

# *Regression*

*Supervised learning is a method in which we teach the machine using labelled data*



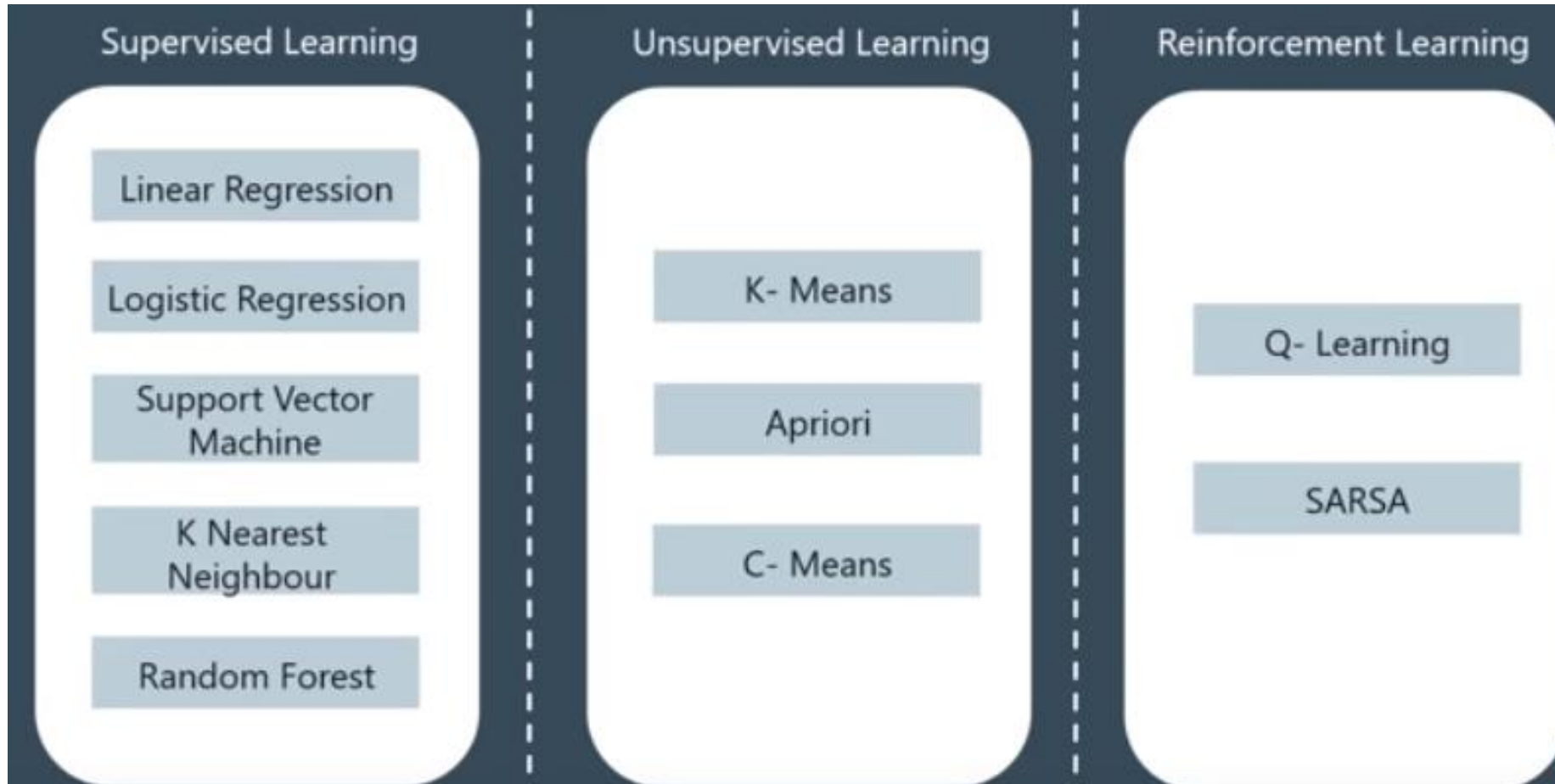
*In unsupervised learning the machine is trained on unlabelled data without any guidance*



*In Reinforcement learning an agent interacts with its environment by producing actions & discovers errors or rewards*



# Algorithm



# Applications

## Supervised Learning

Risk Evaluation



Forecast Sales



## Unsupervised Learning

Recommendation Systems



Anomaly Detection



## Reinforcement Learning

Self driving cars



Gaming



# What is Regression?

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**“Regression analysis is a form of predictive modelling technique which investigates the relationship between a dependent and independent variable”**

It is a technique used to:

- Estimate a relationship between variables
- Predict the value of one variable (dependent variable) on the basis of other variables (independent variables)



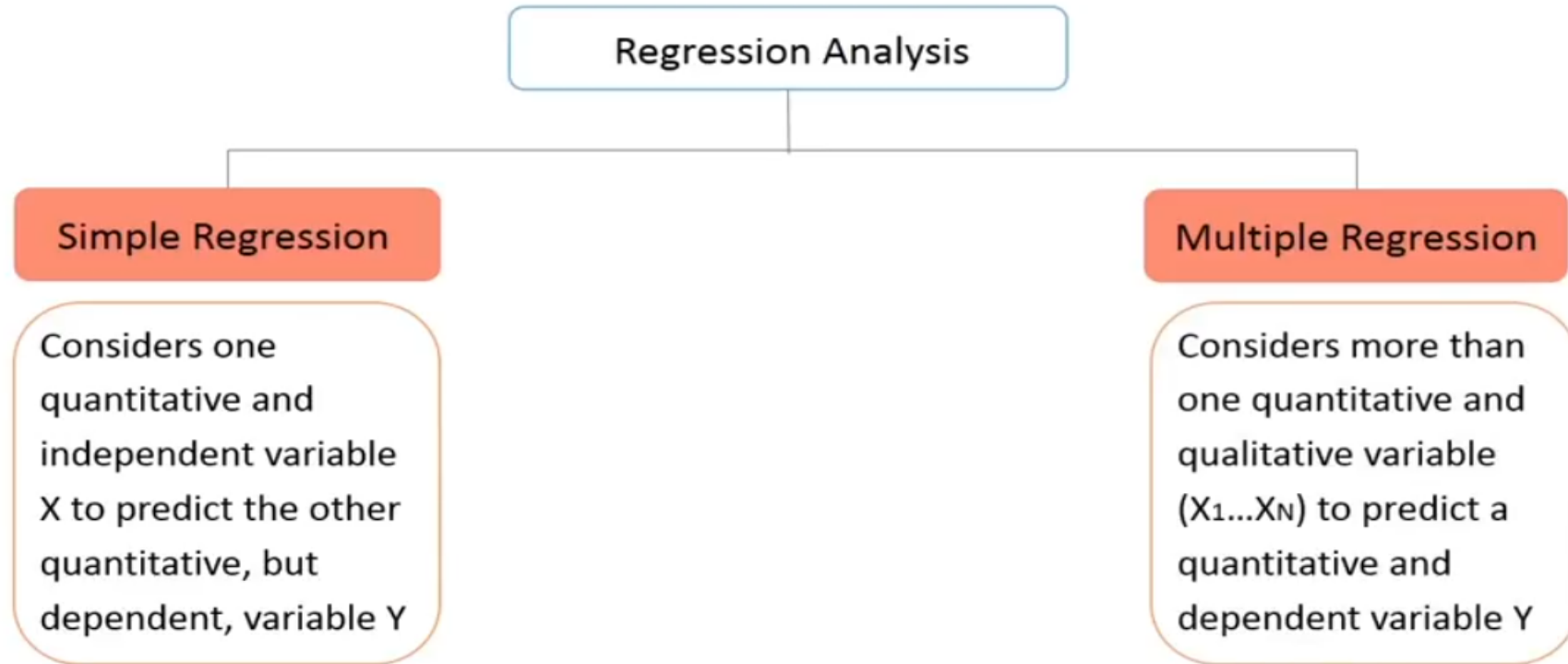
**Example:**

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

Here,  $Y$  is a dependent variable, whereas  $\beta_0$ ,  $\beta_1$ ,  $x$ , and  $\varepsilon$  are independent variables.

# Types Regression analysis

These are:



# Simple Regression analysis

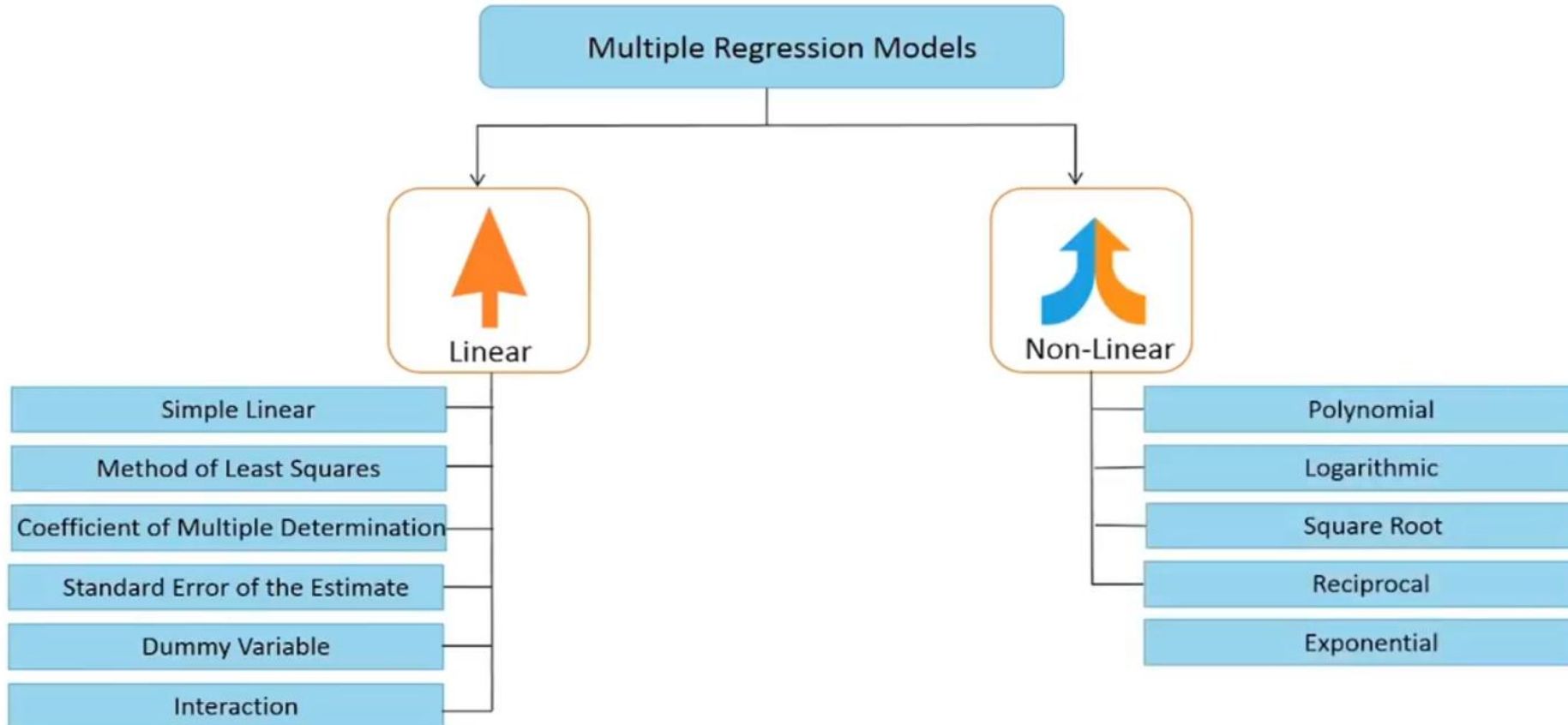
In this analysis:

- A relationship between a dependent variable and an independent variable is modeled
- The output is a function to predict the dependent variable on the basis of the values of independent variables
- A straight line is fit to the data



# Multiple Regression Models

These are:





Uses of  
regression  
analysis

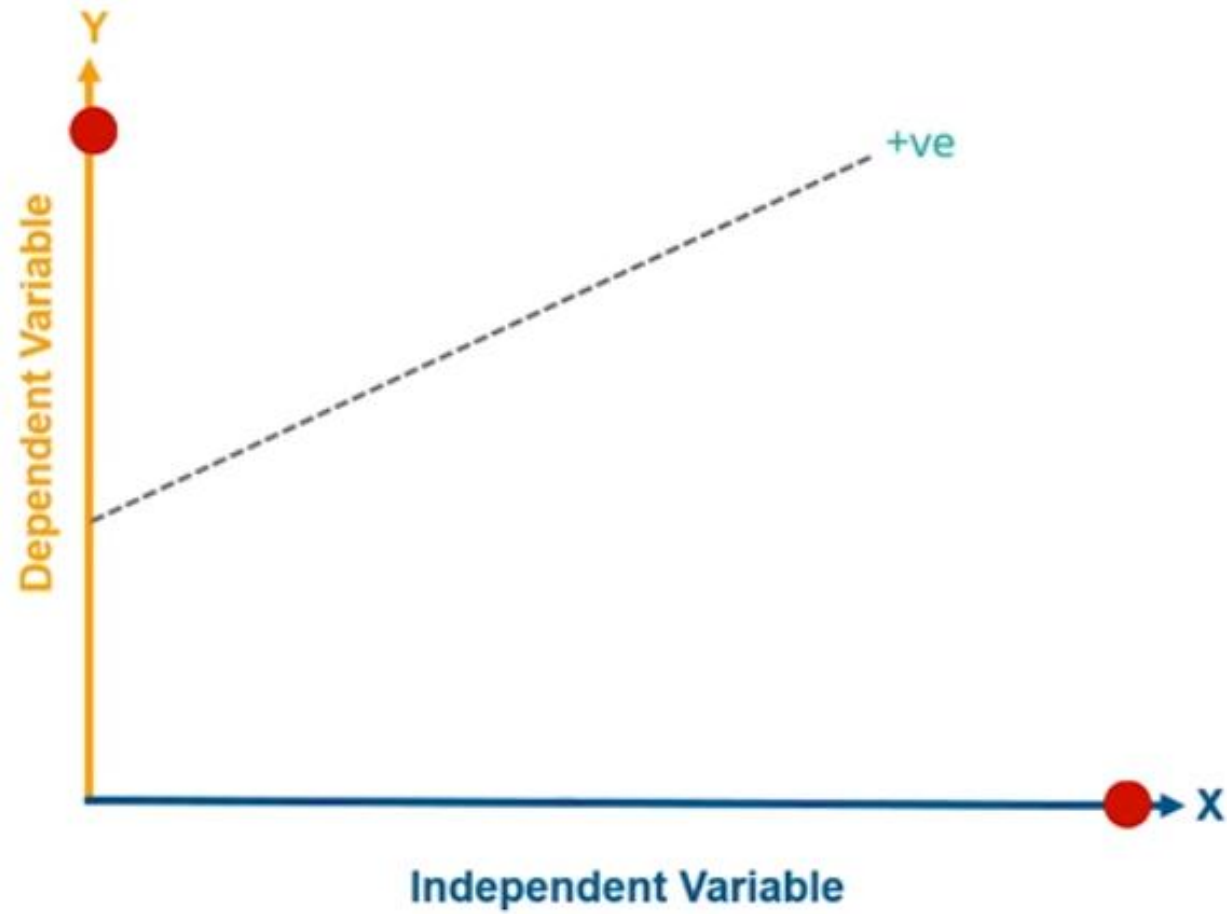
Three major uses for regression analysis are

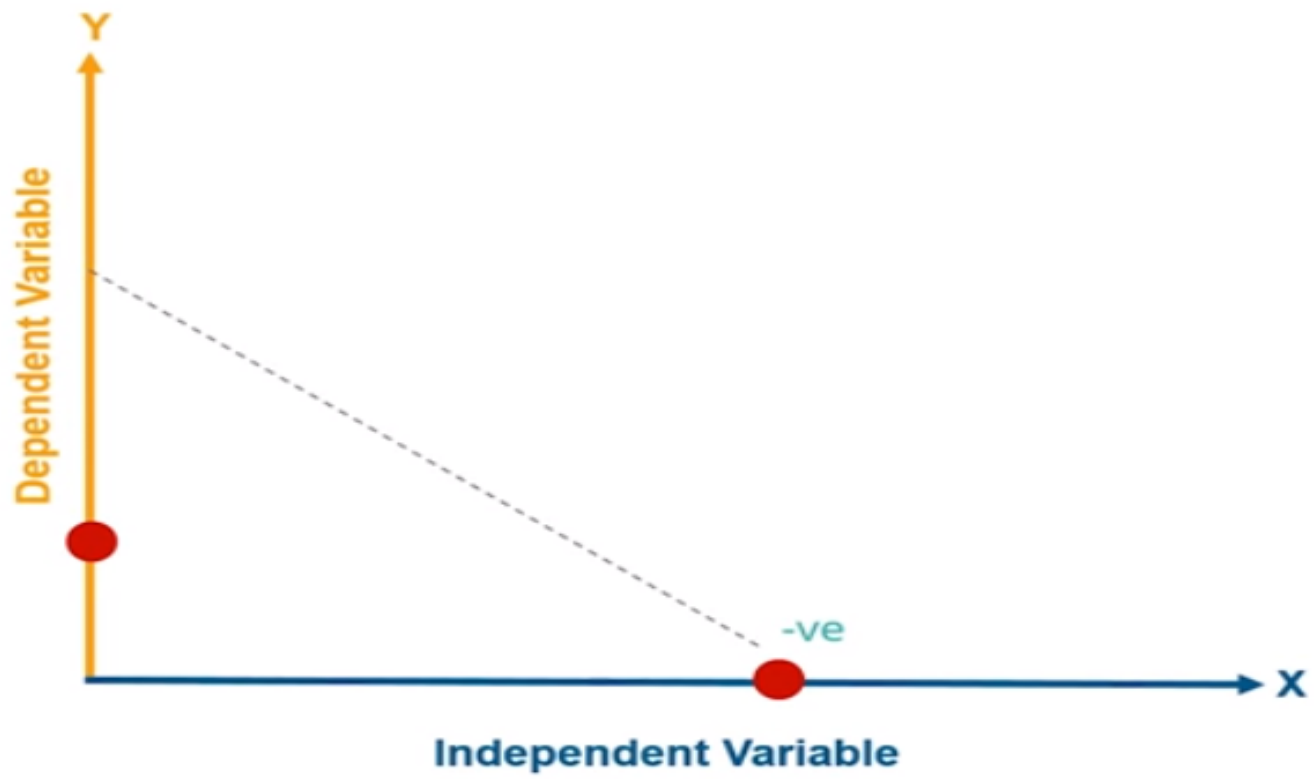
- Determining the strength of predictors
- Forecasting an effect, and
- Trend forecasting

