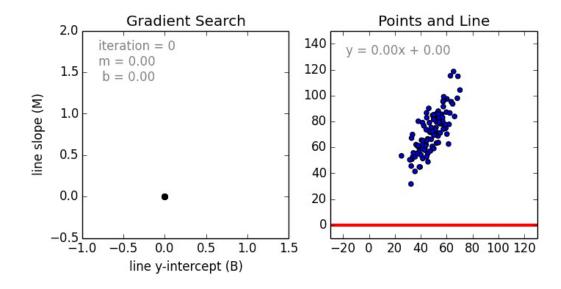
Linear Regression





- Simple Linear regression
- Least Squares Method
- Coefficient of determination
- Model Assumptions
- Testing for Significance
- Using the Estimated regression Equation for Estimation an Prediction



USES OF REGRESSION

Major uses of regression analysis are:-

- Determining the strength of predictors
- Forecasting an effect, and
- Trend Forecasting
- Evaluating Trends and Sales estimates
- Analyzing the Impact of price Changes
- Assessment of risk in financial services and insurance domain







Linear Regression

- The key point in the linear regression is that our dependent value should be continuous
- and cannot be a discreet value. However, the independent variable(s) can be
- measured on either a categorical or continuous measurement scale.
- There are two types of linear regression models.
- They are: simple regression and multiple regression.

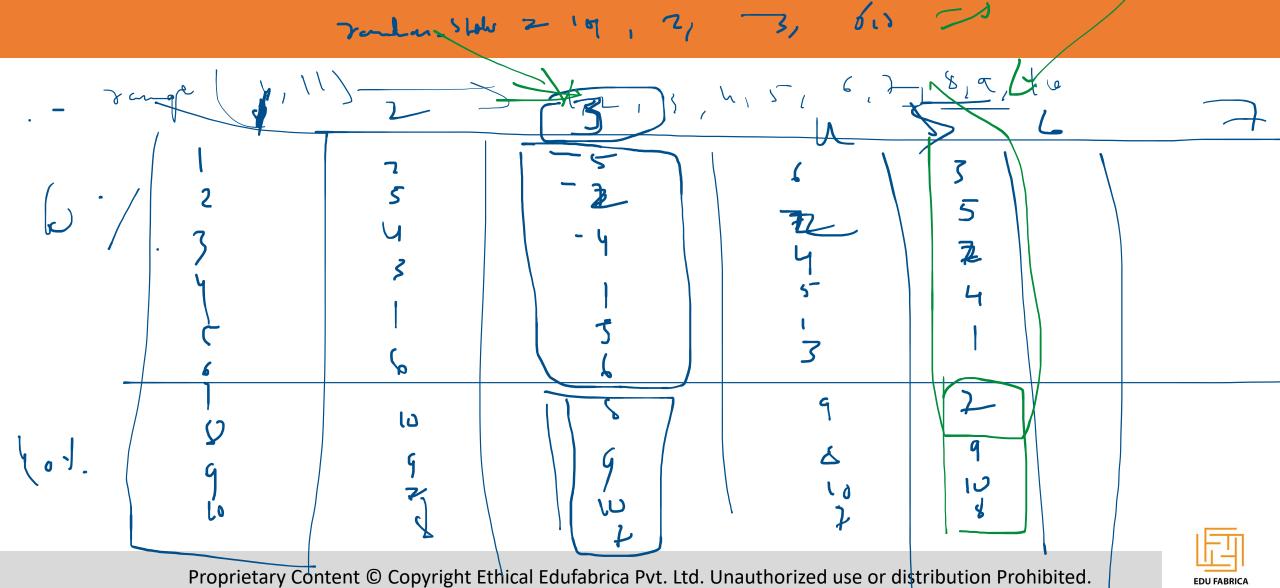


SELECTION CRITERIA

- Classification and Regression Capabilities
- Data Quality
- Computational Complexity
- Comprehensible and Transparent



Random State



| | ENGINESIZE | CYLINDERS | FUELCONSUMPTION_COMB | CO2EMISSIONS |
|---|-------------------|-----------|----------------------|--------------|
| 0 | 2.0 | 4 | 8.5 | 196 |
| 1 | 2.4 | 4 | 9.6 | 221 |
| 2 | 1.5 | 4 | 5.9 | 136 |
| 3 | 3.5 | 6 | 11.1 | 255 |
| 4 | 3.5 | 6 | 10.6 | 244 |
| 5 | 3.5 | 6 | 10.0 | 230 |
| 6 | 3.5 | 6 | 10.1 | 232 |
| 7 | 3.7 | 6 | 11.1 | 255 |
| 8 | 3.7 | 6 | 11.6 | 267 |
| 9 | 2.4 | 4 | 9.2 | ? |



| С | ENGINESIZE | CYLINDERS | FUELCONSUMPTION_COMB | CO2EMISSIONS |
|---|------------|-----------|----------------------|--------------|
| 0 | 2.0 | 4 | 8.5 | 196 |
| 1 | 2.4 | 4 | 9.6 | 221 |
| 2 | 1.5 | 4 | 5.9 | 136 |
| 3 | 3.5 | 6 | 11.1 | 255 |
| 4 | 3.5 | 6 | 10.6 | 244 |
| 5 | 3.5 | 6 | 10.0 | 230 |
| 6 | 3.5 | 6 | 10.1 | 232 |
| 7 | 3.7 | 6 | 11.1 | 255 |
| 8 | 3.7 | 6 | 11.6 | 267 |
| 9 | 2.4 | 4 | 9.2 | ? |

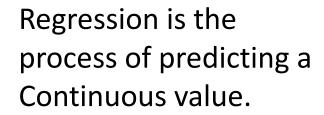




| | ENGINESIZE | CYLINDERS | FUELCONSUMPTION_COMB | CO2EMISSIONS |
|---|------------|-----------|----------------------|--------------|
| 0 | 2.0 | 4 | 8.5 | 196 |
| 1 | 2.4 | 4 | 9.6 | 221 |
| 2 | 1.5 | 4 | 5.9 | 136 |
| 3 | 3.5 | 6 | 11.1 | 255 |
| 4 | 3.5 | 6 | 10.6 | 244 |
| 5 | 3.5 | 6 | 10.0 | 230 |
| 6 | 3.5 | 6 | 10.1 | 232 |
| 7 | 3.7 | 6 | 11.1 | 255 |
| 8 | 3.7 | 6 | 11.6 | 267 |
| 9 | 2.4 | 4 | 9.2 | ? |



| L | ENGINESIZE | CYLINDERS | FUELCONSUMPTION_COMB | CO2EMISSIONS |
|---|------------|-----------|----------------------|--------------|
| 0 | 2.0 | 4 | 8.5 | 196 |
| 1 | 2.4 | 4 | 9.6 | 221 |
| 2 | 1.5 | 4 | 5.9 | 136 |
| 3 | 3.5 | 6 | 11.1 | 255 |
| 4 | 3.5 | 6 | 10.6 | 244 |
| 5 | 3.5 | 6 | 10.0 | 230 |
| 6 | 3.5 | 6 | 10.1 | 232 |
| 7 | 3.7 | 6 | 11.1 | 255 |
| 8 | 3.7 | 6 | 11.6 | 267 |
| 9 | 2.4 | 4 | 9.2 | (?) |





X : Independent Variable

Y : Dependent Variable

| 200 | ENGINESIZE | CYLINDERS | FUELCONSUMPTION_COMB | CO2EMISSIONS |
|-----|------------|-----------|----------------------|--------------|
| 0 | 2.0 | 4 | 8.5 | 196 |
| 1 | 2.4 | 4 | 9.6 | 221 |
| 2 | 1.5 | 4 | 5.9 | 136 |
| 3 | 3.5 | 6 | 11.1 | 255 |
| 4 | 3.5 | 6 | 10.6 | 244 |
| 5 | 3.5 | 6 | 10.0 | 230 |
| 6 | 3.5 | 6 | 10.1 | 232 |
| 7 | 3.7 | 6 | 11.1 | 255 |
| 8 | 3.7 | 6 | 11.6 | 267 |
| 9 | 2.4 | 4 | 9.2 | (?) |
| | | | | |

Regression is the process of predicting a Continuous value.

