

# Linear Regression

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# Introduction to Machine Learning

## An Overview

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## Basics

# Independent and Dependent Variable

- **Independent Variable** - A variable whose value does not change by the effect of other variables and is used to manipulate the dependent variable. It is often denoted as X.
- **Dependent variable** - A variable whose value change when there is any manipulation in the values of independent variables. It is often denoted as Y.

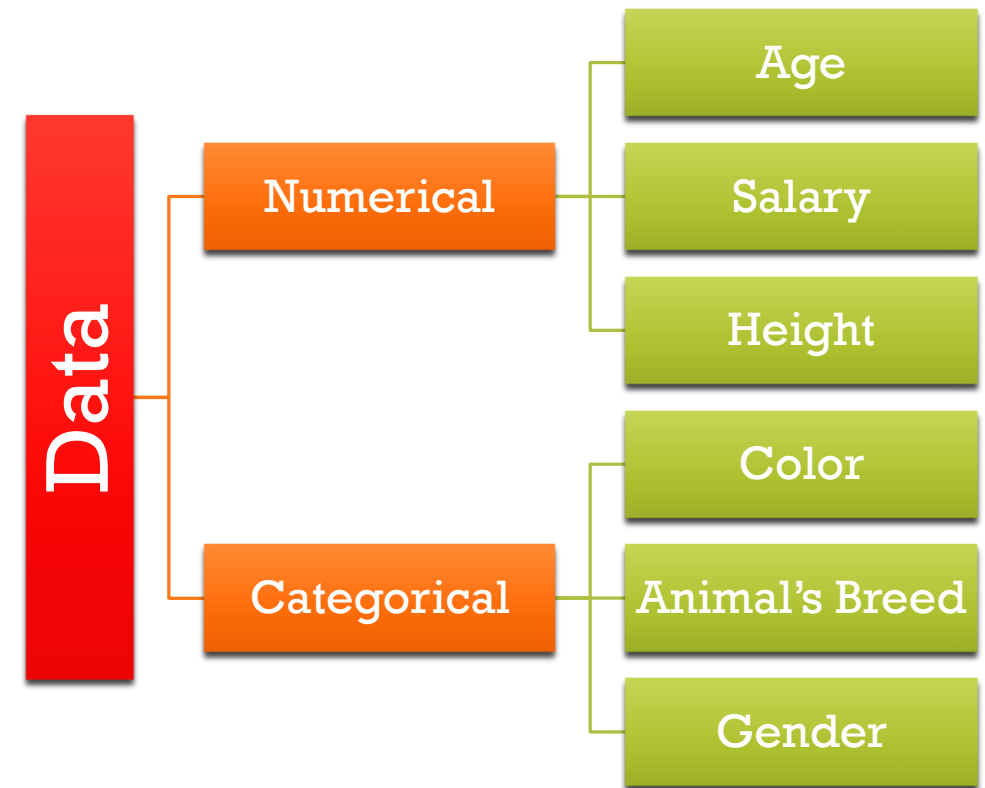
Based on amount of rainfall, How much would be the crop yield?

