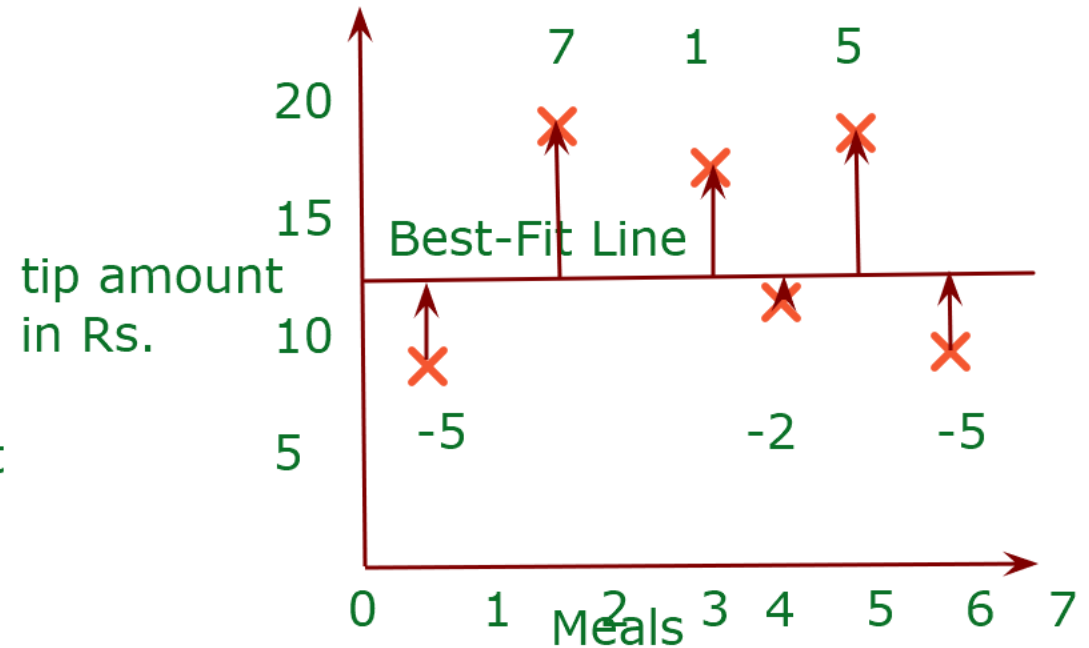
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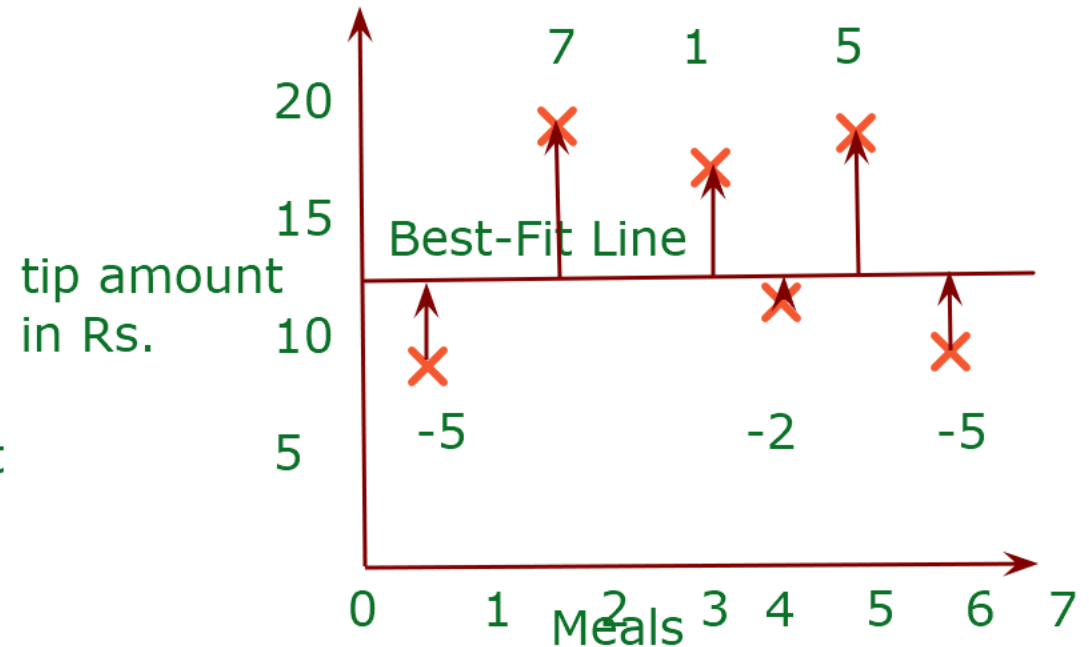
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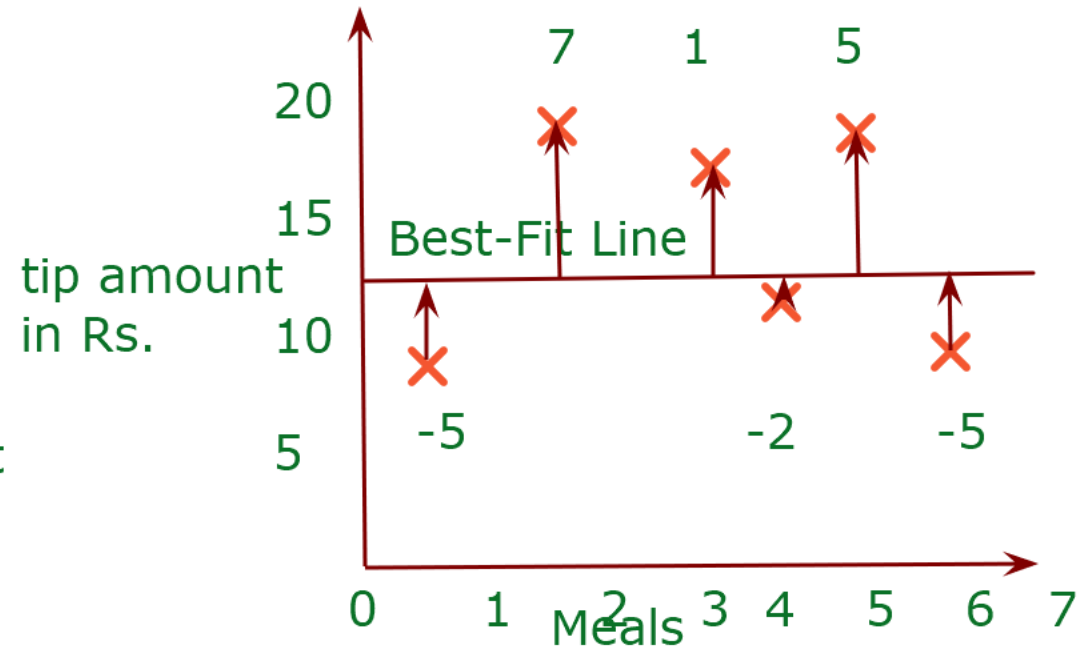