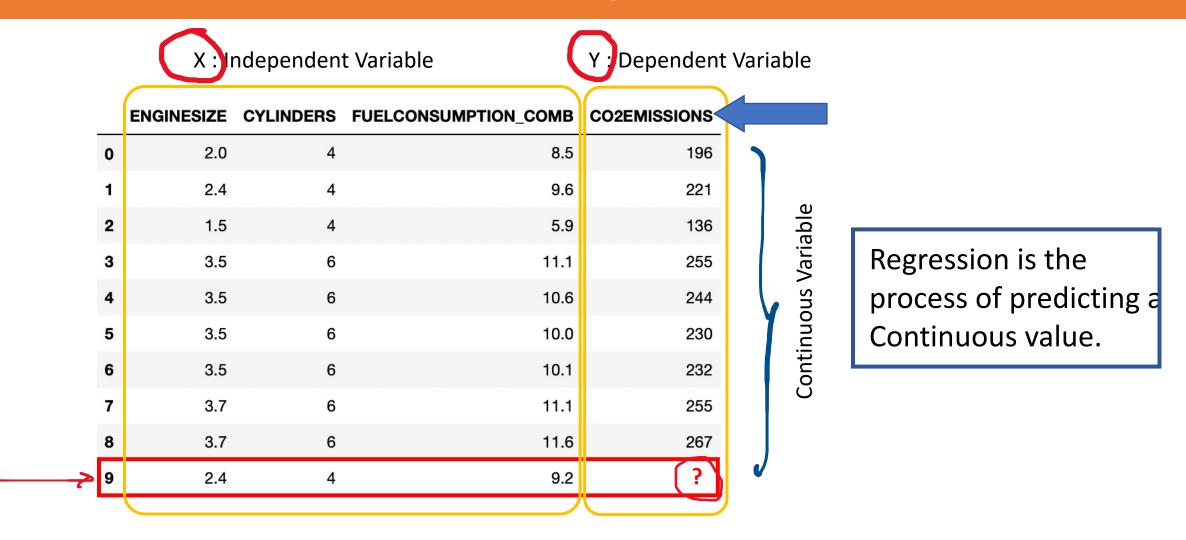


Y: Dependent Variable X : Independent Variable ENGINESIZE CYLINDERS FUELCONSUMPTION_COMB CO2EMISSIONS 2.0 8.5 196 0 4 2.4 9.6 221 2 1.5 5.9 136 Regression is the 3.5 11.1 3 6 255 process of predicting a 10.6 3.5 6 244 4 Continuous value. 3.5 10.0 6 230 3.5 6 10.1 232 6 3.7 11.1 6 255 7 3.7 11.6 267 6 2.4 9.2 9 4



X : Independent Variable Y : Dependent Variable ENGINESIZE CYLINDERS FUELCONSUMPTION_COMB CO2EMISSIONS 2.0 8.5 196 0 4 2.4 9.6 221 2 1.5 5.9 136 Regression is the 3.5 11.1 3 6 255 process of predicting a 10.6 3.5 6 244 4 Continuous value. 3.5 10.0 6 230 3.5 6 10.1 232 6 3.7 11.1 6 255 7 3.7 11.6 267 6 2.4 9.2 9 4







Types of Linear Regression

Simple Linear Regression

- Simple Linear Regression
- Simple Non Linear Regression

Predict co2emission vs EngineSize of all cars

Multiple Regression

- Multiple Linear Regression
- Multiple Non Linear Regression

Predict co2emission vs EngineSize and Cylinders of all cars



Empirical Model

- As an illustration, consider the data in the table
- In this table y is the purity of oxygen produced in a chemical distillation process, and x is the percentage of hydro carbons that are present tin the main condenser of the distillation unit.

| Hydrocarbon Level (X) | Purity (Y) |
|-----------------------|------------|
| 0.99 | 90.01 |
| 1.02 | 89.05 |
| 1.15 | 91.43 |
| 1.29 | 93.74 |
| 1.46 | 96.73 |
| 1.36 | 94.45 |
| 0.87 | 87.59 |
| 1.23 | 91.77 |
| 1.55 | 99.42 |
| 1.4 | 93.65 |



Linear vs Logistic

| Basis | Linear Regression | Logistic Regression | |
|------------------------------|--|---|--|
| Core Concept | The data is modelled using a straight line | The probability of some obtained event is represented as a linear function of a combination of predictor variables. | |
| Used with | Continuous Variables | Categorical Variable | |
| Output / prediction | Values of the Variable | Probability of occurrence of the event. | |
| Accuracy and Goodness of fit | Measured by Loss, R squared, Adjusted R squared etc. | Accuracy, Precision, Recall, F1 Score, ROC curve, Confusion Matrix , etc. | |



Simple Linear Regression Model

- Regression analysis. is a form of predictive modelling technique which investigates the relationship the relationship between a dependent and independent variable.
- The equation that describes how y is related to x and an error term is called the regression model.
- The Simple Linear Regression Model is :

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

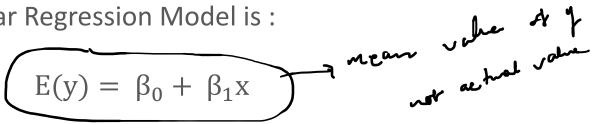
Where:

 β_0 and β_1 are called parameters of the model, ϵ is a random variable called the error term.



Simple Linear Regression Model

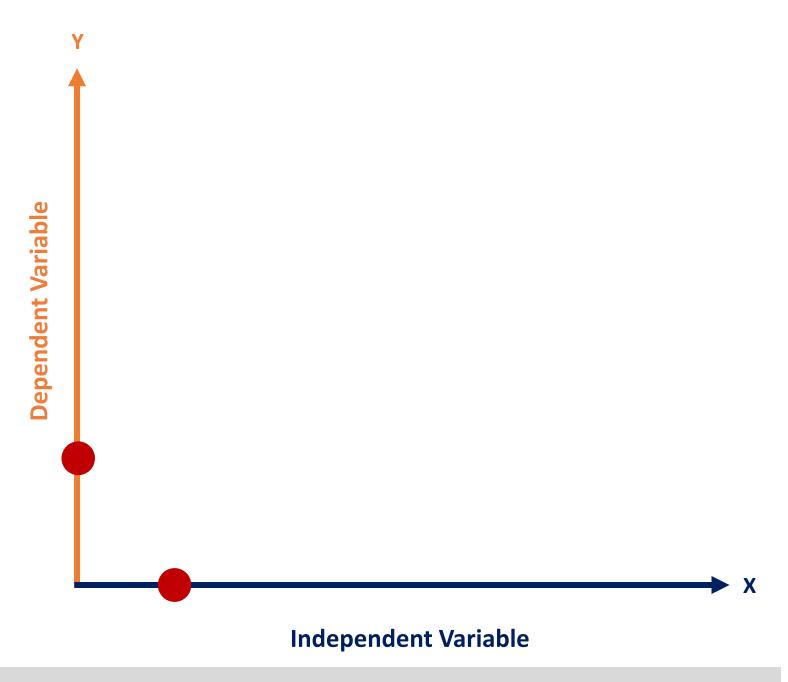
The Simple Linear Regression Model is:



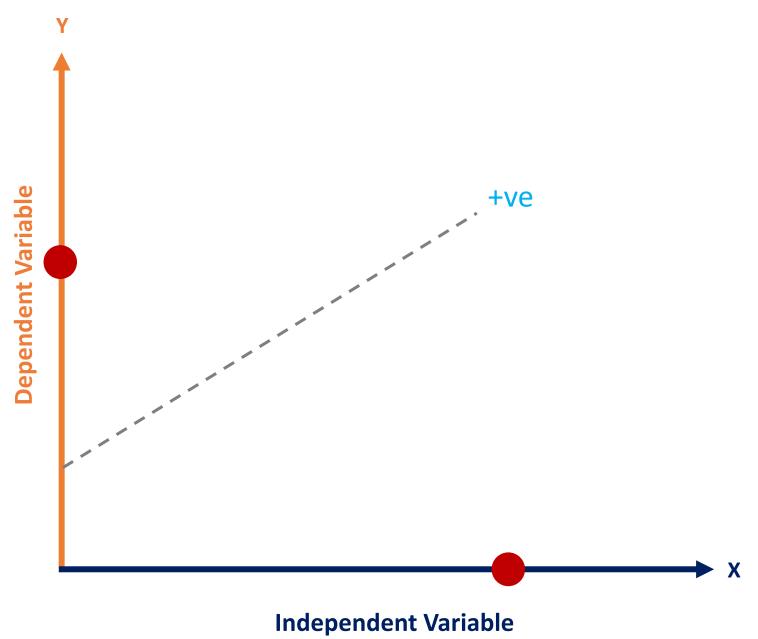
- Graph of the regression equation is a straight line
- β_0 is the intercept of the regression line
- B_1 is the slope of the regression line
- E(y) is the expected value of y for a given x value



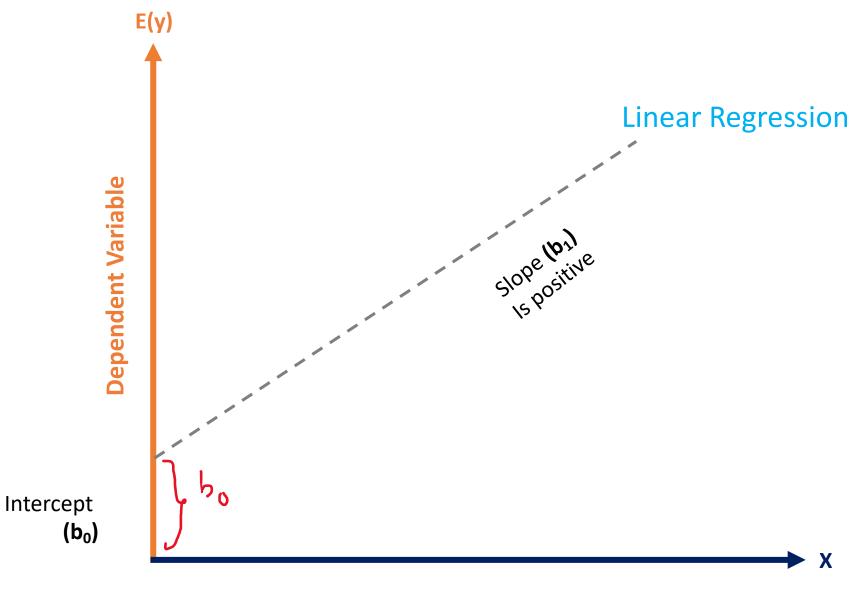
Simple Linear Regression Model











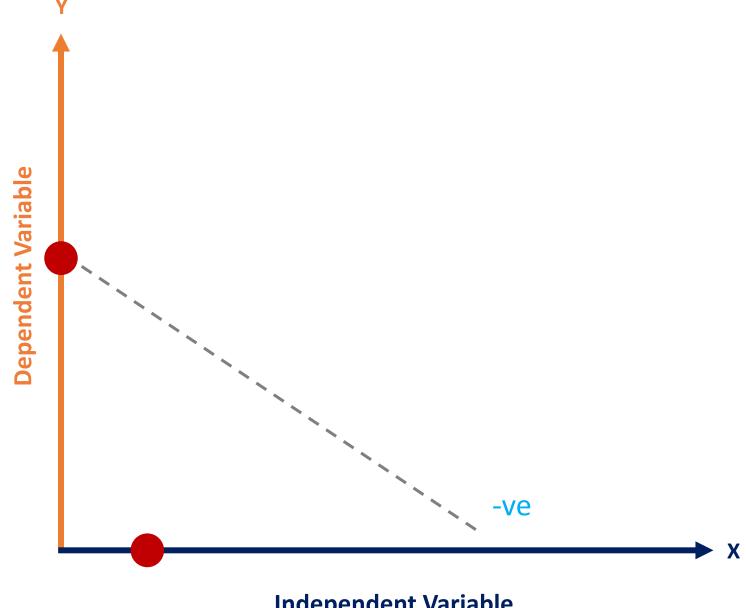






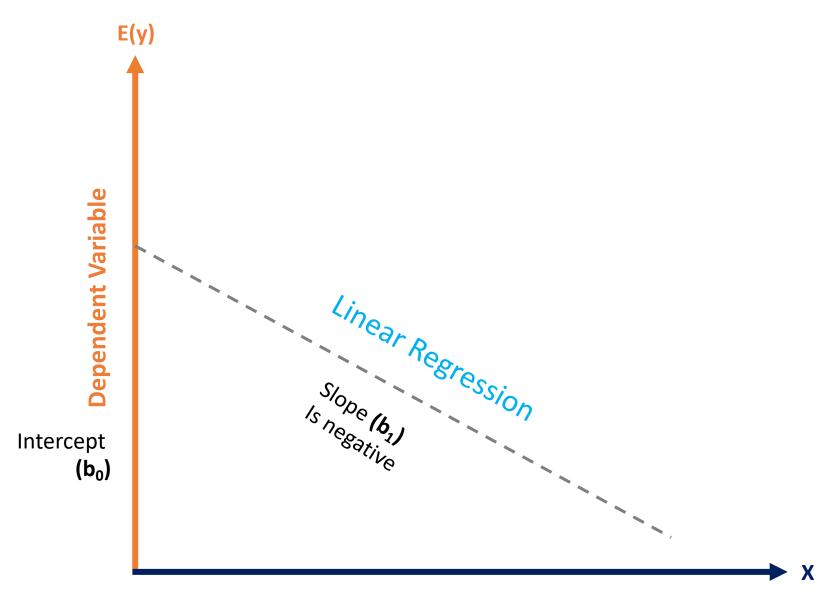












Independent Variable



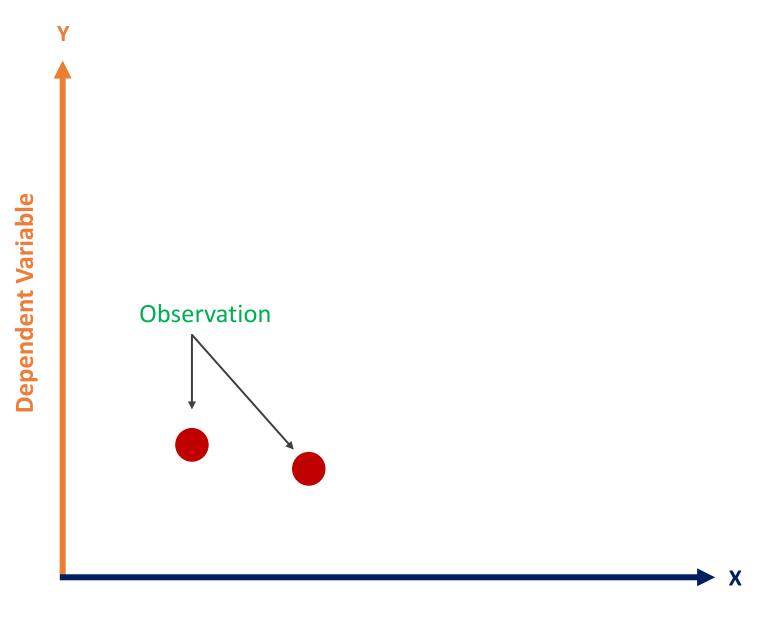
Estimated Simple Linear Regression Equation