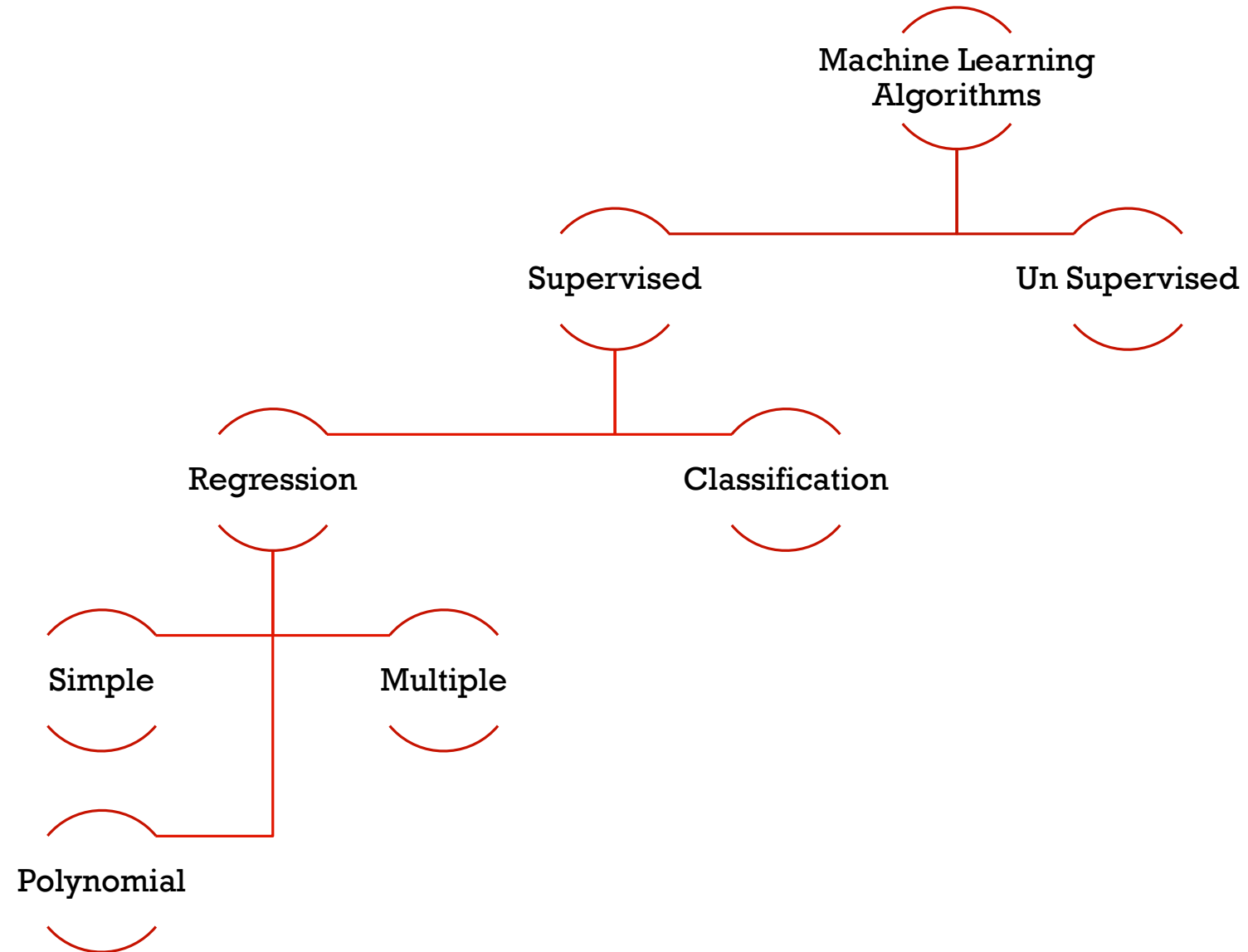


Crop Yielded depends upon the  
amount of rainfall received



# Numerical and Categorical values

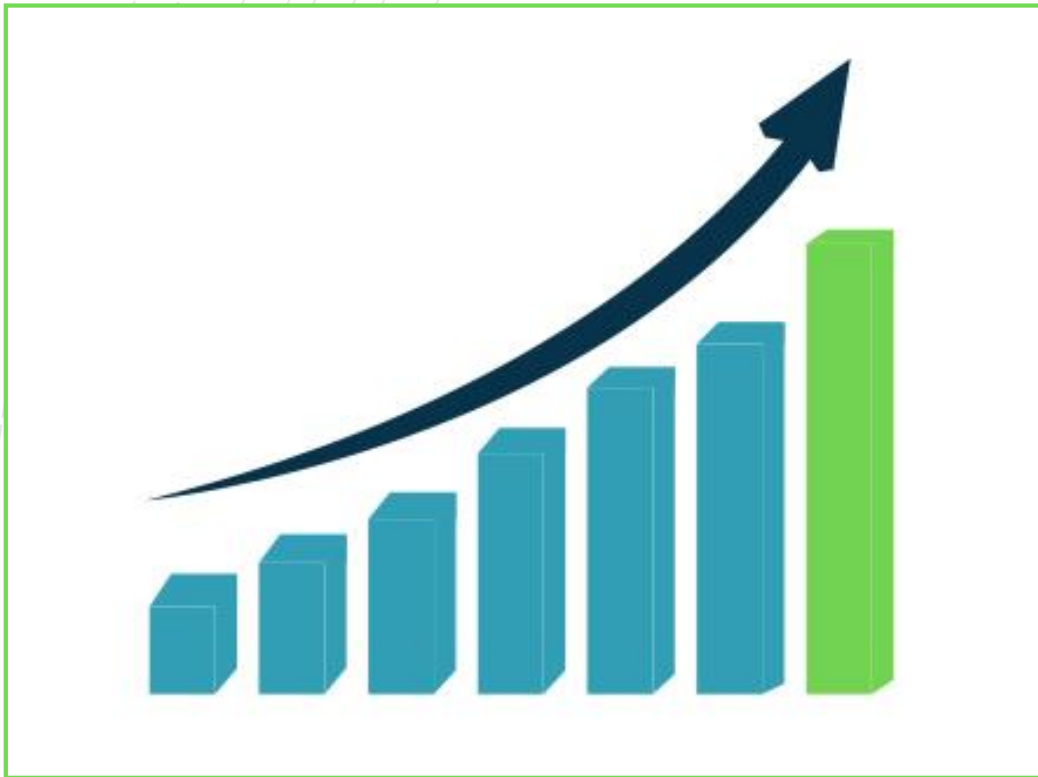




# Application of Linear Regression



# Economic Growth



- Used to determine the Economic growth of a country or a state in the coming quarter, can also be used to predict the GDP of a country.

# Product Price



- Can be used to predict what would be the price of the product in the future.

# Housing Sales



- To estimate the number of houses a builder would sell and at what price in the coming months.