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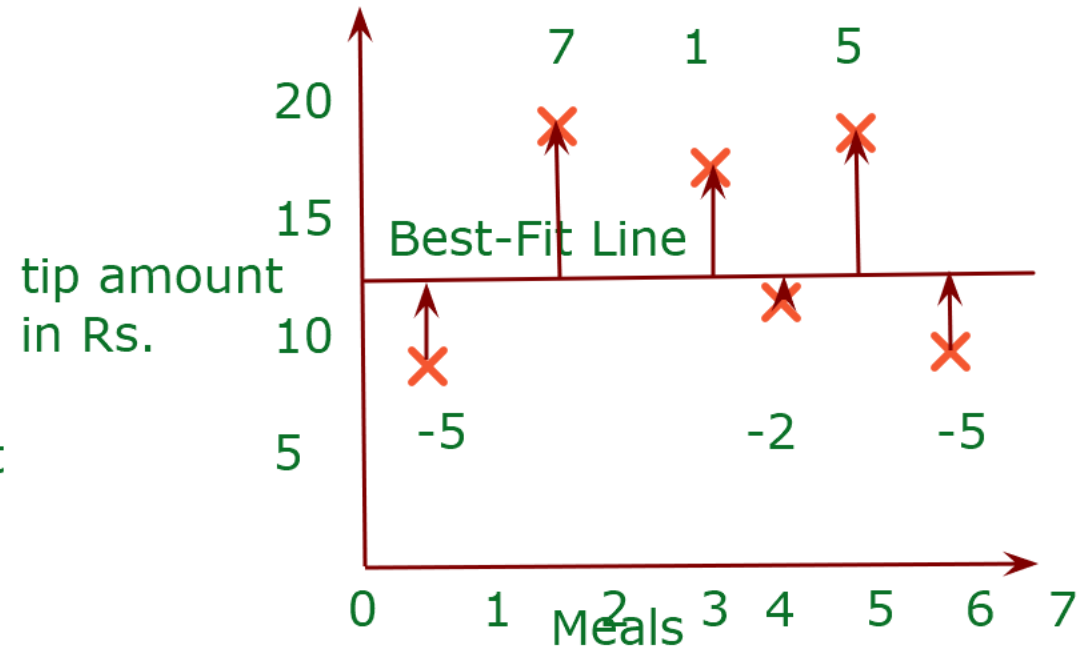
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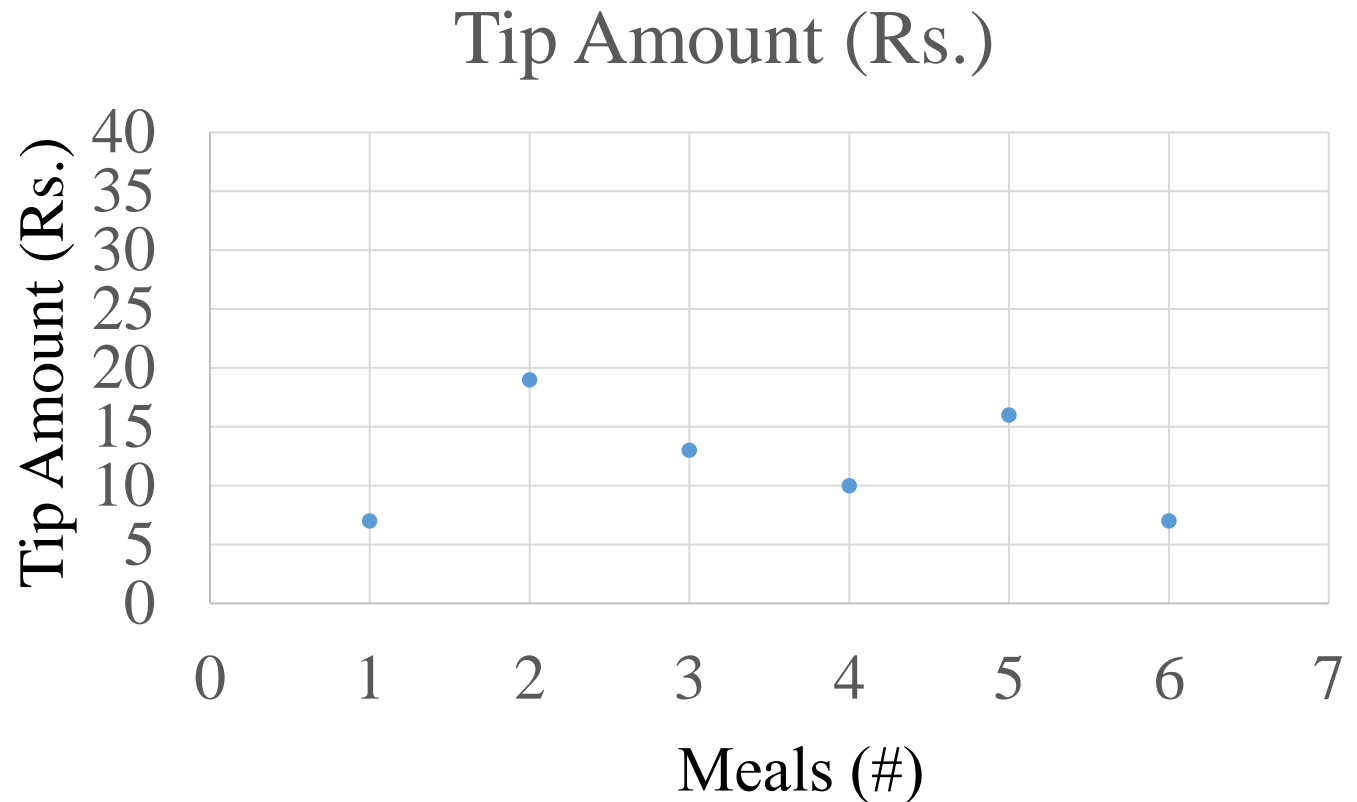
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**Scatter Plot:** Visualize the data to observe the pattern