• The estimated Simple Linear Regression equation is :

$$\hat{y} = b_0 + b_1 x$$

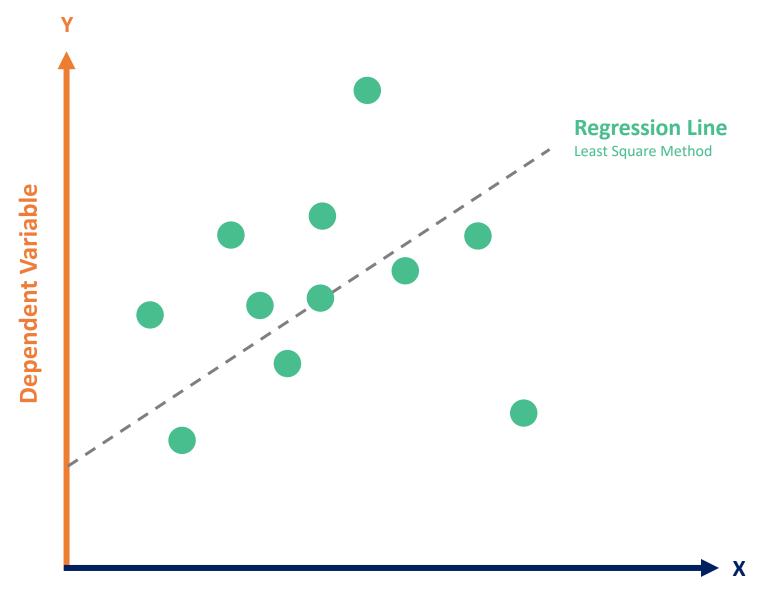
- The graph is called the estimated regression line
- b₀ is the intercept of the line
- b₁ is the slope of the line
- \hat{y} is the estimated value of y for a given x value