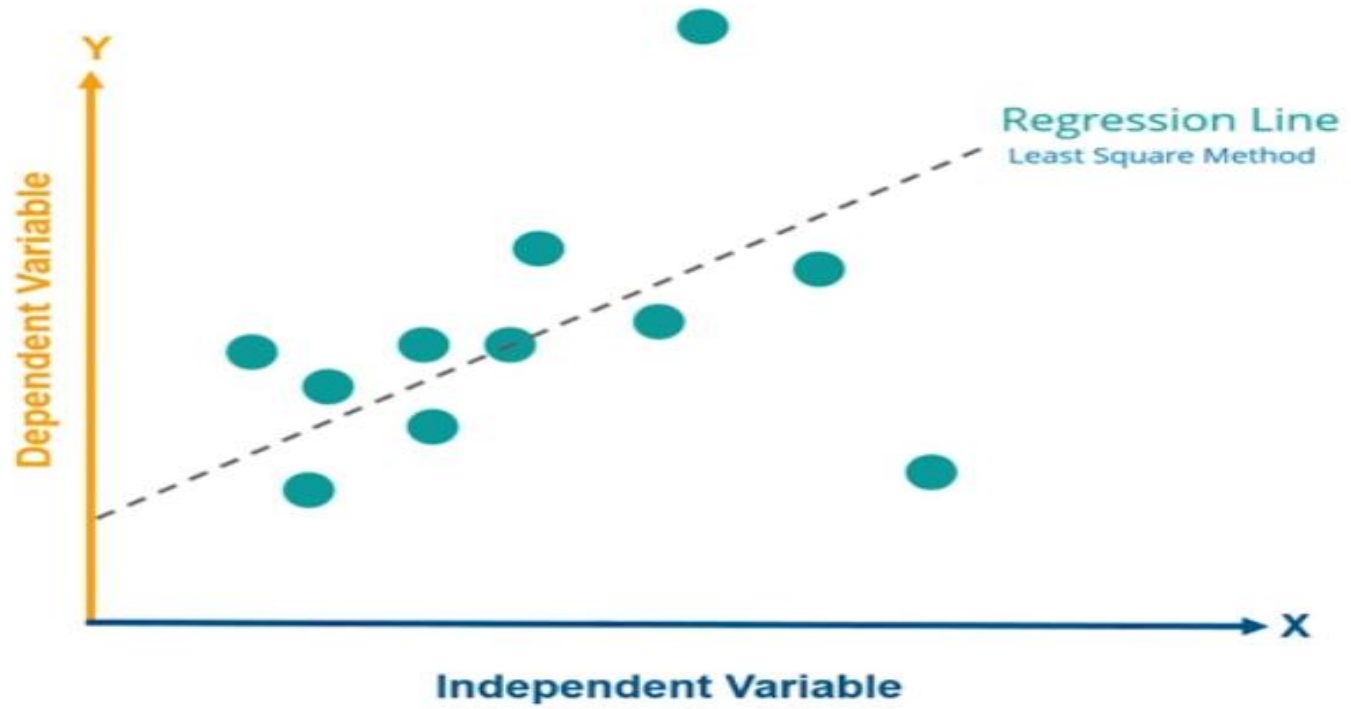
# Where is linear regression used

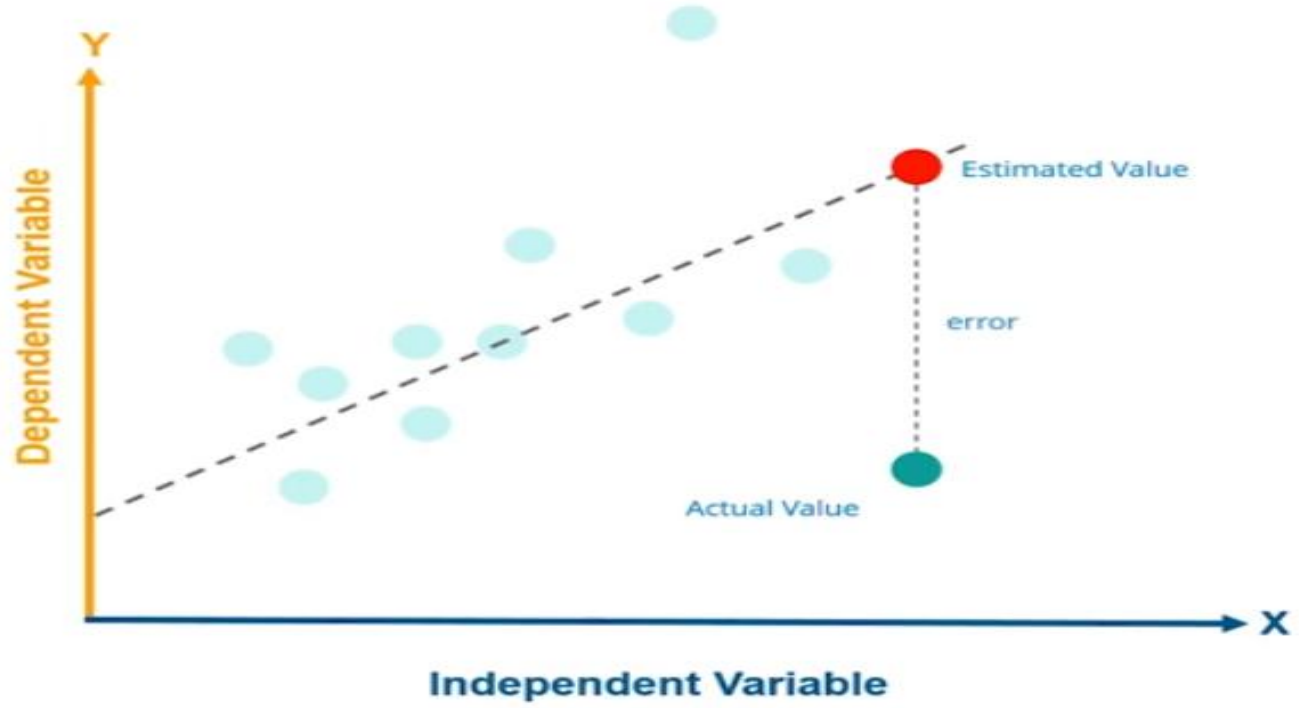
- Evaluating Trends and Sales Estimates
- Analyzing the Impact of Price Changes
- Assessment of risk in financial services and insurance domain