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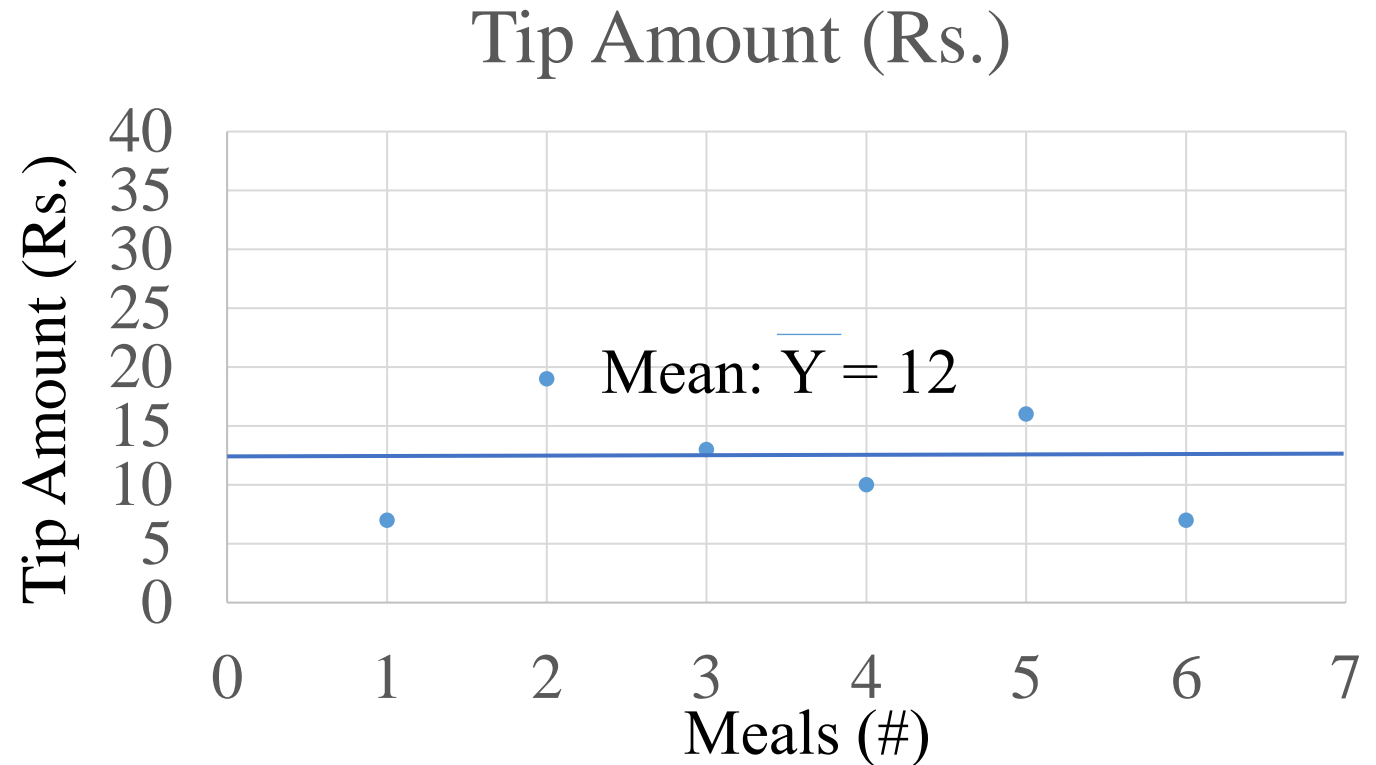
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**Best Predictor?**