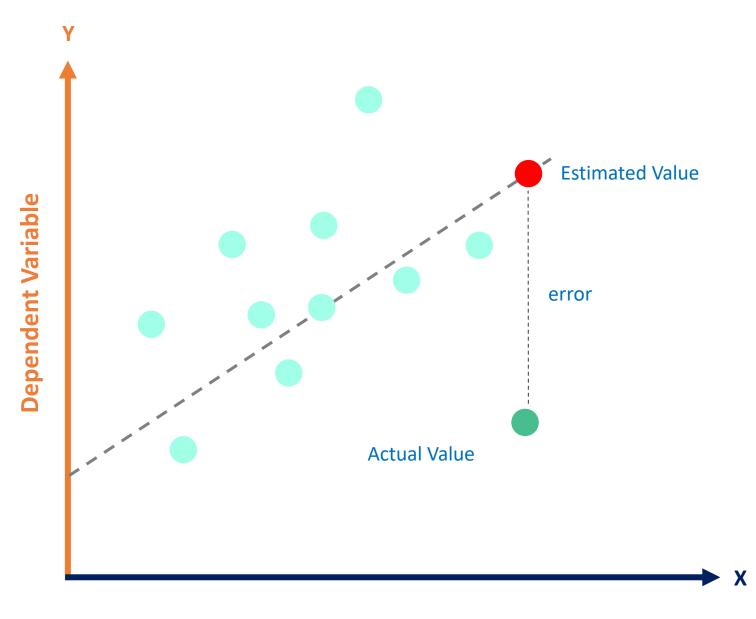
Independent Variable





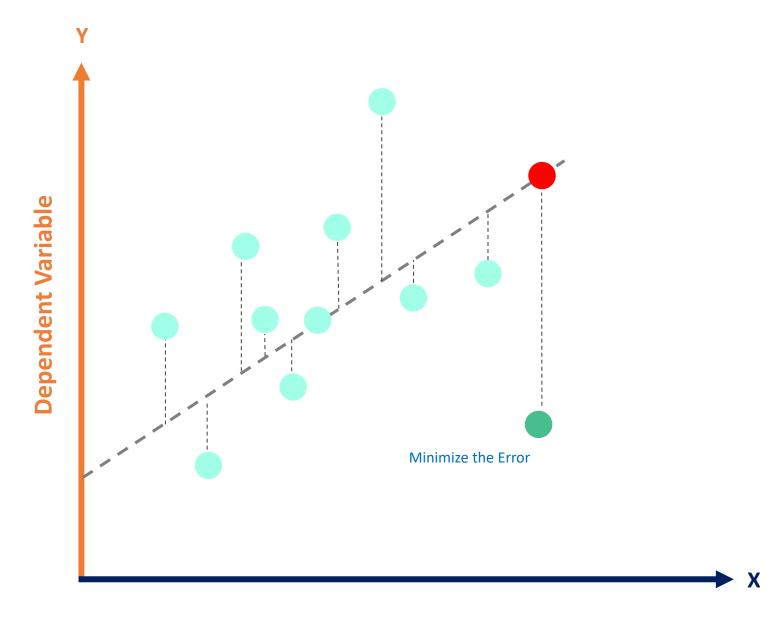






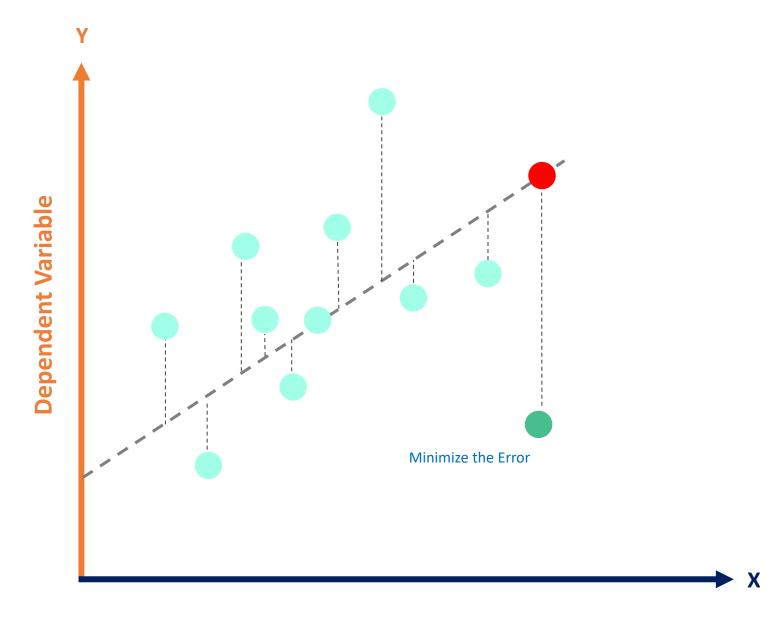






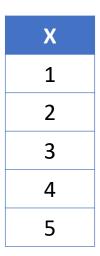






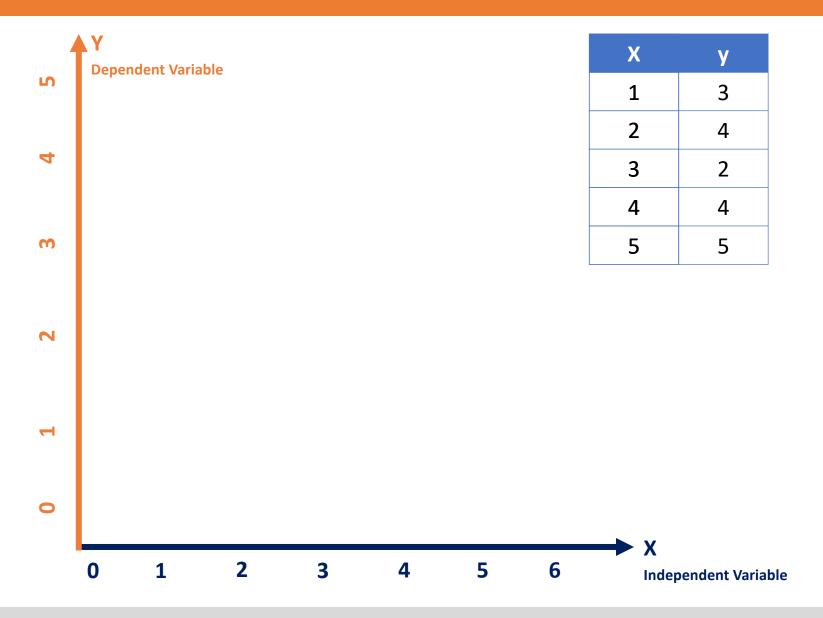




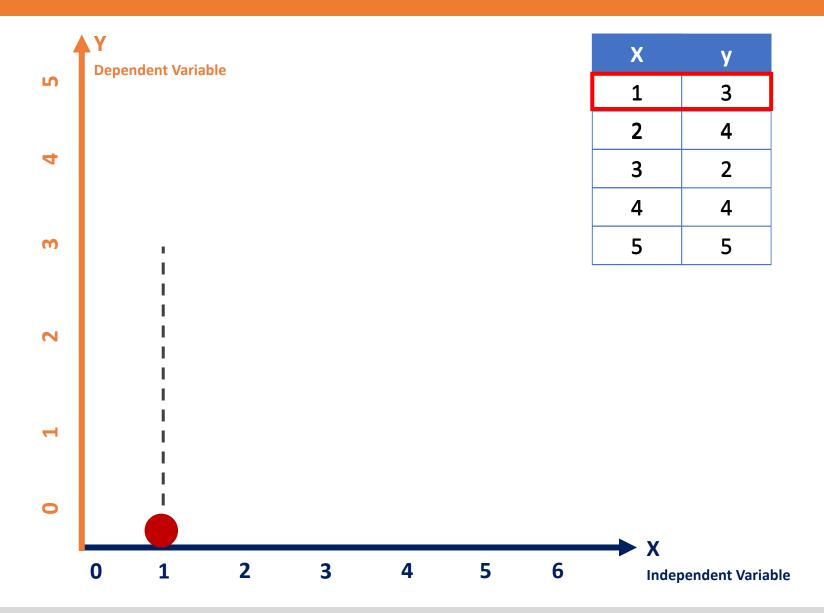




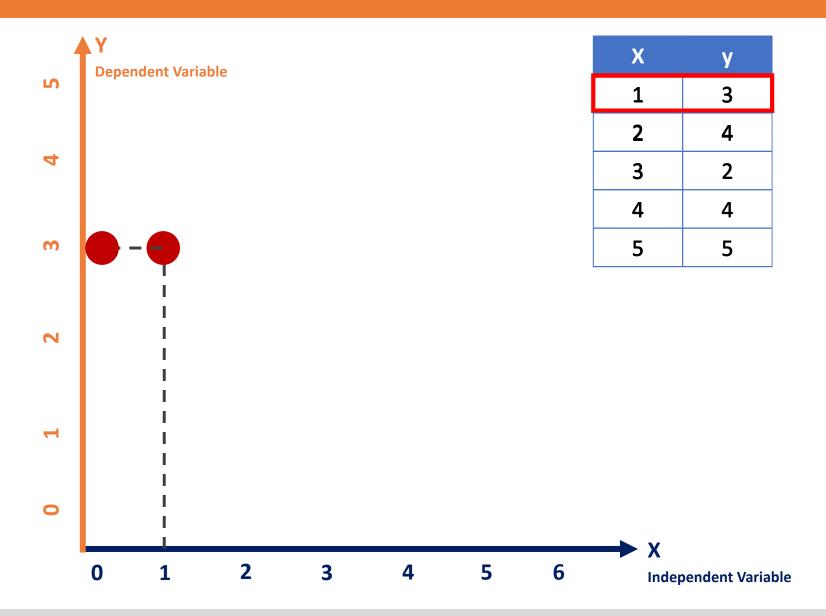




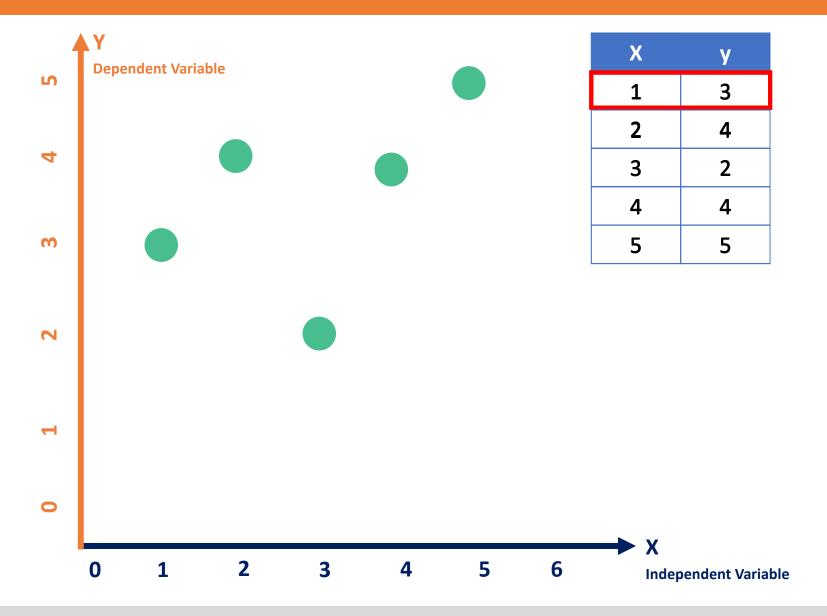




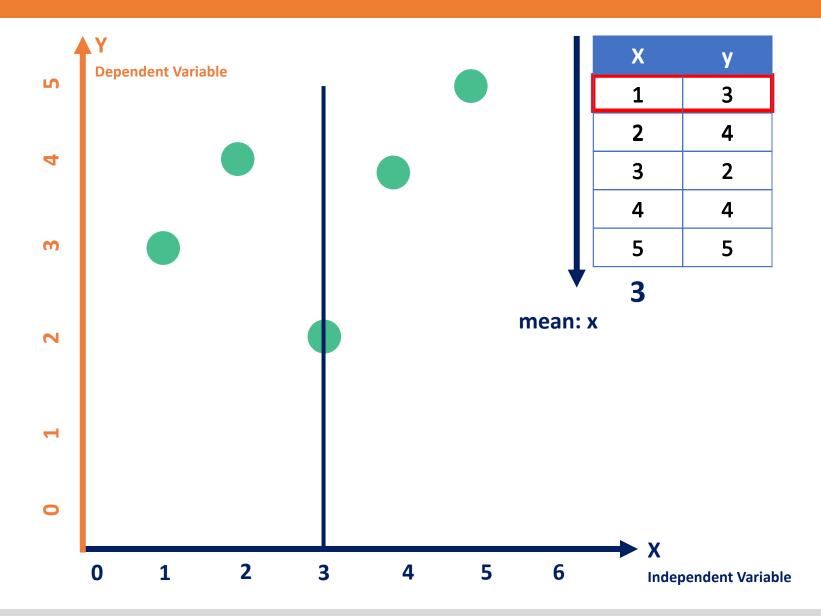




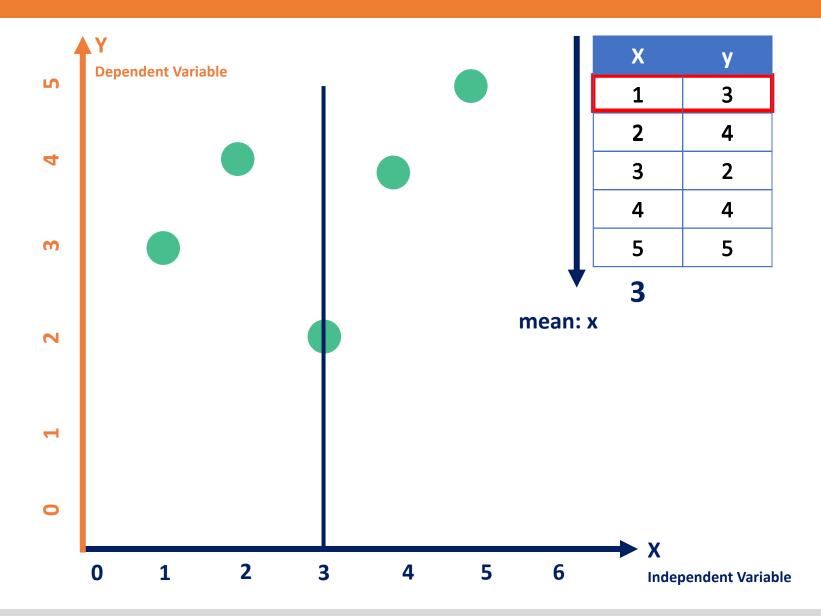




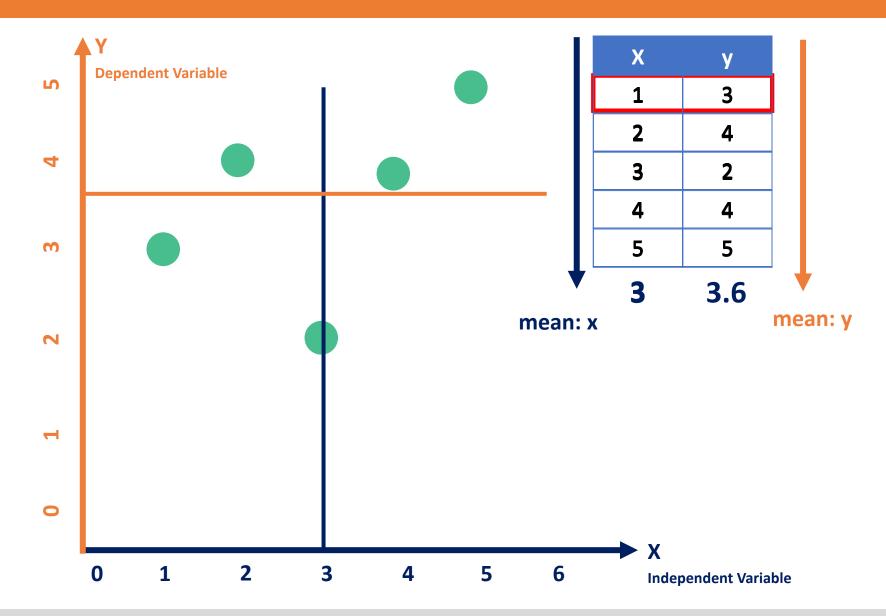