# Score Prediction



- To predict the number of runs a player would score in the coming matches based on previous performance.

# Understanding Linear Regression

Linear Regression is a statistical model used to predict the relationship between independent and dependent variables.

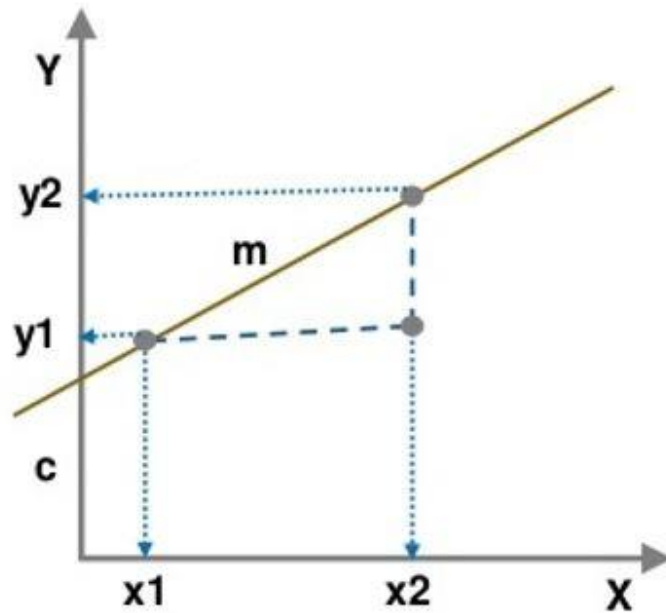
## Examine 2 Factors

Which variables in particular are significant predictors of the outcome variables?

How significant the Regression line to make predictions with highest possible accuracy.