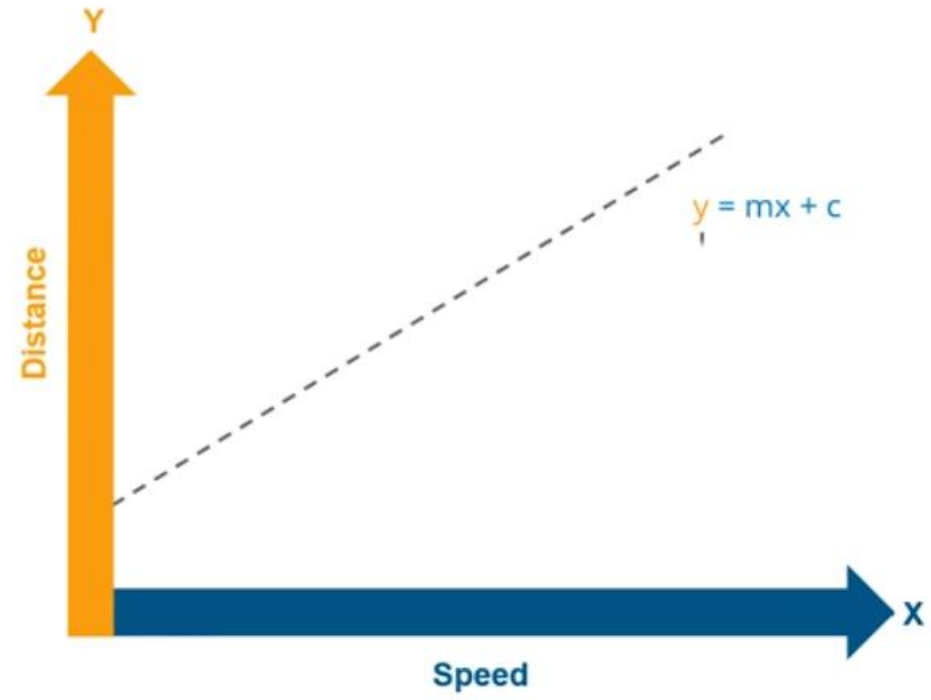
# Linear regression algorithm



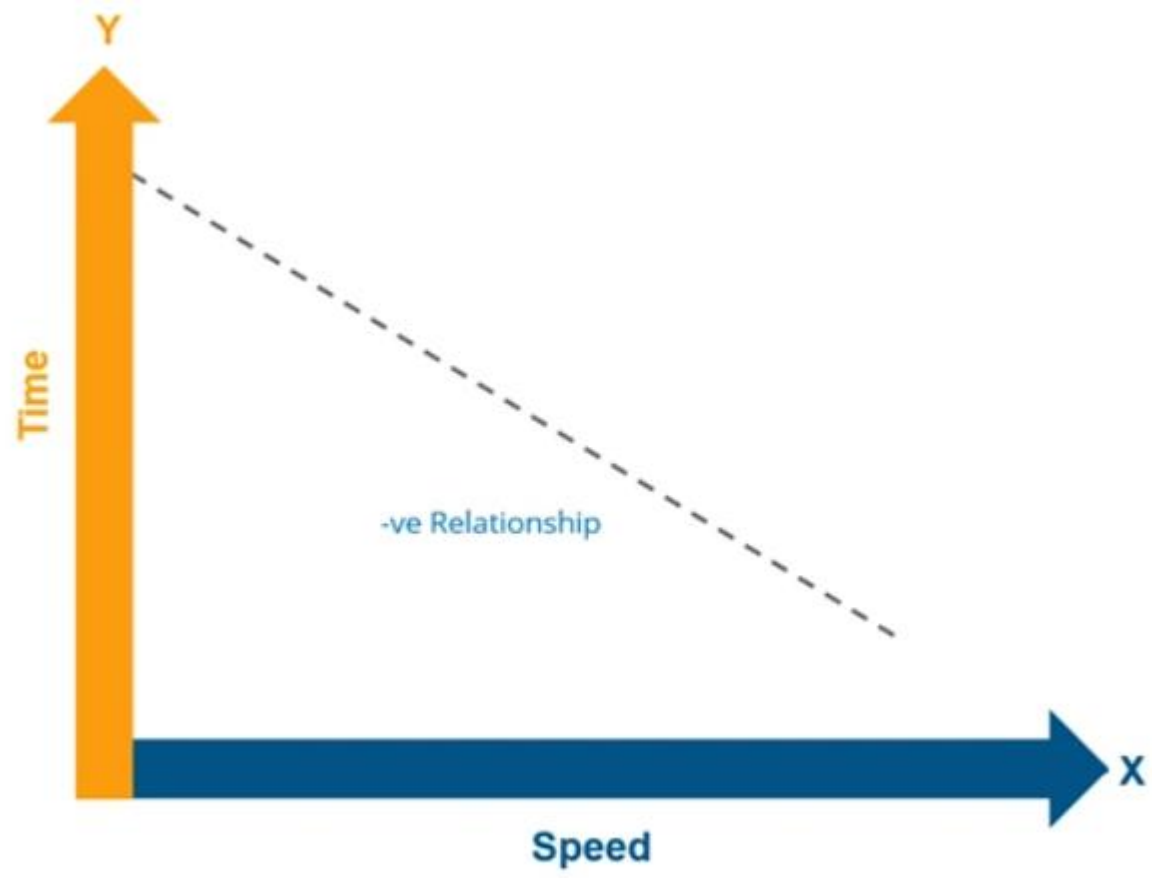


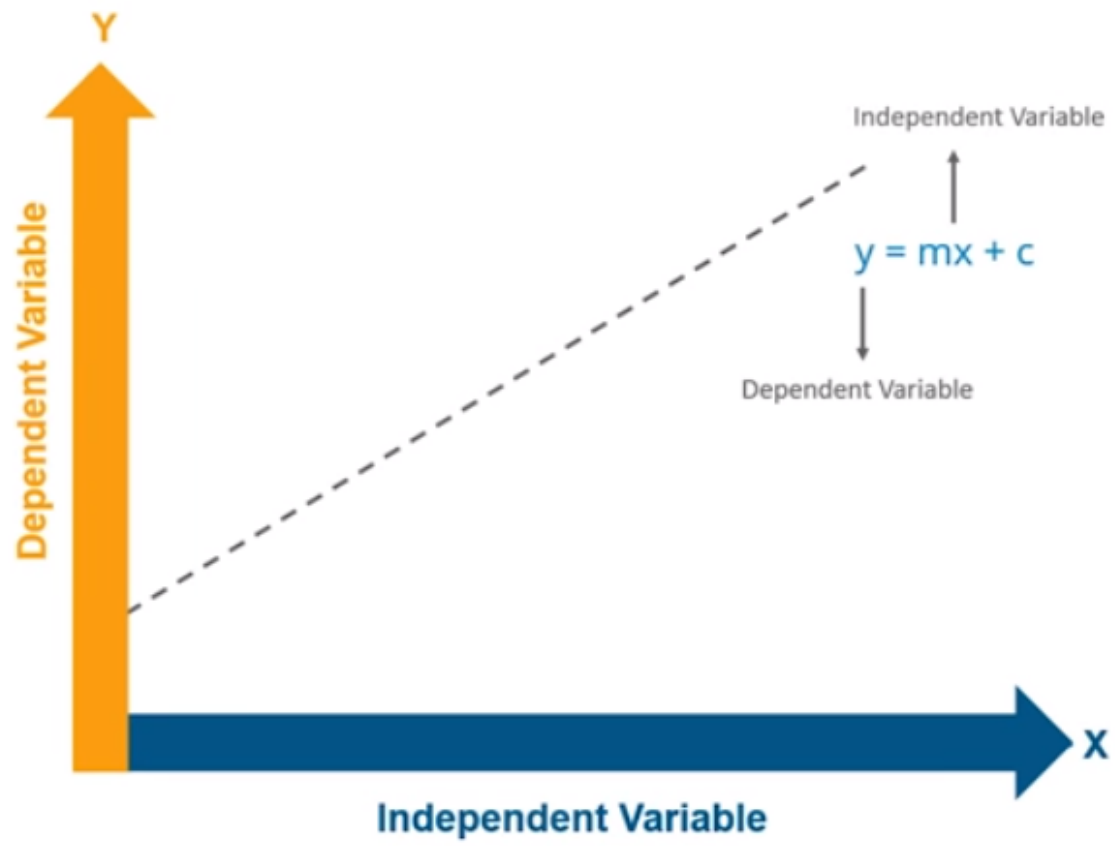


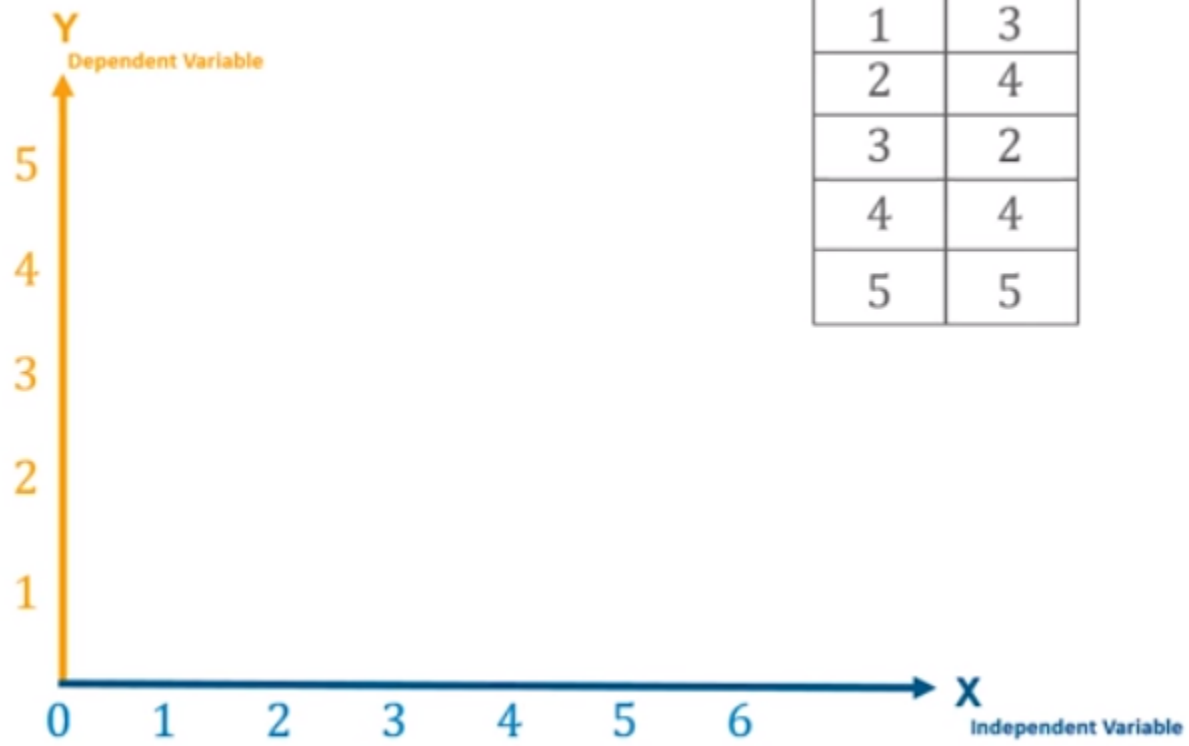


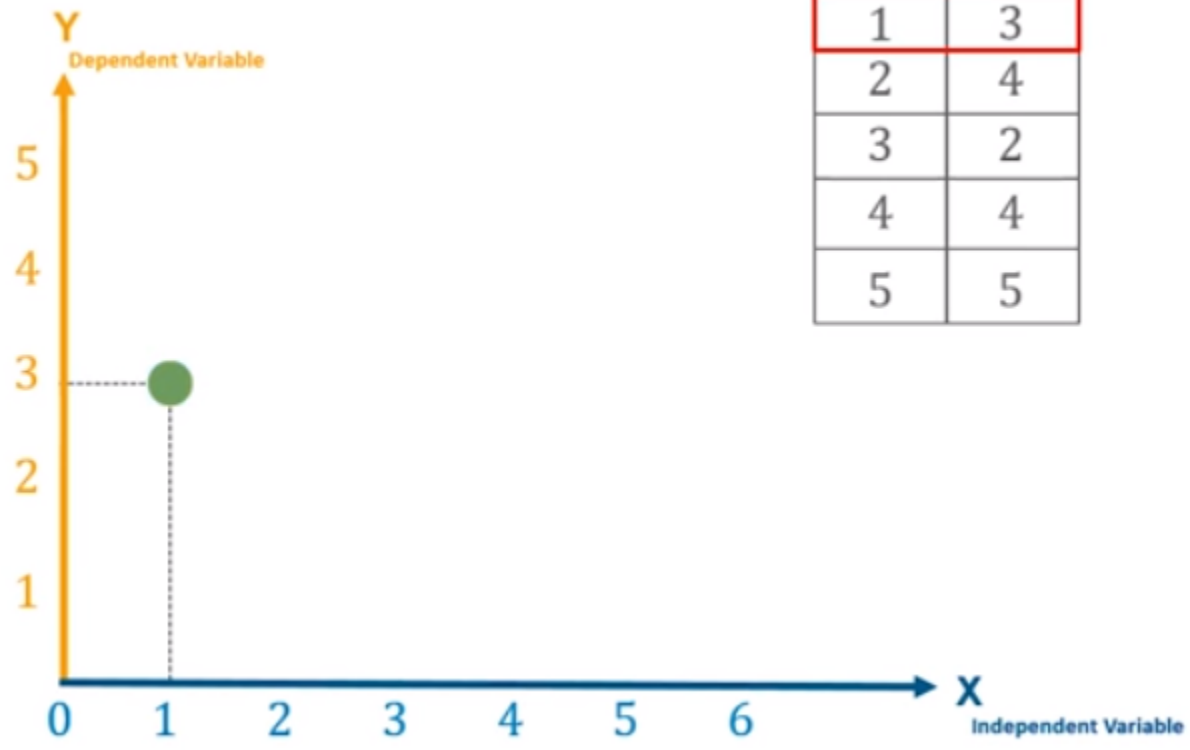




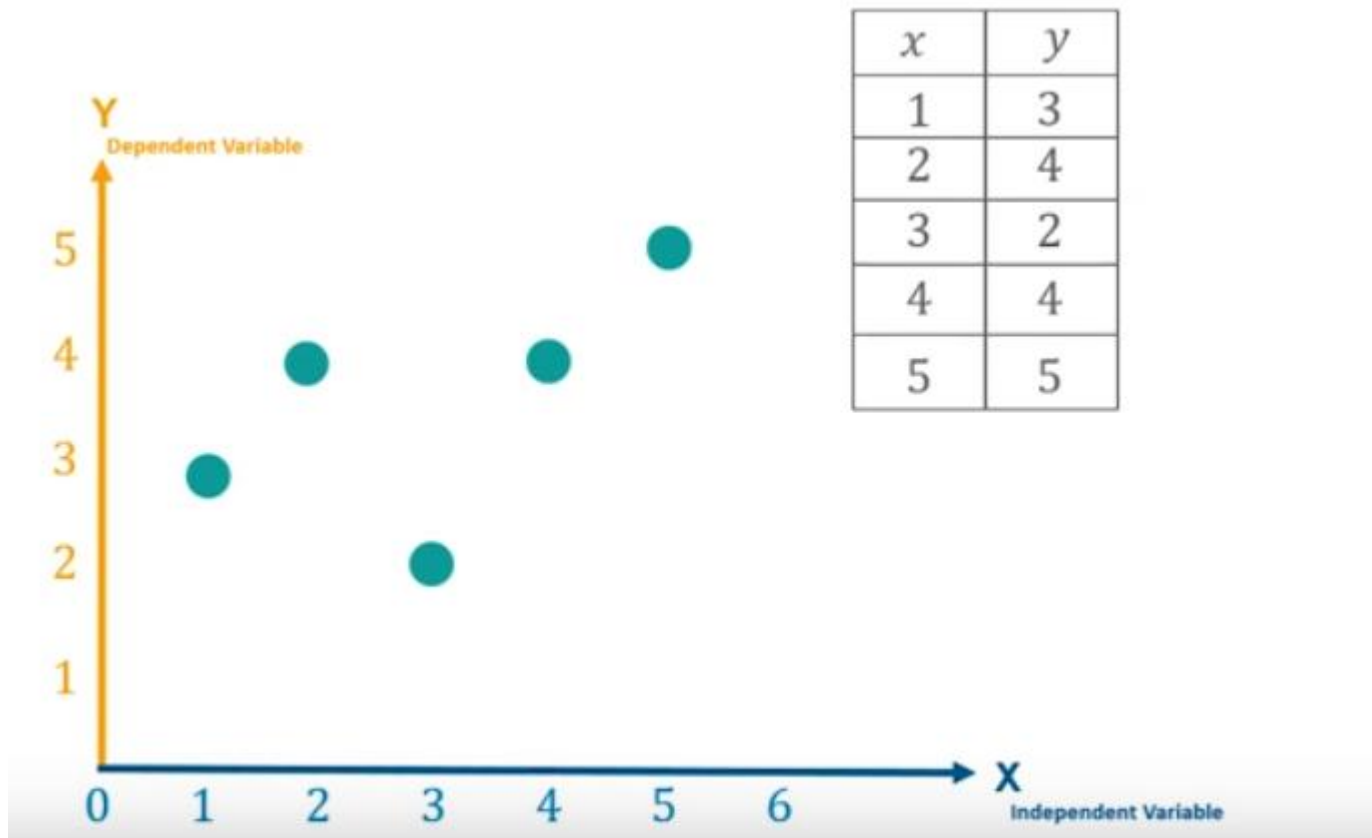


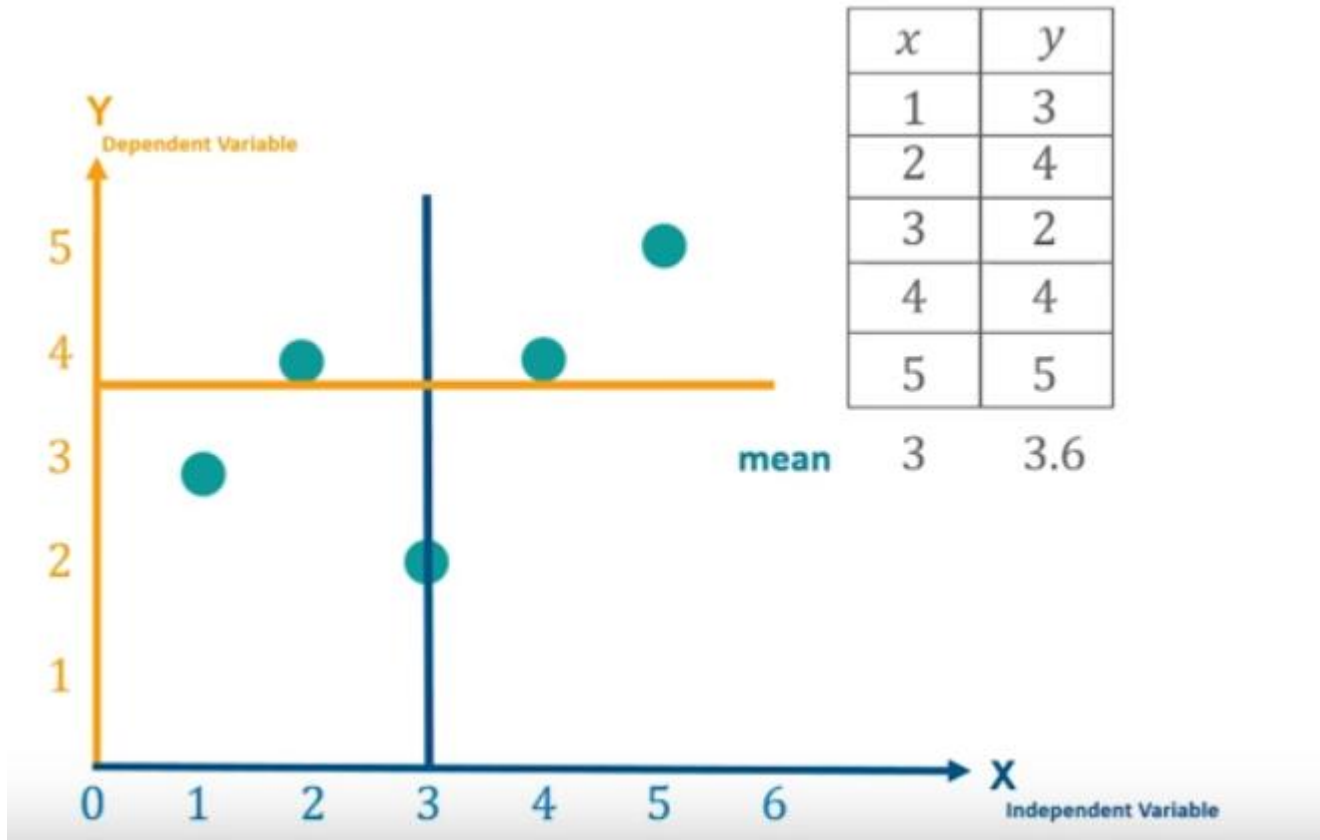


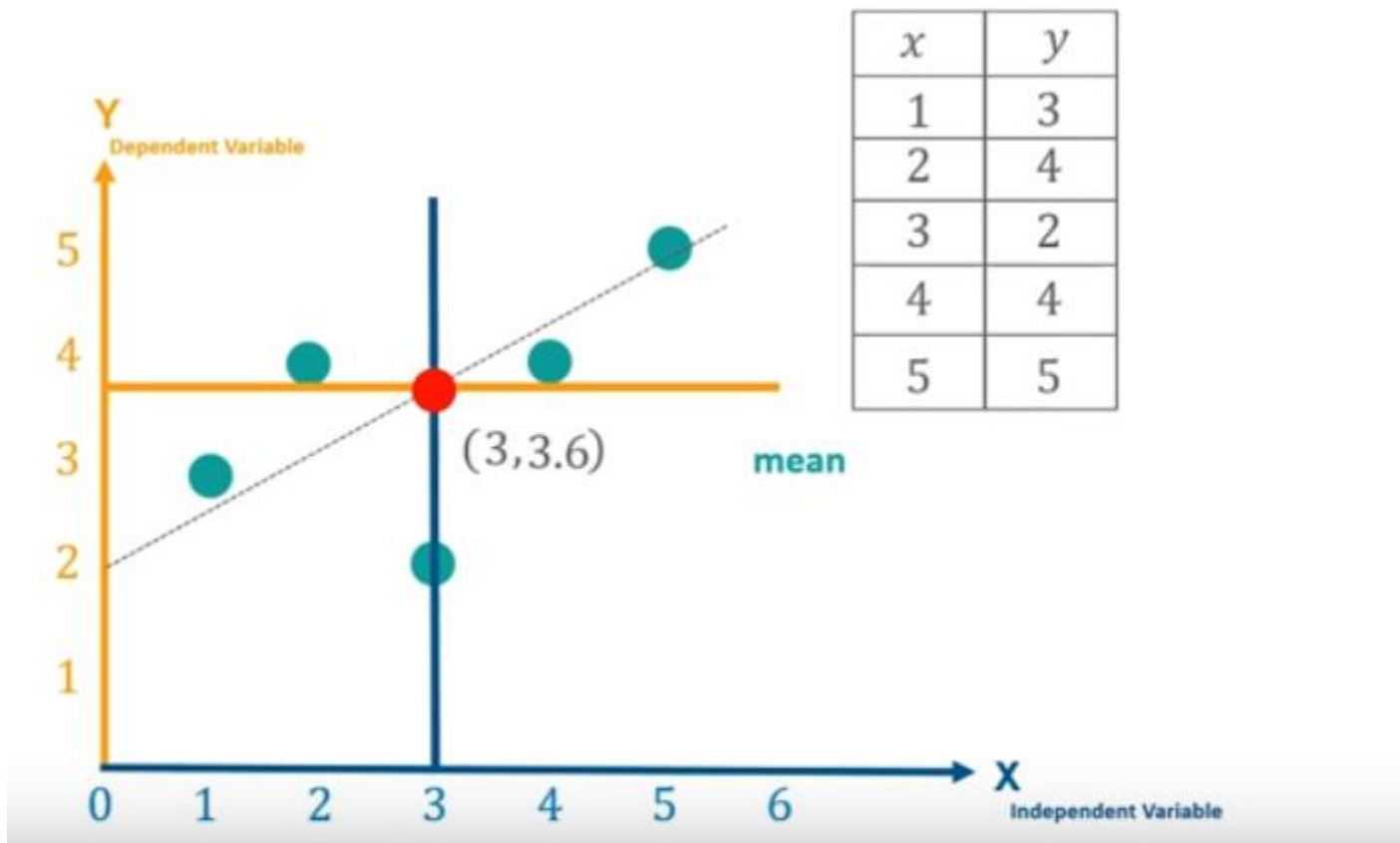


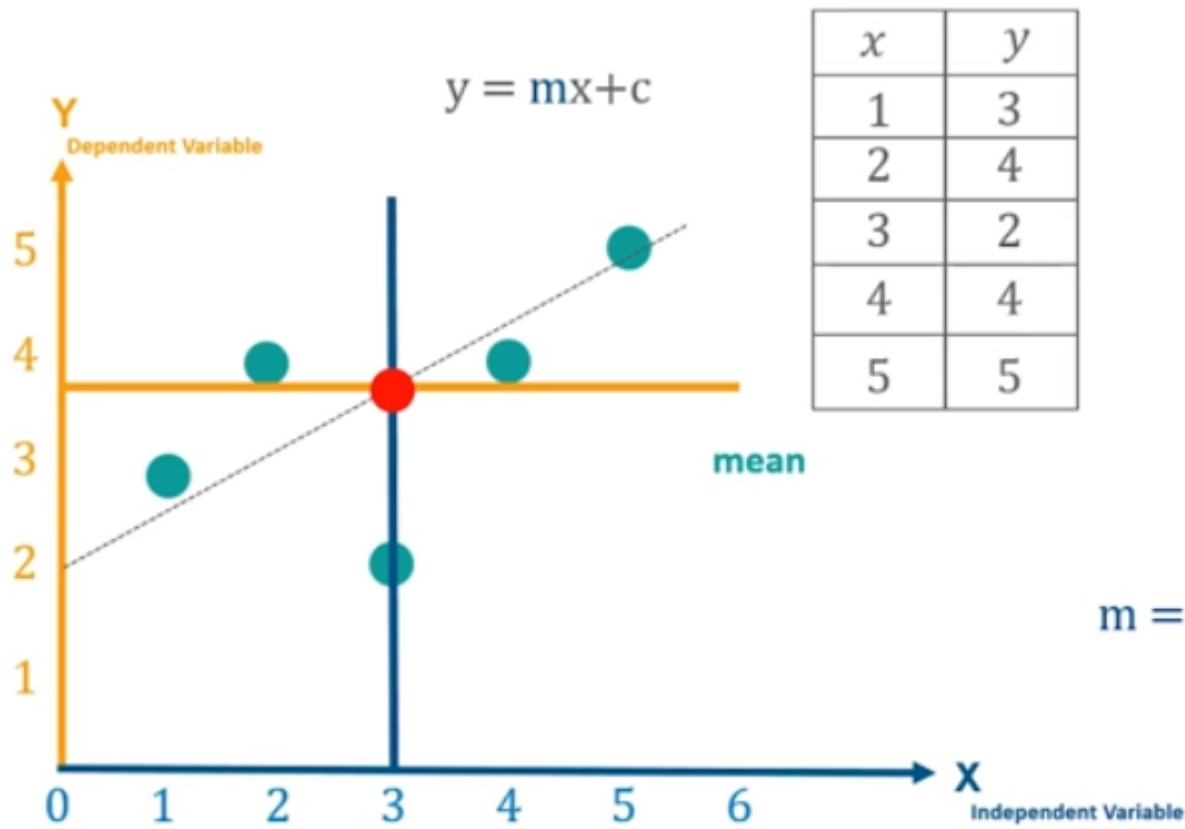


$x$	$y$
1	3
2	4
3	2
4	4
5	5



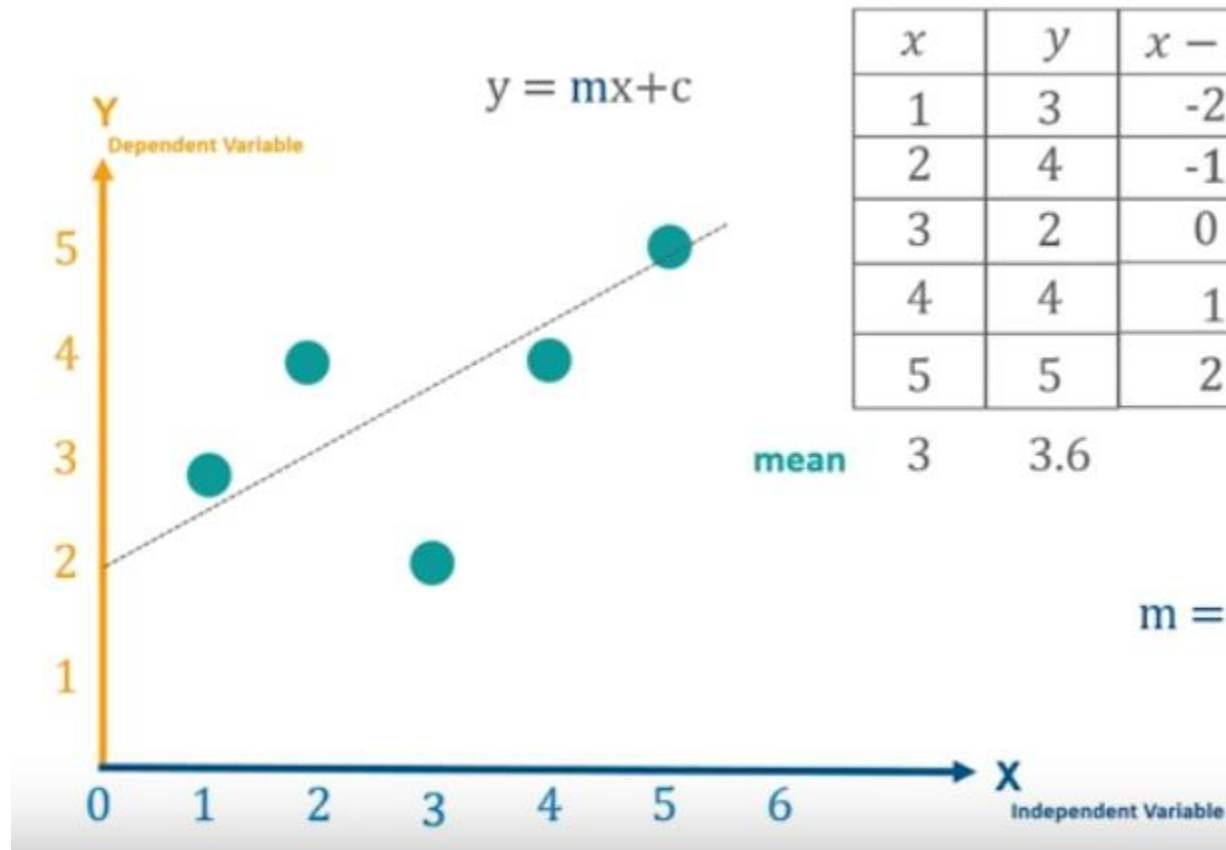