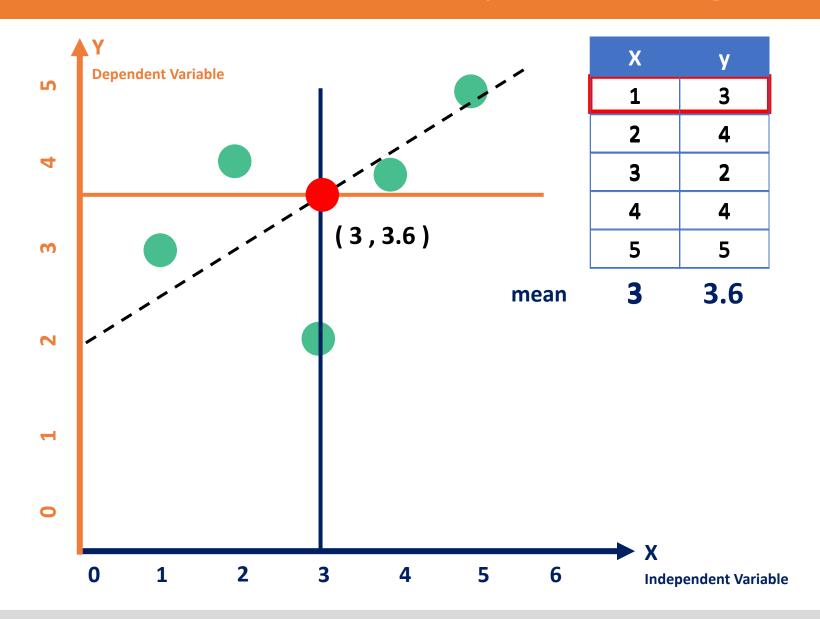




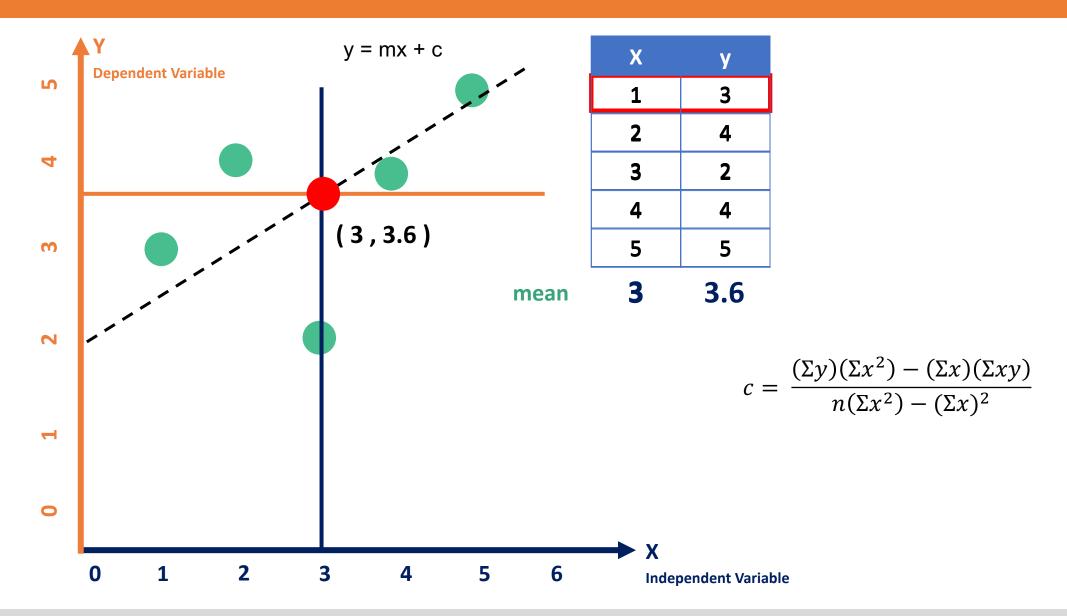




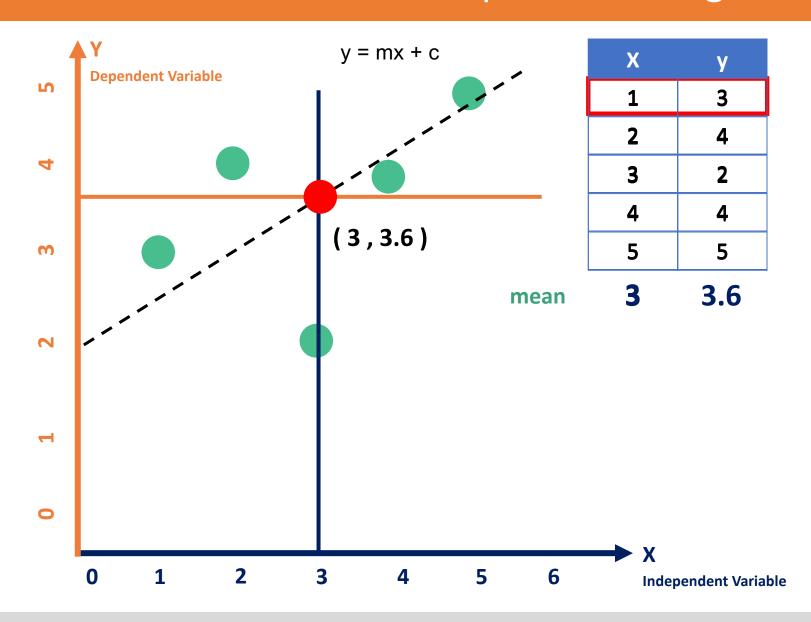








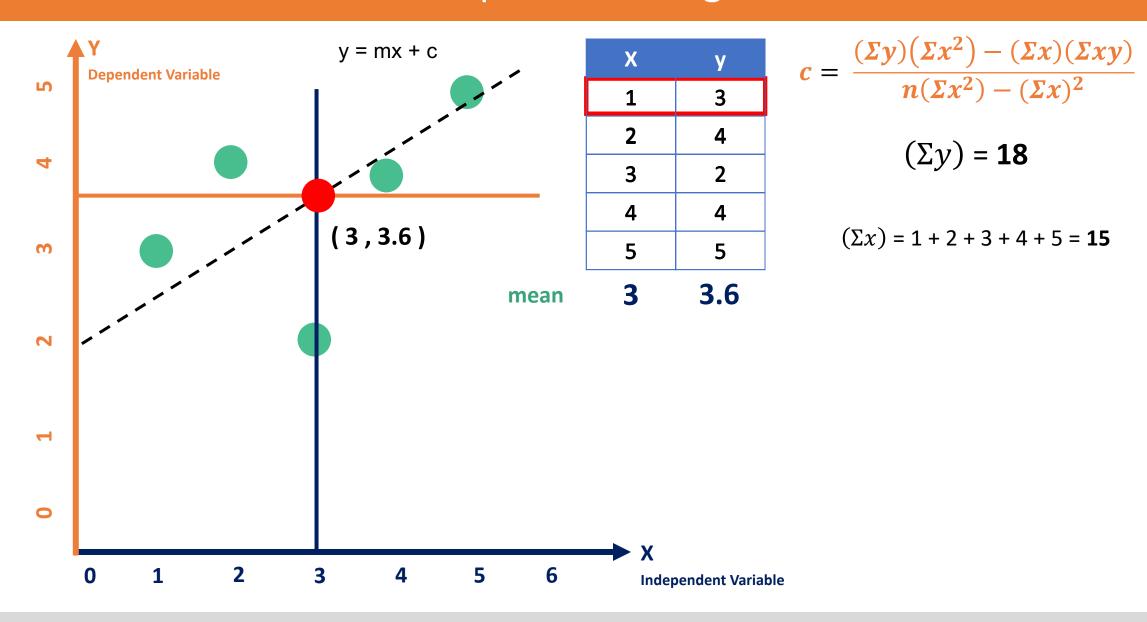


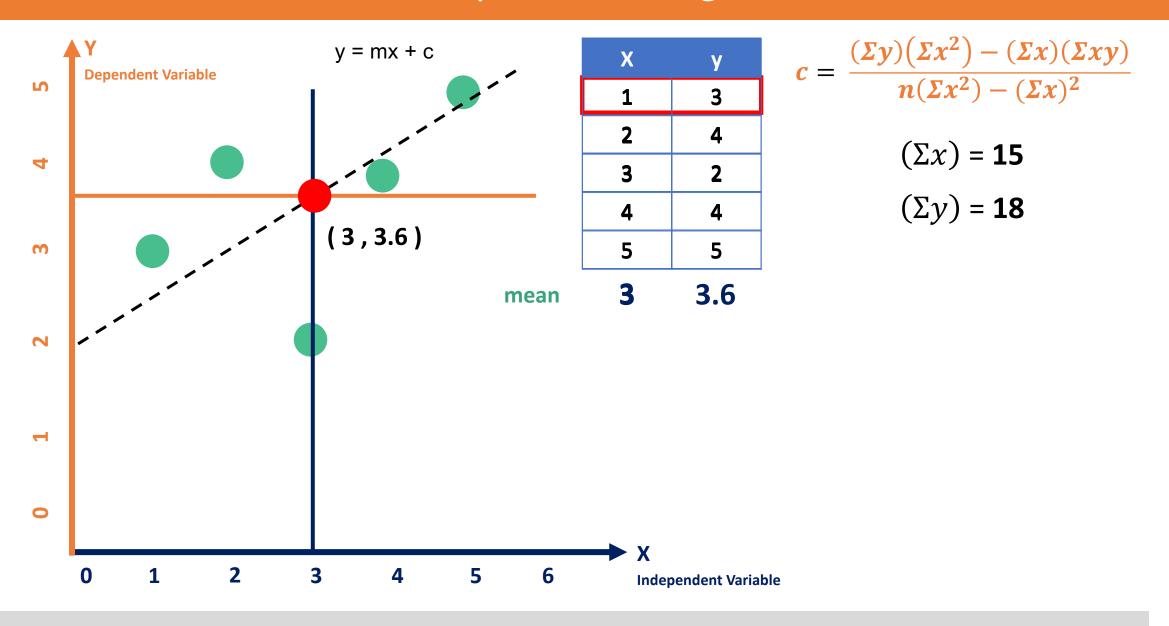


$$c = \frac{(\Sigma y)(\Sigma x^2) - (\Sigma x)(\Sigma xy)}{n(\Sigma x^2) - (\Sigma x)^2}$$

$$(\Sigma y)$$
 = 3 + 4 + 2 + 4 + 5 = **18**







| X | У |
|---|---|
| 1 | 3 |
| 2 | 4 |
| 3 | 2 |
| 4 | 4 |
| 5 | 5 |
| 3 | |



| X | У | xy | x^2 |
|---|---|----|-------|
| 1 | 3 | 3 | 1 |
| 2 | 4 | 8 | 4 |
| 3 | 2 | 6 | 9 |
| 4 | 4 | 16 | 16 |
| 5 | 5 | 25 | 25 |

$$c = \frac{(\Sigma y)(\Sigma x^2) - (\Sigma x)(\Sigma xy)}{n(\Sigma x^2) - (\Sigma x)^2}$$

$$(\Sigma x)$$
 = **15** (Σy) = **18**

$$(\Sigma xy) = 3 + 8 + 6 + 16 + 25 = 58$$

$$y = mx + c$$



| X | у | xy | x^2 |
|---|---|----|-------|
| 1 | 3 | 3 | 1 |
| 2 | 4 | 8 | 4 |
| 3 | 2 | 6 | 9 |
| 4 | 4 | 16 | 16 |
| 5 | 5 | 25 | 25 |

$$c = \frac{(\Sigma y)(\Sigma x^2) - (\Sigma x)(\Sigma xy)}{n(\Sigma x^2) - (\Sigma x)^2}$$