# Regression Equation

The simplest form of a simple linear regression equation with one dependent and one independent variable is represented by:



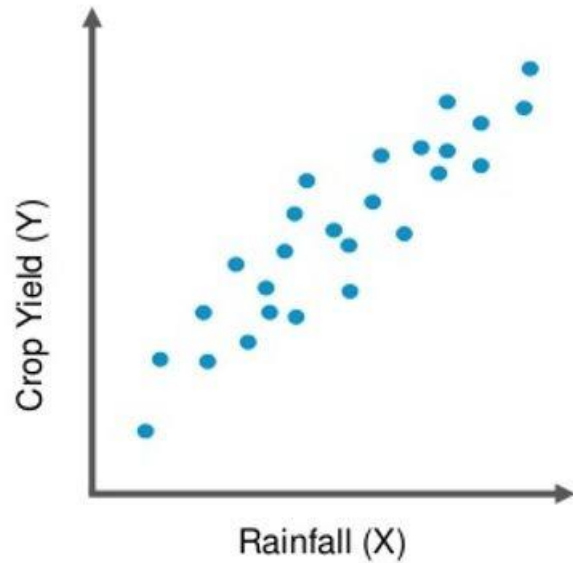
$$y = mx + c$$

- $y \rightarrow$  Dependent Variables
- $x \rightarrow$  Independent Variables
- $m \rightarrow$  Slope

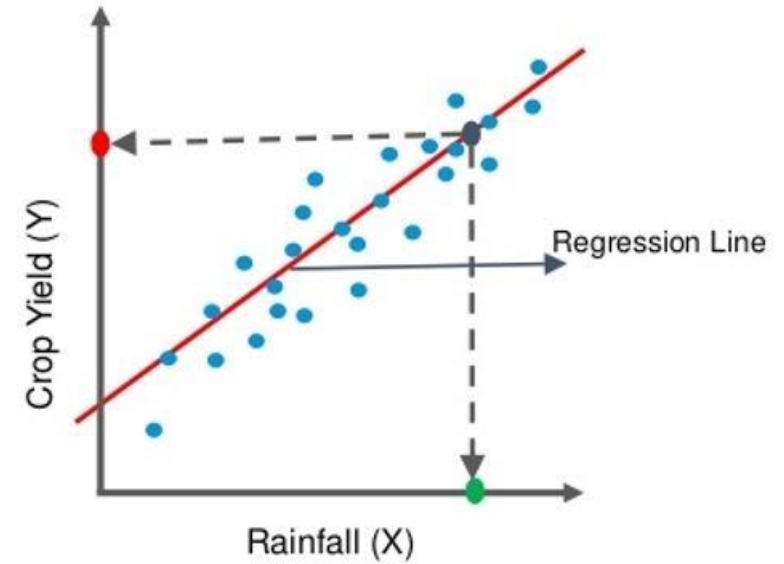
$$\frac{y_2 - y_1}{x_2 - x_1}$$

- $C \rightarrow$  Coefficient of the Line

# Prediction Using the Regression Line.



Plotting the amount of Crop Yield based on the amount of Rainfall



The Red point on the y axis is the amount of Crop Yield you can expect for some amount of Rainfall (X) represented by Green dot.