$$m = \frac{\sum (x - \bar{x})(y - \bar{y})}{\sum (x - \bar{x})^2}$$

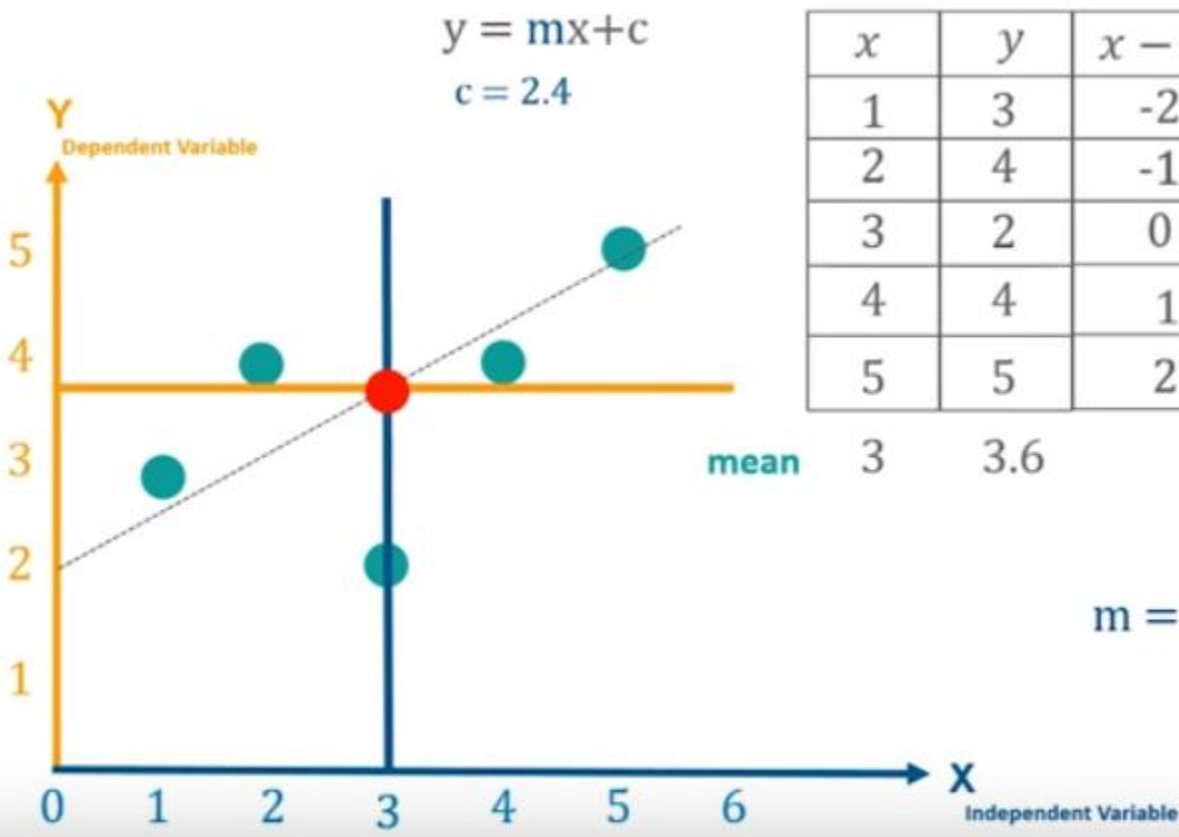




$x$	$y$	$x - \bar{x}$	$y - \bar{y}$	$(x - \bar{x})^2$	$(x - \bar{x})(y - \bar{y})$
1	3	-2	-0.6	4	1.2
2	4	-1	0.4	1	-0.4
3	2	0	-1.6	0	0
4	4	1	0.4	1	0.4
5	5	2	1.4	4	2.8
		$\Sigma = 10$	$\Sigma = 4$		

$$m = \frac{\Sigma (x - \bar{x})(y - \bar{y})}{\Sigma (x - \bar{x})^2} = \frac{4}{10}$$

- $Y = mx + c$
- $Y = 3.6, m = 0.4, x = 3$
- $3.6 = 0.4 * 3 + c$
- $3.6 - 1.2 = c$
- $C = 2.4$