**Can we use Mean of the Tip Amount?**



- Mean is the best estimate for predicting the tip amount when no other information is available with us, the variability in the tip amount can only be explained by the tips themselves



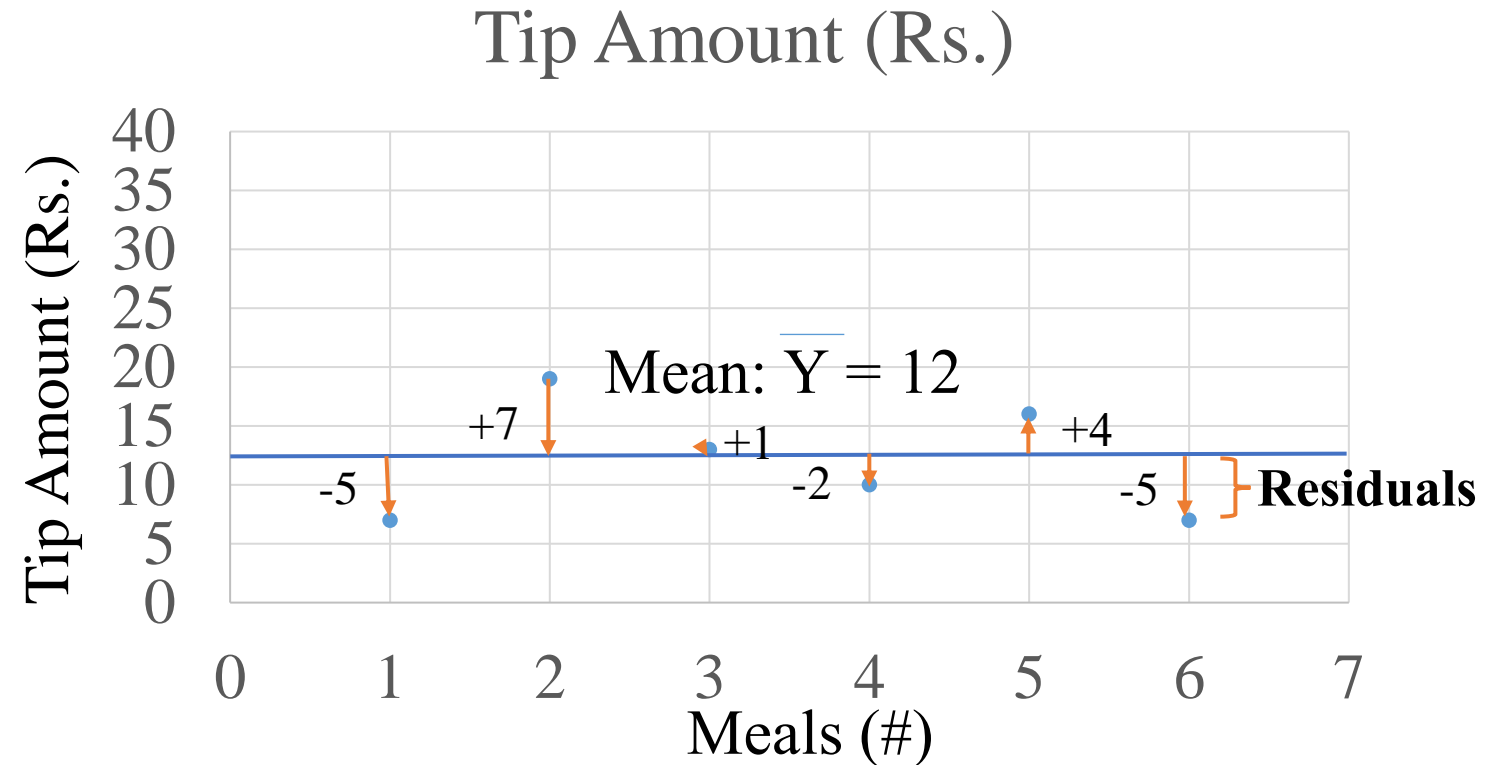
# “Goodness of Fit” for the Tips

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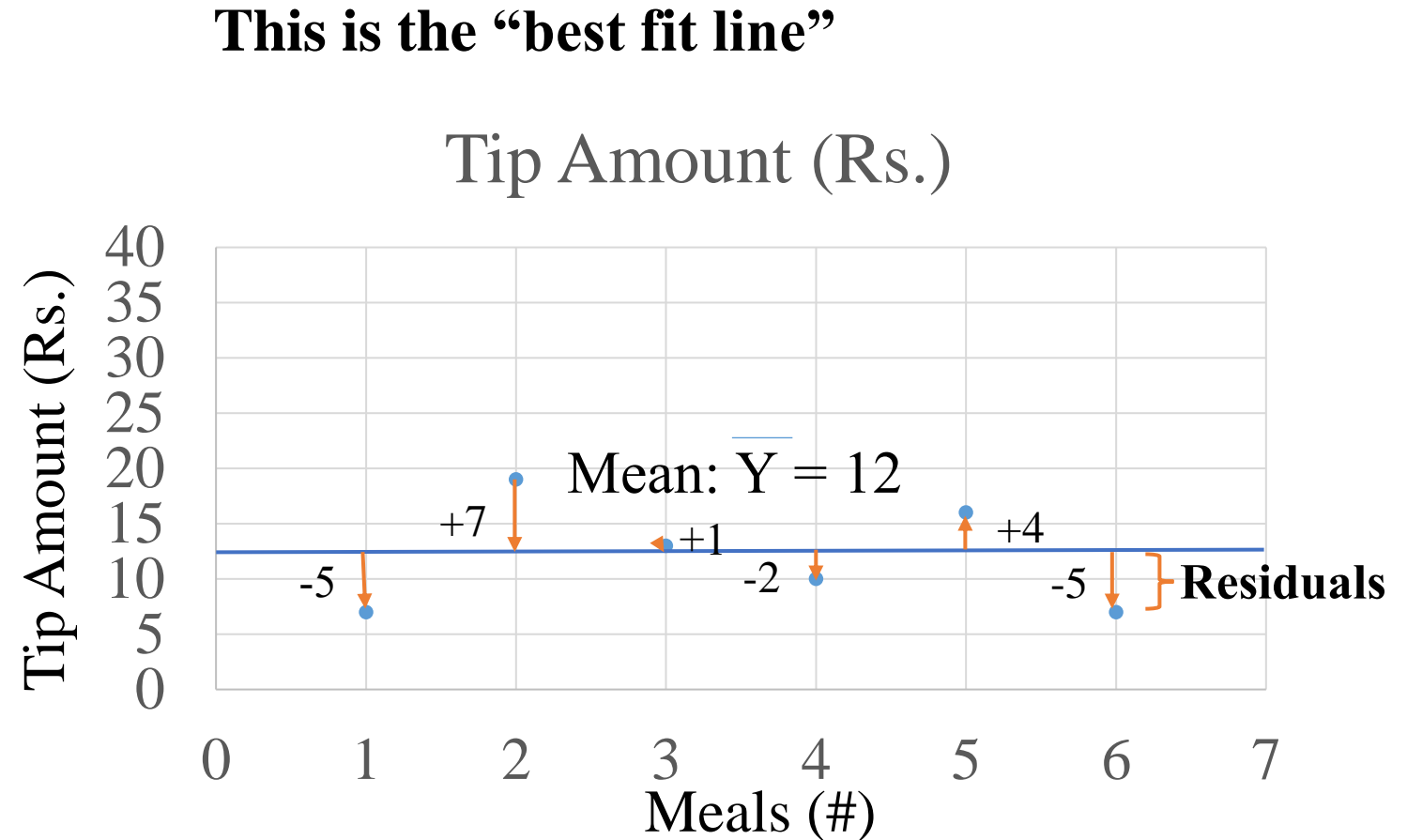
**Mean of Tips:** Some data points are below and some are above it



- The distance every data point is from mean is called as **residuals**
- The distance from the “**best fit line**” to the observed values are called as “**residuals or errors**”

# “Goodness of Fit” for the Tips

Meal (#)	Tip Amount (in Rs.)	Predicted Tip amount (in Rs.)	Difference or error
1.	7	12	25
2.	19	12	49
3.	13	12	1
4.	10	12	4
5.	16	12	16
6.	7	12	25



- **Residuals** will add up & gives **zero**: (i.e.  $-5 + 7 + 1 - 2 + 4 - 5 = 0$ )
- Residual square: **make them positive** and to **emphasize on larger deviations**
- **Sum of Squared errors (SSE) = 120/-**