# Intuition behind the regression Line

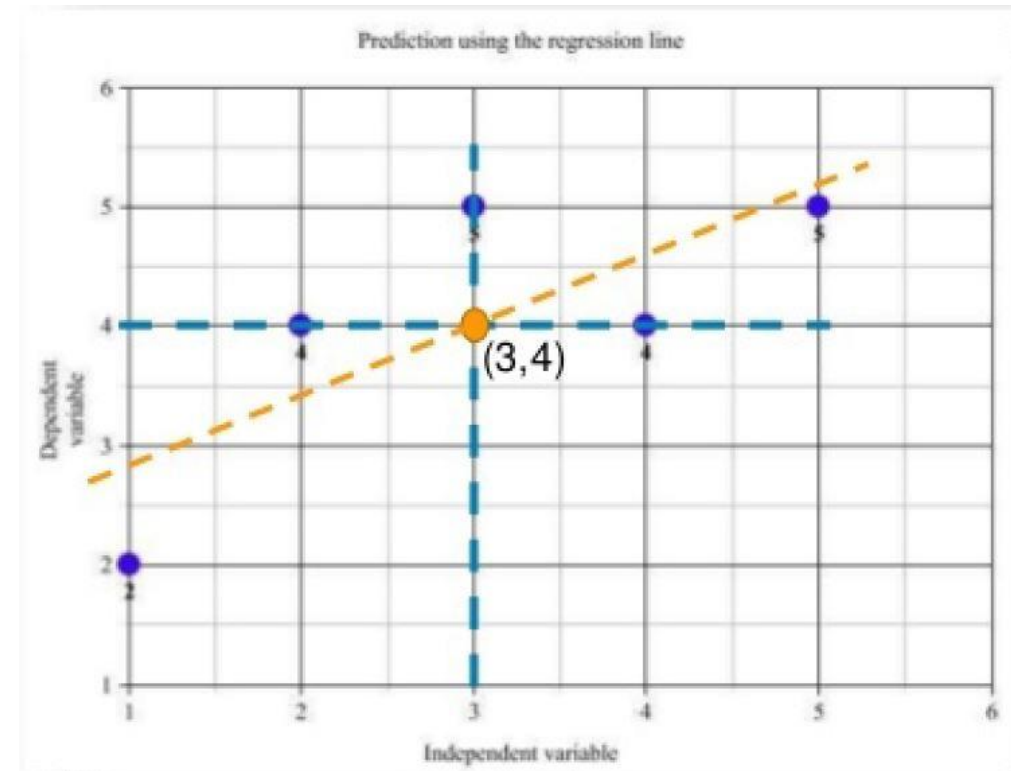
Regression line should ideally pass through the mean of X and Y.

Independent Variable X	Dependent Variable Y
1	2
2	4
3	5
4	4
5	5

Mean

3

4



Regression Line

# Intuition behind the regression Line

Drawing the Equation of the Regression line

X	Y	$x^2$	$y^2$	$x * y$
1	2	1	4	2
2	4	4	16	8
3	5	9	25	15
4	4	16	16	16
5	5	25	25	25

$$\Sigma = 15$$

$$\Sigma = 20$$

$$\Sigma = 25$$

$$\Sigma = 86$$

$$\Sigma = 66$$

$$y = mx + c$$

$$y = 0.6 * 3 + 2.2$$

$$= 4$$

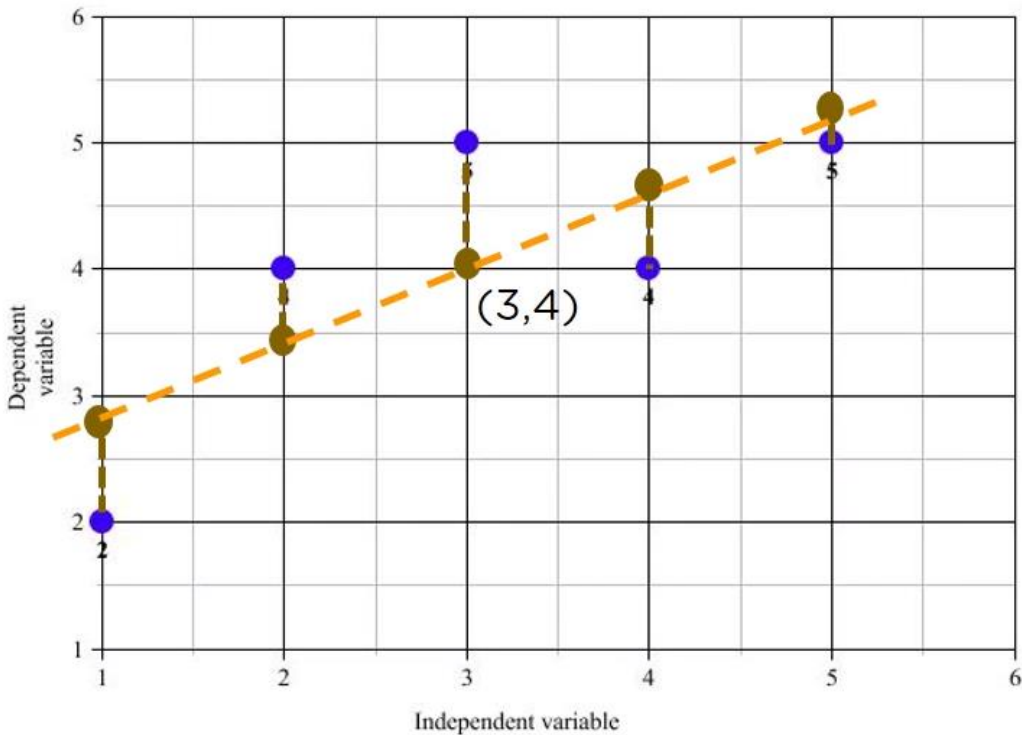
$$m = \frac{(n * \Sigma xy) - (\Sigma x * \Sigma y)}{(n * \Sigma x^2) - (\Sigma x)^2} = \frac{(5 * 66) - (15 * 20)}{(5 * 25) - 225} = 0.6$$