$$(\Sigma x)$$
 = **15** (Σy) = **18**

$$(\Sigma xy) = 58$$

$$(\Sigma x^2) = 1 + 4 + 9 + 16 + 25 = 55$$



| X | у | xy | x^2 |
|---|---|----|-------|
| 1 | 3 | 3 | 1 |
| 2 | 4 | 8 | 4 |
| 3 | 2 | 6 | 9 |
| 4 | 4 | 16 | 16 |
| 5 | 5 | 25 | 25 |

$$c = \frac{(\Sigma y)(\Sigma x^2) - (\Sigma x)(\Sigma xy)}{n(\Sigma x^2) - (\Sigma x)^2}$$

$$(\Sigma x)$$
 = **15** (Σy) = **18**

$$(\Sigma xy) = 58$$
 $(\Sigma x^2) = 55$



| X | У | xy | x^2 |
|---|---|----|-------|
| 1 | 3 | 3 | 1 |
| 2 | 4 | 8 | 4 |
| 3 | 2 | 6 | 9 |
| 4 | 4 | 16 | 16 |
| 5 | 5 | 25 | 25 |

$$c = \frac{(\Sigma y)(\Sigma x^2) - (\Sigma x)(\Sigma xy)}{n(\Sigma x^2) - (\Sigma x)^2}$$

$$c = \frac{(18)(55) - (15)(58)}{5(55) - (225)}$$



| X | У | xy | x^2 |
|---|---|----|-------|
| 1 | 3 | 3 | 1 |
| 2 | 4 | 8 | 4 |
| 3 | 2 | 6 | 9 |
| 4 | 4 | 16 | 16 |
| 5 | 5 | 25 | 25 |

$$y = mx + c$$

$$c = \frac{(\Sigma y)(\Sigma x^2) - (\Sigma x)(\Sigma xy)}{n(\Sigma x^2) - (\Sigma x)^2}$$

$$c = \frac{(18)(55) - (15)(58)}{5(55) - (225)} = \frac{990 - 870}{275 - (225)}$$

$$c = \frac{120}{50} = 12 / 5 = 2.4$$

$$C = 2.4$$



| X | у | xy | x^2 |
|---|---|----|-------|
| 1 | 3 | 3 | 1 |
| 2 | 4 | 8 | 4 |
| 3 | 2 | 6 | 9 |
| 4 | 4 | 16 | 16 |
| 5 | 5 | 25 | 25 |

$$y = mx + c$$

$$m = \frac{n(\Sigma xy) - (\Sigma x)(\Sigma y)}{n(\Sigma x^2) - (\Sigma x)^2}$$

$$m = \frac{5(58) - (15)(18)}{5(55) - (15)^2}$$

$$m=\frac{290-270}{275-225}$$

$$m=\frac{20}{70}=\frac{2}{7}$$
= 0.28



| X | У | xy | x ² |
|----|----|-----|----------------|
| 1 | 3 | 3 | 1 |
| 2 | 4 | 8 | 4 |
| 3 | 2 | 6 | 9 |
| 4 | 4 | 16 | 16 |
| 5 | 5 | 25 | 25 |
| 6 | 4 | 24 | 36 |
| 7 | 12 | 84 | 49 |
| 8 | 15 | 120 | 64 |
| 9 | 18 | 162 | 81 |
| 10 | 20 | 200 | 100 |

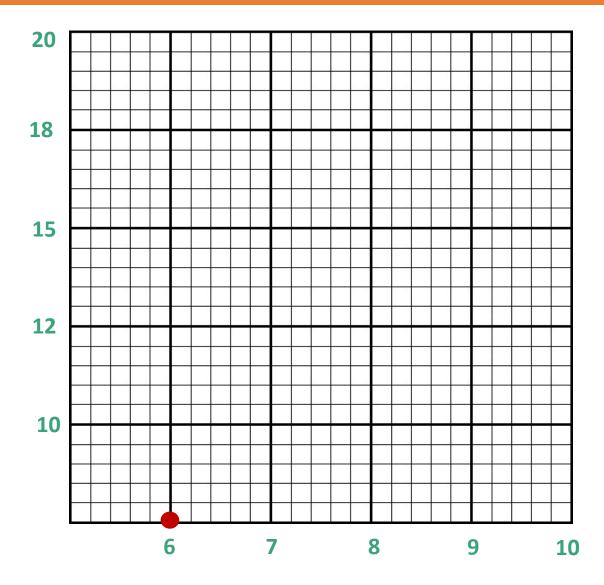
$$y = mx + c$$
;
 $m = 0.28$; $c = 2.4$

$$y = 0.28x + 2.4$$

Let's calculate y for

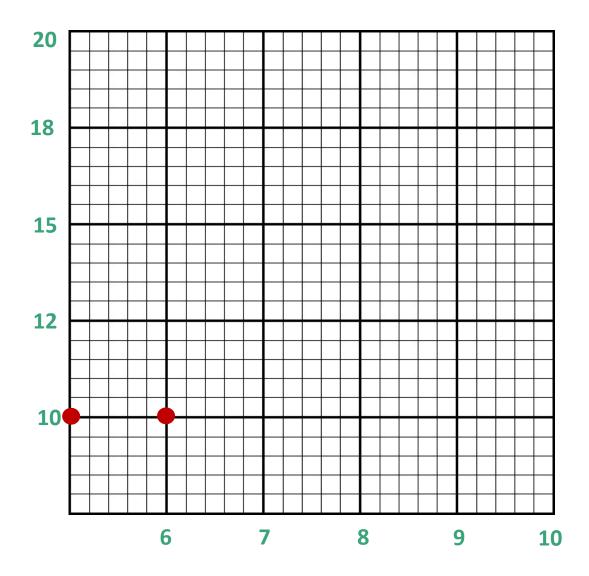
$$X = \{6,7,8,9,10\}$$





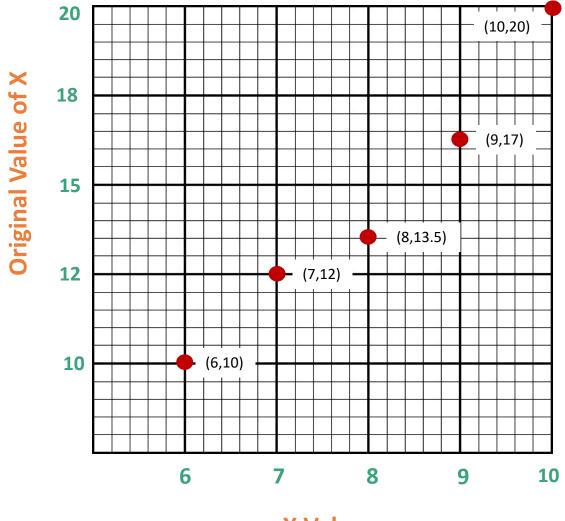
| X | У | xy | x ² |
|----|----|-----|----------------|
| 6 | 10 | 24 | 36 |
| 7 | 12 | 84 | 49 |
| 8 | 15 | 120 | 64 |
| 9 | 18 | 162 | 81 |
| 10 | 20 | 200 | 100 |





| X | У | xy | x^2 |
|----|----|-----|-------|
| 6 | 10 | 24 | 36 |
| 7 | 12 | 84 | 49 |
| 8 | 15 | 120 | 64 |
| 9 | 18 | 162 | 81 |
| 10 | 20 | 200 | 100 |





| X | У | xy | x^2 |
|----|------|-----|-------|
| 6 | 10 | 24 | 36 |
| 7 | 12 | 84 | 49 |
| 8 | 13.5 | 120 | 64 |
| 9 | 17 | 162 | 81 |
| 10 | 20 | 200 | 100 |

X Value