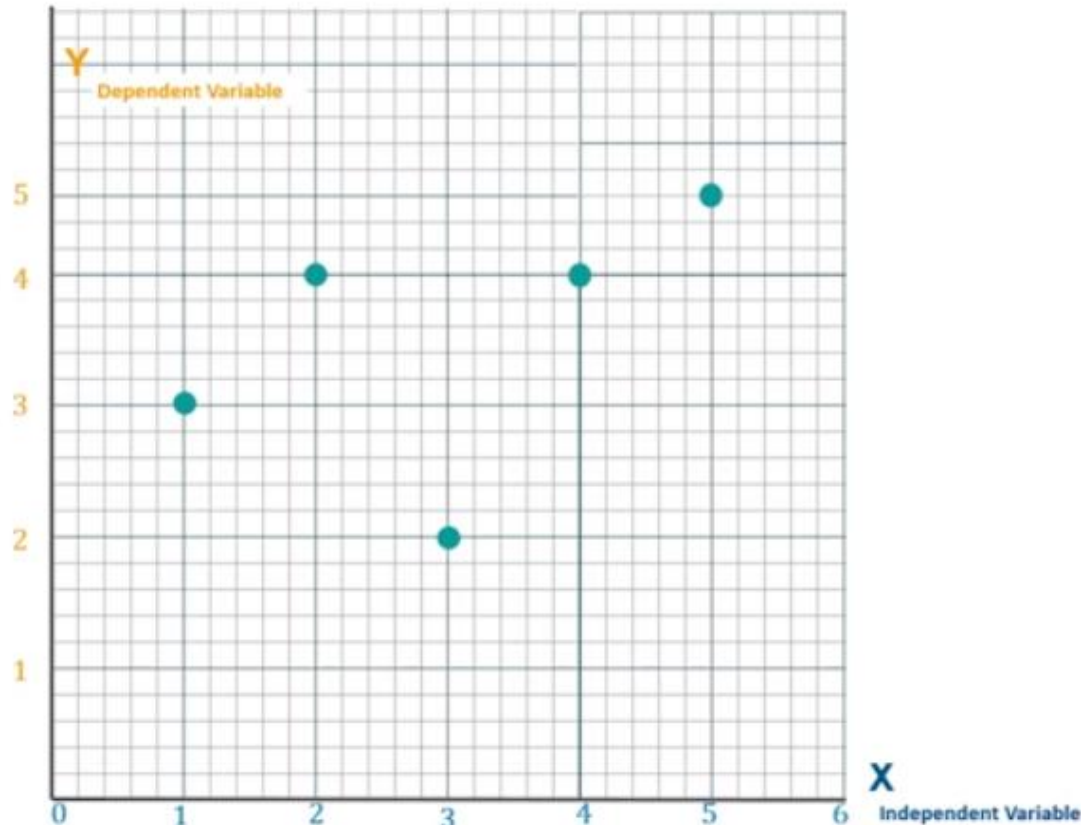


$x$	$y$	$x - \bar{x}$	$y - \bar{y}$	$(x - \bar{x})^2$	$(x - \bar{x})(y - \bar{y})$
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$\Sigma$	10			$\Sigma = 10$	$\Sigma = 4$

$$m = \frac{\sum (x - \bar{x})(y - \bar{y})}{\sum (x - \bar{x})^2} = \frac{4}{10}$$

$m = 0.4$   
 $c = 2.4$   
 $y = 0.4x + 2.4$

# Mean Square Error



$$m = 0.4$$
$$c = 2.4$$
$$y = 0.4x + 2.4$$

For given  $m = 0.4$  &  $c = 2.4$ , lets predict values for y for  $x = \{1,2,3,4,5\}$