$$c = \frac{(\Sigma y * \Sigma x^2) - (\Sigma x * \Sigma xy)}{(n * \Sigma x^2) - (\Sigma x)^2} = 2.2$$

# Intuition behind the regression Line

Lets find out the predicted values of Y for corresponding values of X using the linear equation where  $m=0.6$  and  $c=2.2$

Prediction using the regression line



$Y_{pred}$

$$Y = 0.6 * 1 + 2.2 = 2.8$$

$$Y = 0.6 * 2 + 2.2 = 3.4$$

$$Y = 0.6 * 3 + 2.2 = 4$$

$$Y = 0.6 * 4 + 2.2 = 4.6$$

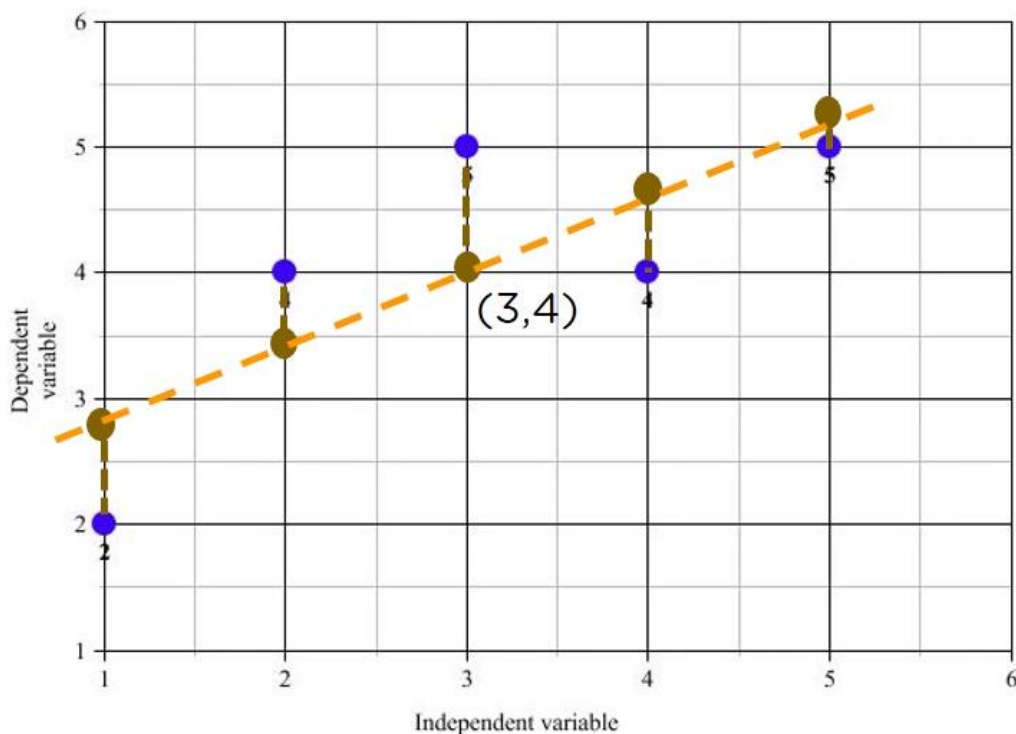
$$Y = 0.6 * 5 + 2.2 = 5.2$$

Here the blue points represent the **actual Y values** and the brown points represent the **predicted Y values**. The distance between the actual and predicted values are known as *residuals or errors*. The best fit line should have the least sum of squares of these errors also known as *e square*.