# Basic Algebra

Slope – intercept form of a line

$$y = mx + b?$$

$x$  = random variable

$m$  = slope of the line rise/run

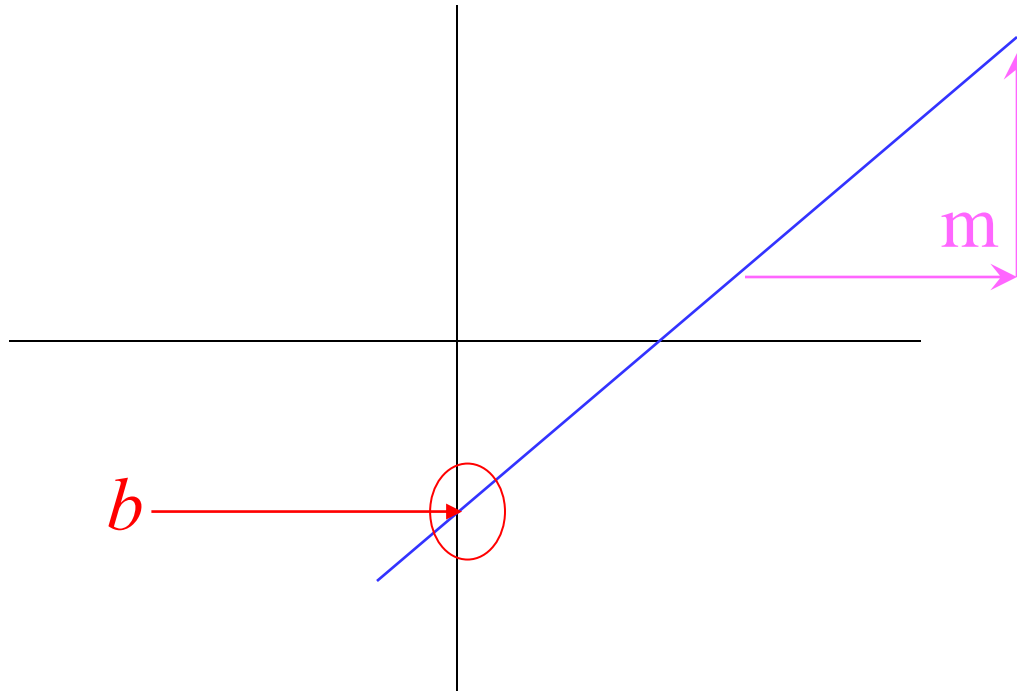
$B$  = intercept where  $x = 0$

$$Y = \beta_0 + \beta_1 x + \epsilon$$

$$\hat{y}_i = b_0 + b_1 x_i + \text{error}$$

$$E(Y) = \beta_0 + \beta_1 x$$

Mean or expected value of  $y$



$$Y = mx + b$$

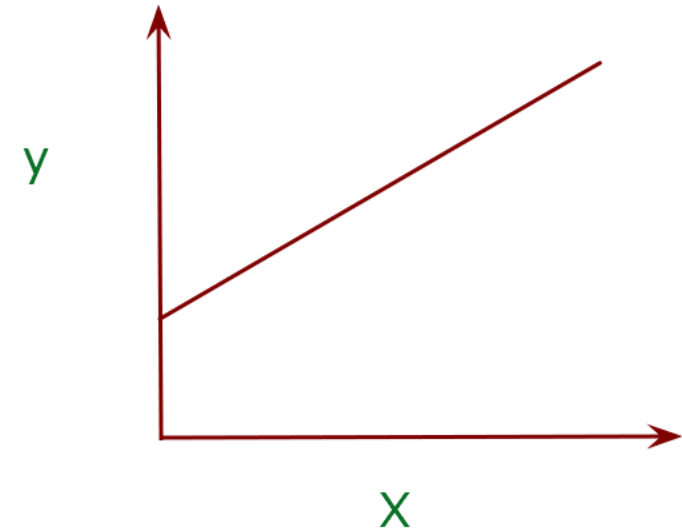
relationship between the regression models

$$\hat{y} = \beta_0 + \beta_1 (x_i) + \text{errors}$$

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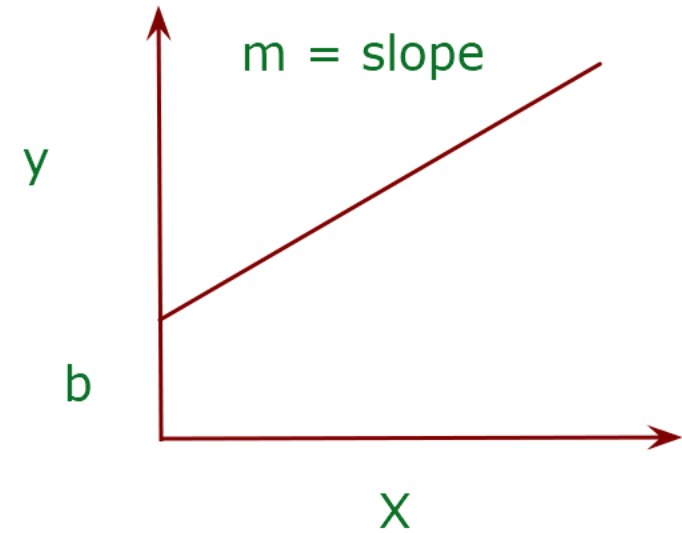
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$$Y = 3X + 6$$