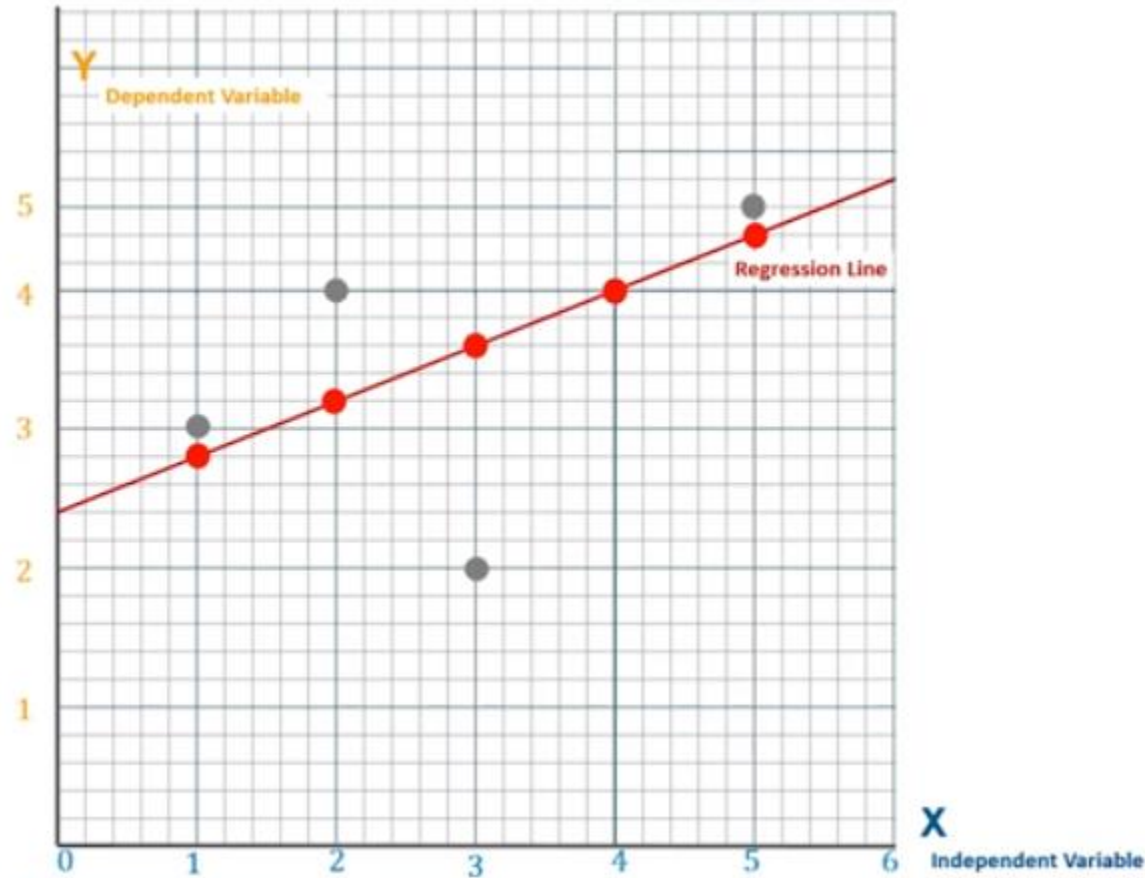
$$y = 0.4 \times 1 + 2.4 = 2.8$$

$$y = 0.4 \times 2 + 2.4 = 3.2$$

$$y = 0.4 \times 3 + 2.4 = 3.6$$

$$y = 0.4 \times 4 + 2.4 = 4.0$$

$$y = 0.4 \times 5 + 2.4 = 4.4$$



$$m = 0.4$$
$$c = 2.4$$
$$y = 0.4x + 2.4$$

For given  $m = 0.4$  &  $c = 2.4$ , lets  
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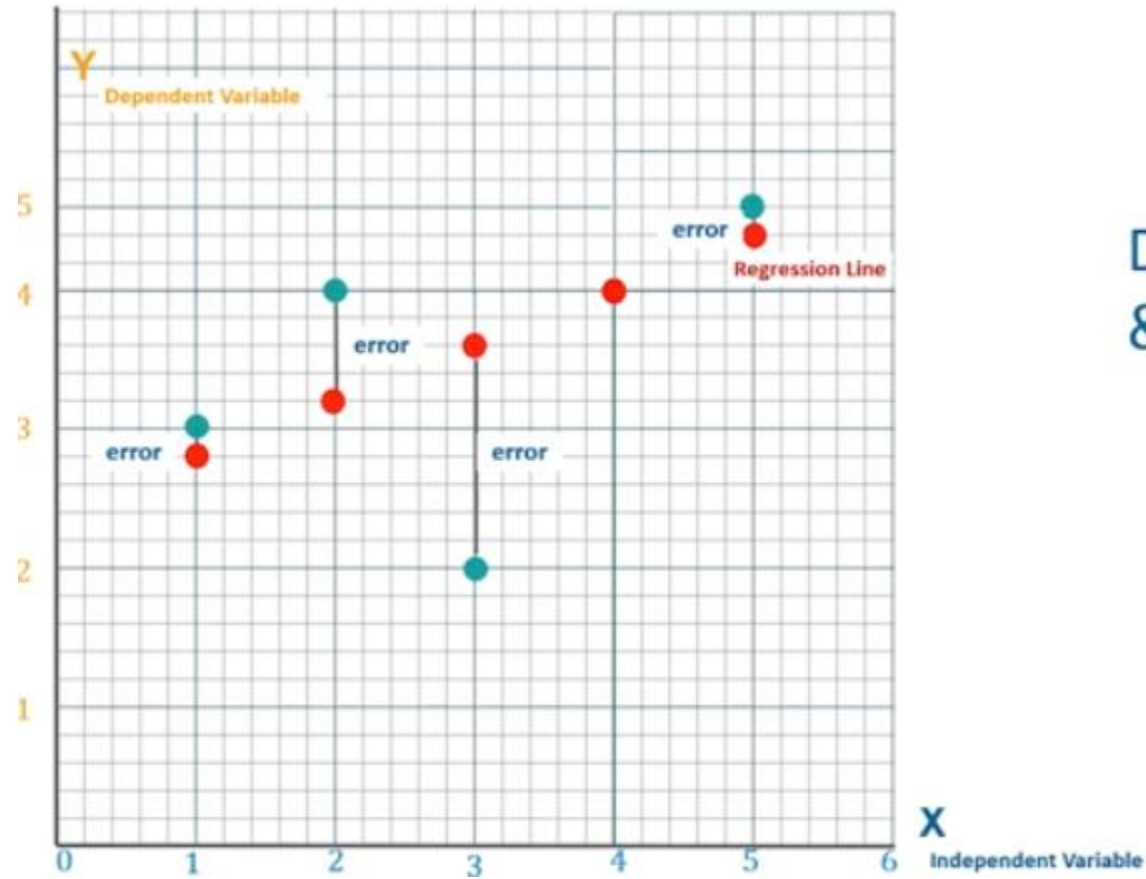
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$$y = 0.4 \times 3 + 2.4 = 3.6$$

$$y = 0.4 \times 4 + 2.4 = 4.0$$

$$y = 0.4 \times 5 + 2.4 = 4.4$$



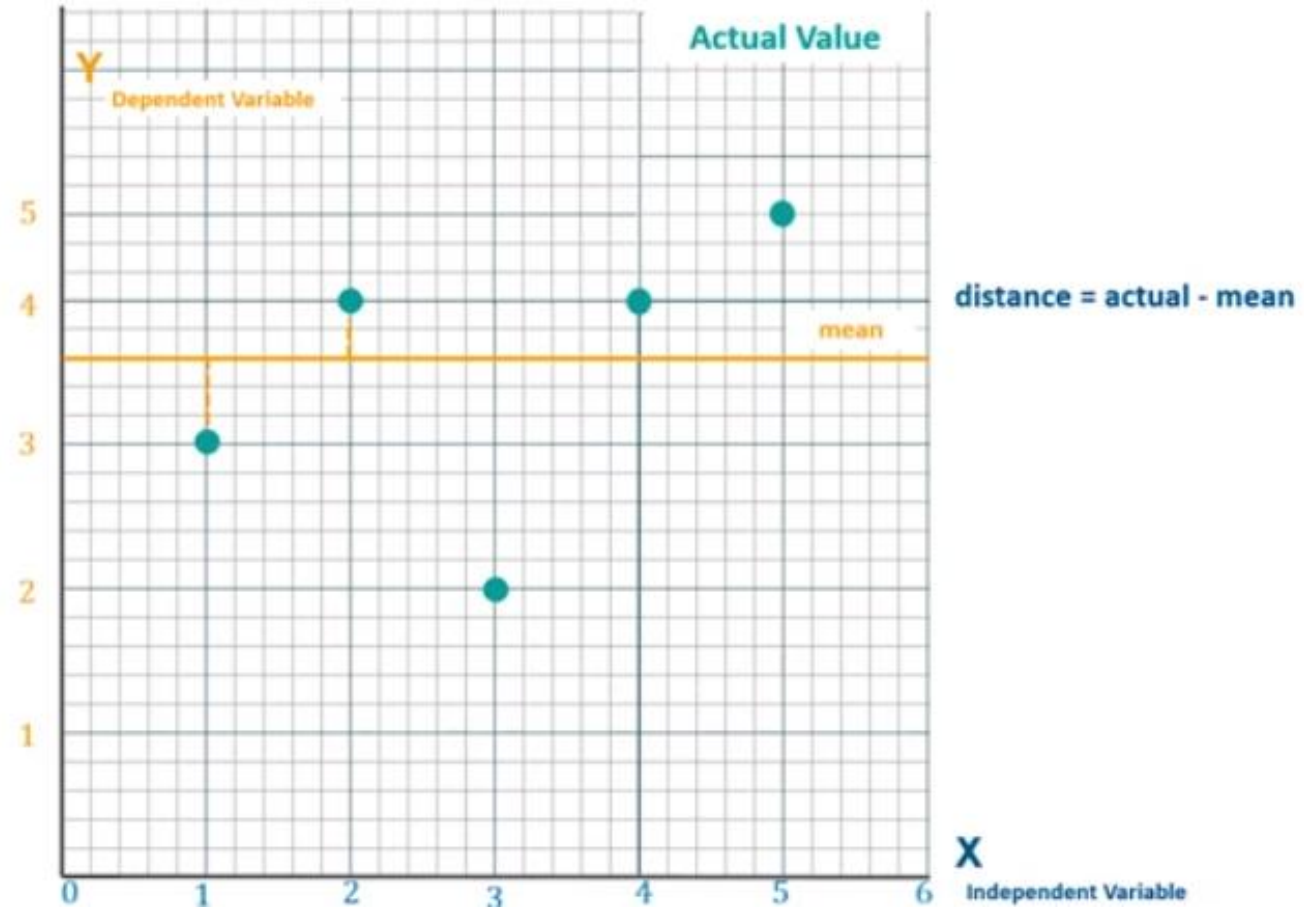
Distance between actual & predicted value

# R Square:-

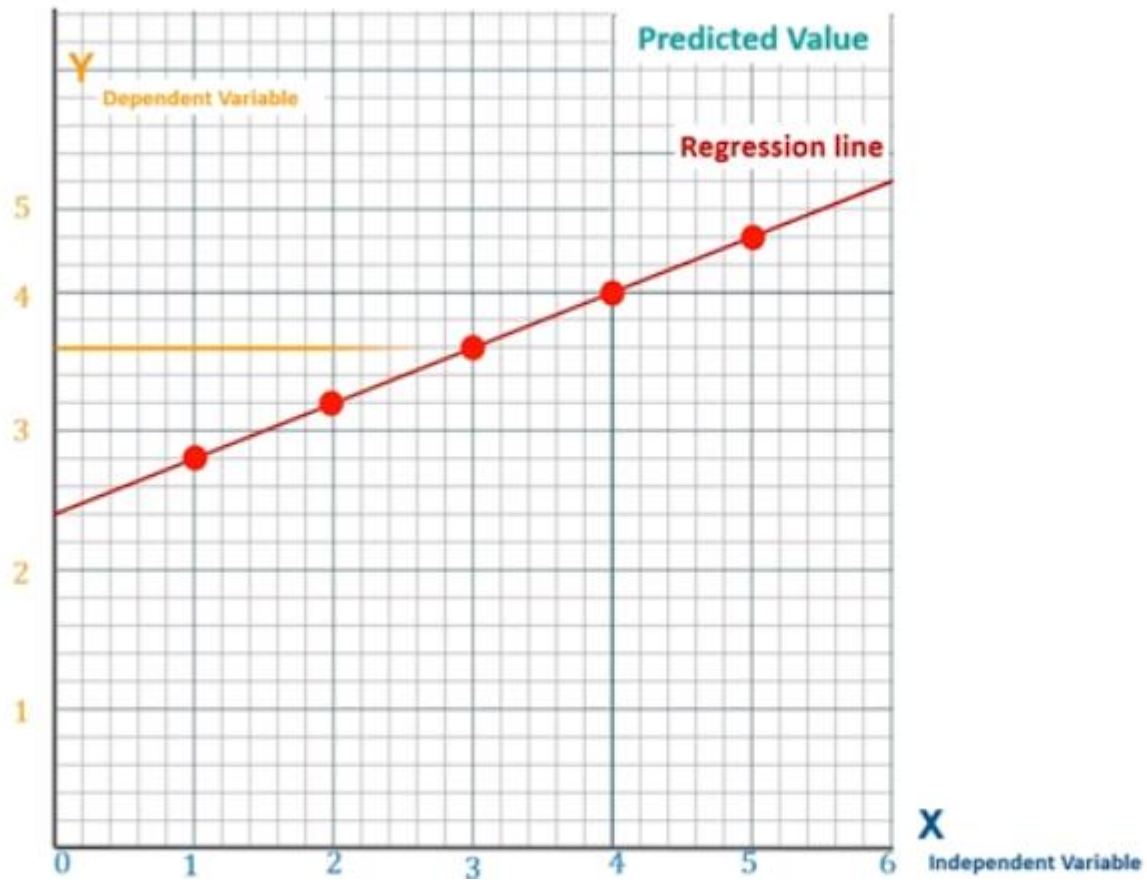
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- **R-squared** value is a statistical measure of how close the data are to the fitted regression line
- It is also known as **coefficient of determination**, or the **coefficient of multiple determination**