# Intuition behind the regression Line

Lets find out the predicted values of Y for corresponding values of X using the linear equation where  $m=0.6$  and  $c=2.2$

Prediction using the regression line



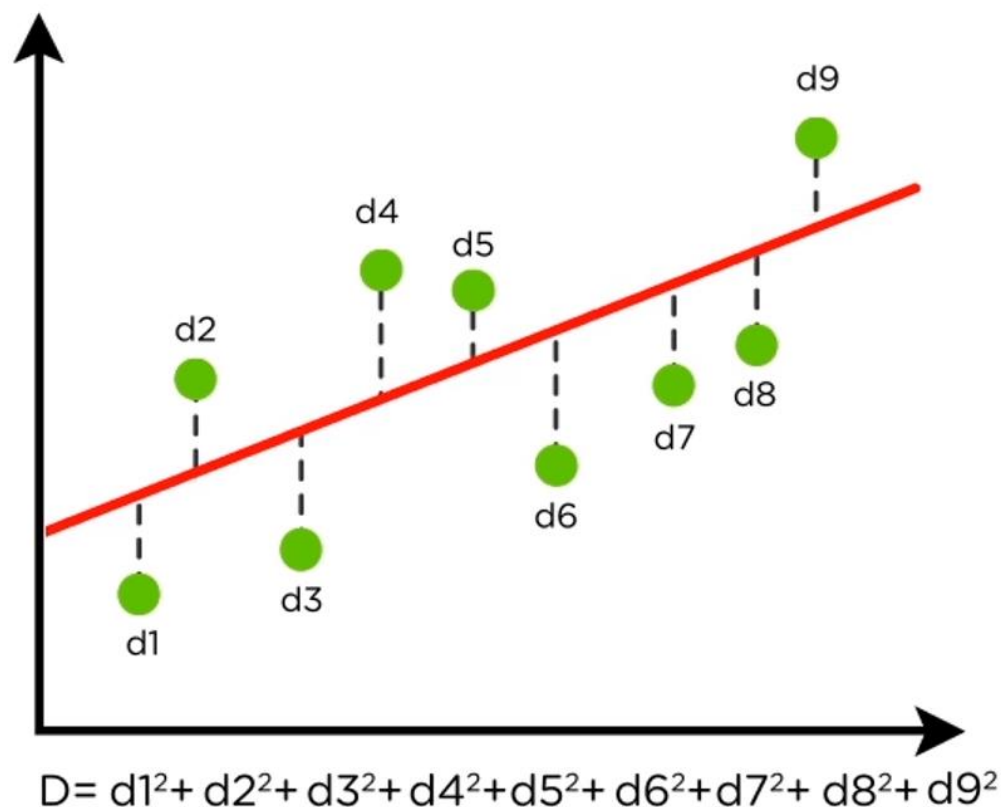
X	Y	$Y_{\text{pred}}$	$(Y - Y_{\text{pred}})$	$(Y - Y_{\text{pred}})^2$
1	2	2.8	-0.8	0.64
2	4	3.4	0.6	0.36
3	5	4	1	1
4	4	4.6	-0.6	0.36
5	5	5.2	-0.2	0.04

$$\sum = 2.4$$

The sum of squared errors for this regression line is 2.4. We check this error for each line and conclude the best fit line having the least square value.

# Finding the Best Fit Line

**Minimizing the Distance:** There are lots of ways to minimize the distance between the line and the data points like Sum of Squared errors, Sum of Absolute errors, Root Mean Square error etc.