Slope of the line = 3



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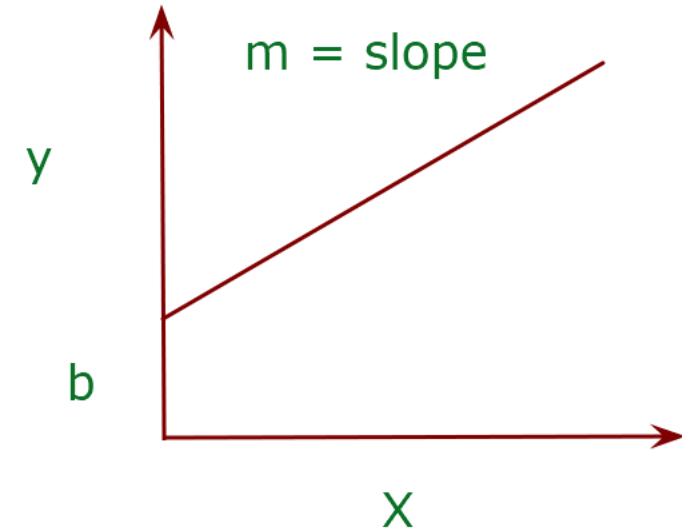
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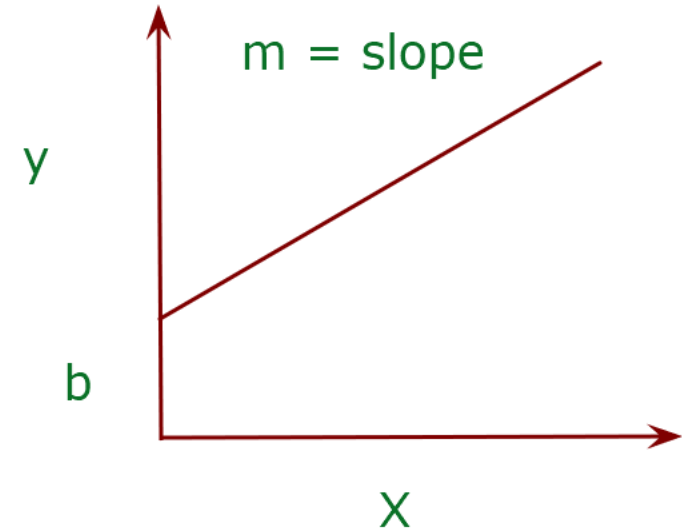
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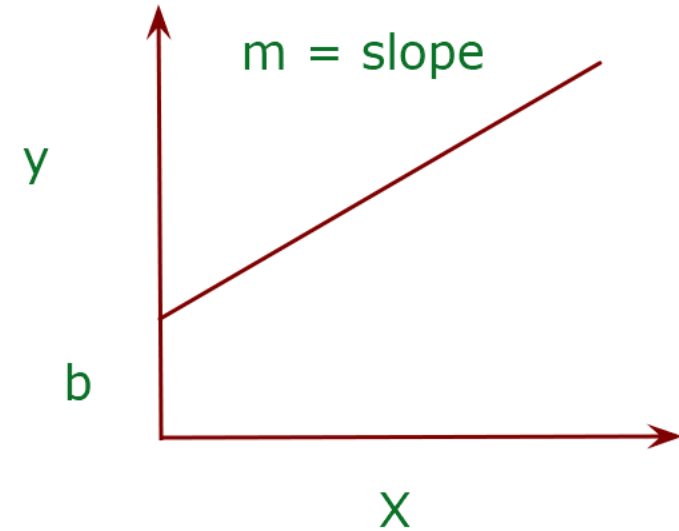