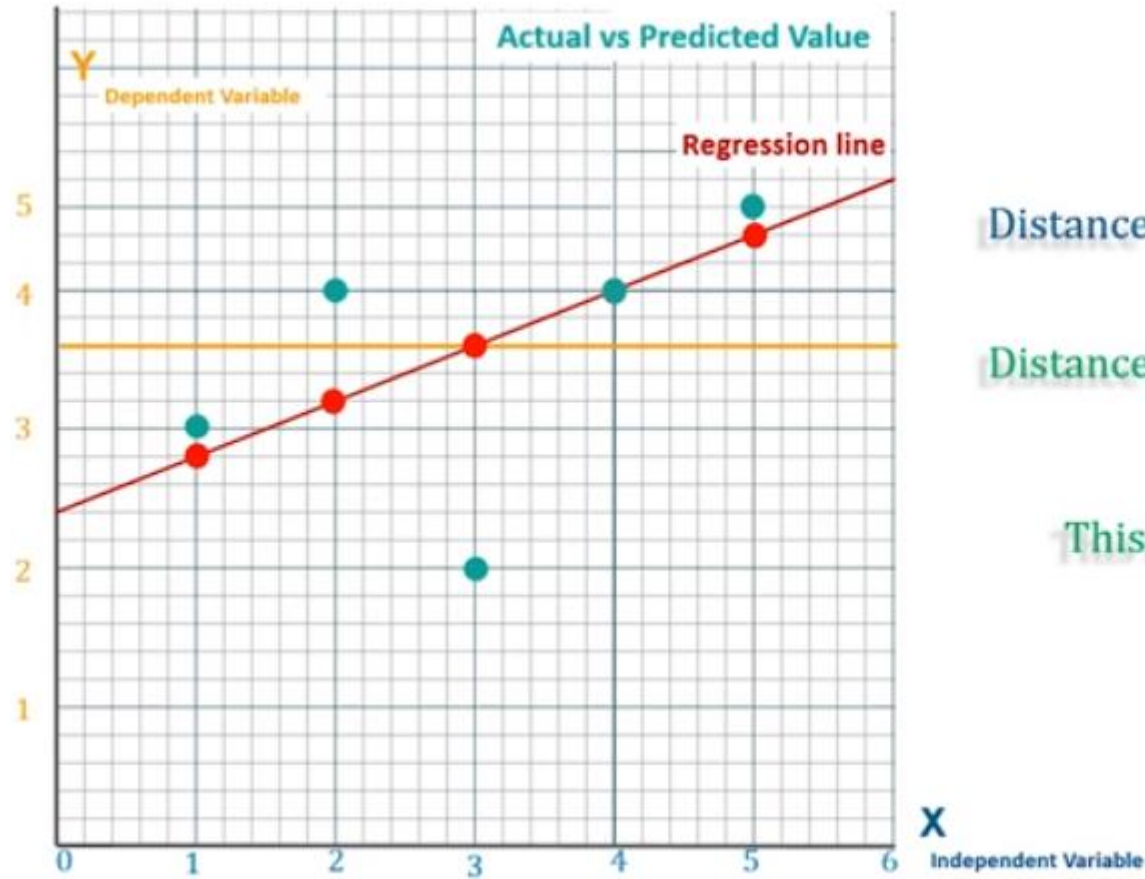
# Calculation of R square







$x$	$y_p$
1	2.8
2	3.2
3	3.6
4	4.0
5	4.4

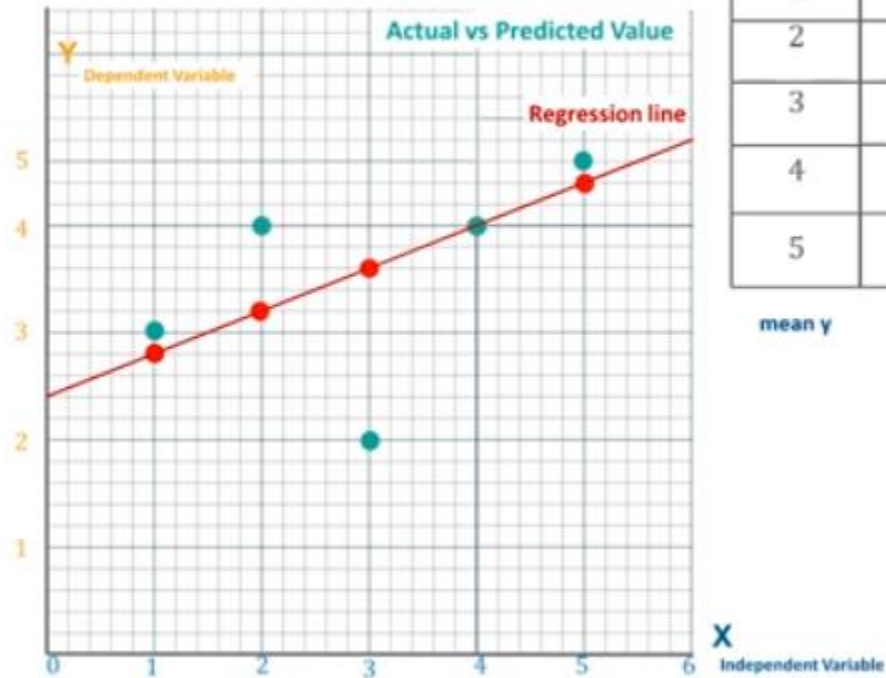


Distance actual - mean

vs

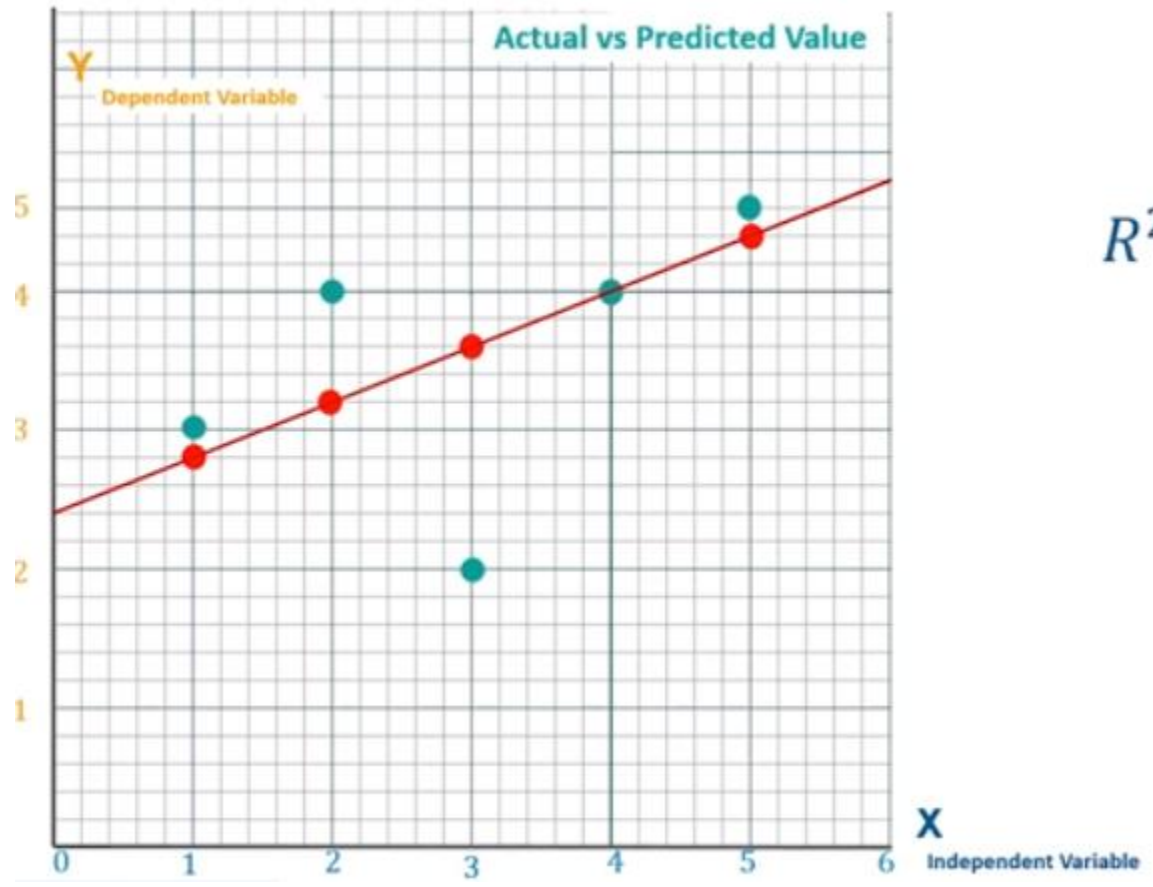
Distance predicted - mean

This is nothing but  $R^2 = \frac{\sum (y_p - \bar{y})^2}{\sum (y - \bar{y})^2}$



x	y	$y - \bar{y}$	$(y - \bar{y})^2$	$y_p$	$(y_p - \bar{y})$	$(y_p - \bar{y})^2$
1	3	-0.6	0.36	2.8	-0.8	0.64
2	4	0.4	0.16	3.2	-0.4	0.16
3	2	-1.6	2.56	3.6	0	0
4	4	0.4	0.16	4.0	0.4	0.16
5	5	1.4	1.96	4.4	0.8	0.64
mean y	3.6		$\Sigma$ 5.2			$\Sigma$ 1.6

$$R^2 = \frac{1.6}{5.2} = \frac{\Sigma (y_p - \bar{y})^2}{\Sigma (y - \bar{y})^2}$$



$$R^2 \approx 0.3$$