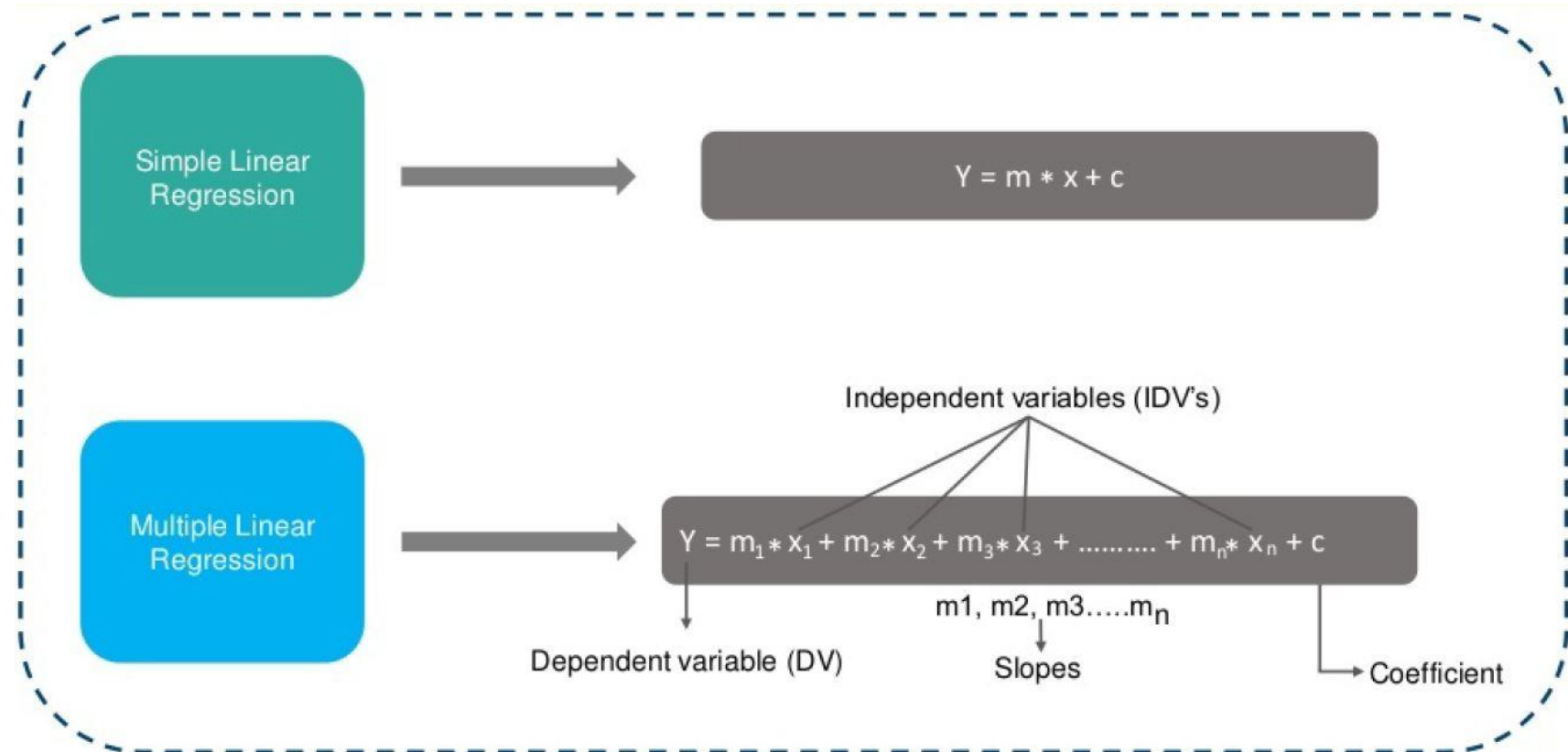
We keep moving this line through the data points to make sure the Best fit line has the least square distance between the data points and the regression line

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# Multiple Linear Regression

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# Multiple Linear Regression



# Implementation of Linear Regression

► Use Case



Let's Understand how  
the Linear Regression  
Works by implementing  
it in Python



- For instance we will use Python 3.7 Version for Windows platform.

You may use any of  
the API depending  
on the configuration  
compatible with your  
system

## JUPYTER NOTEBOOK



## GOOGLE COLAB



Presented By:

**THANK YOU!**  
**Q&A**



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