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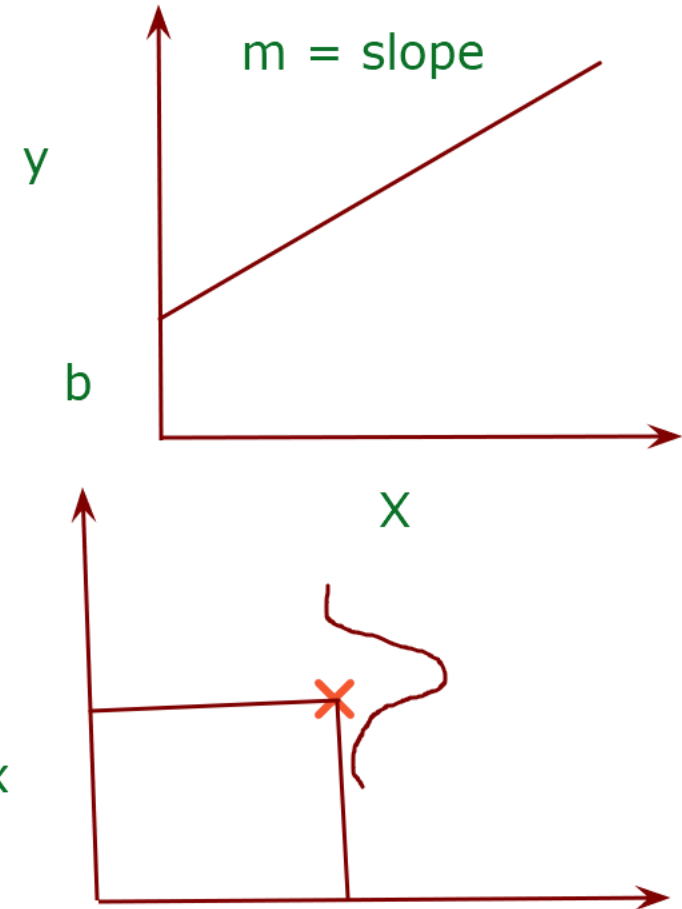
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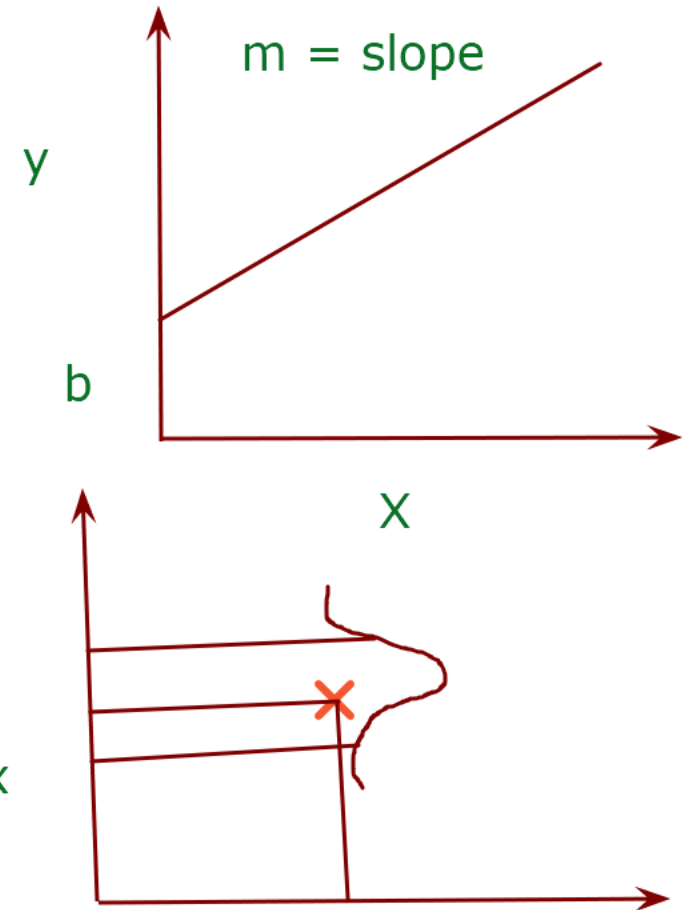
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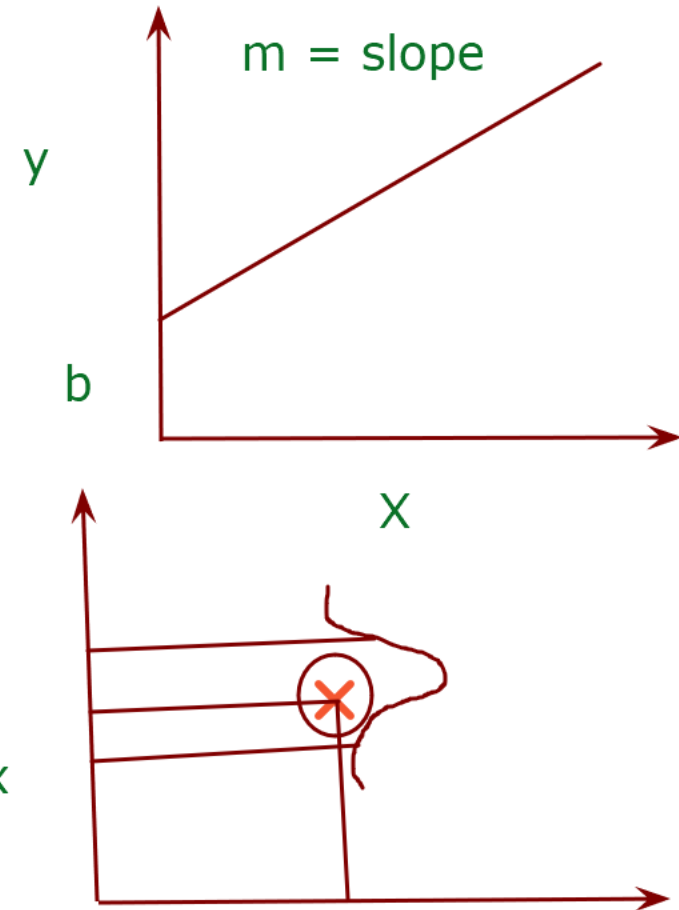
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mean value of the distribution

