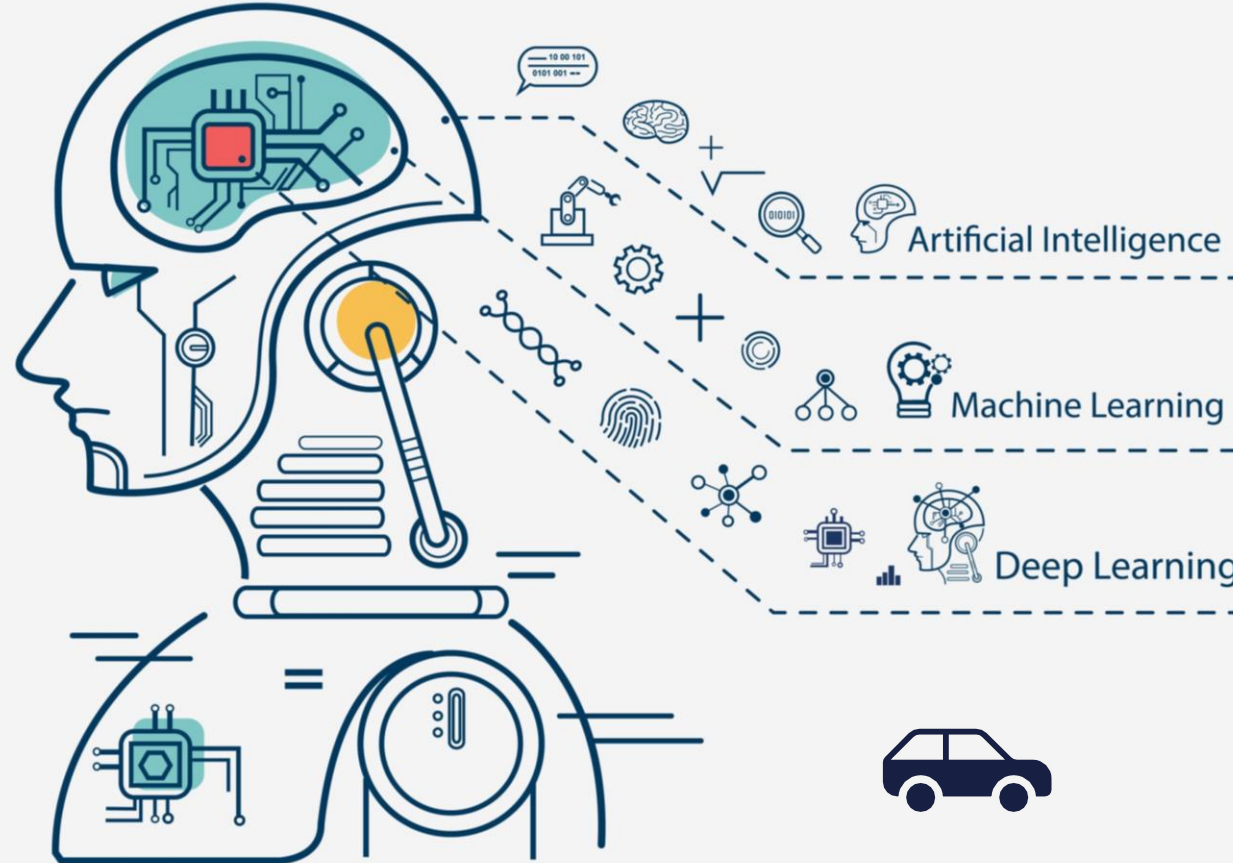




APSSDC

Andhra Pradesh State Skill Development Corporation



MACHINE LEARNING USING PYTHON

DAY2 AGENDA

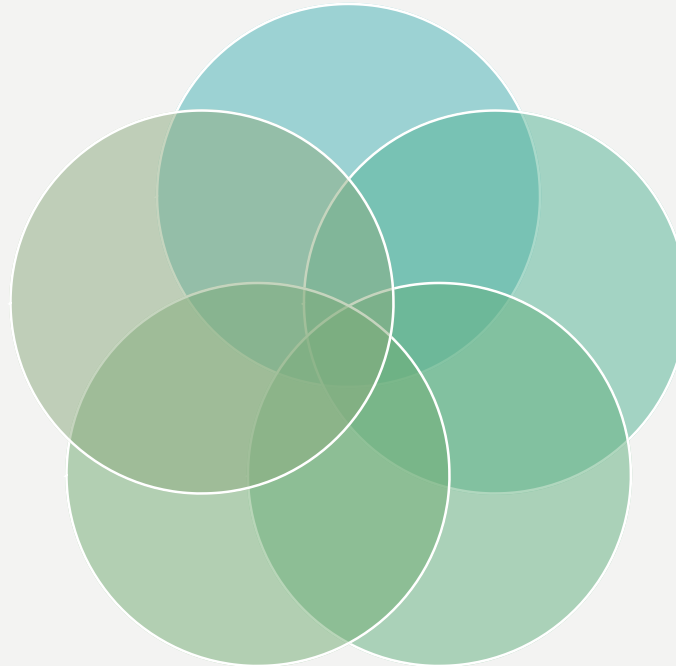
Linear
Regression with
One variable

Linear
Regression
with Multiple
Variables

Evaluation
Metrics in
Regression
Models

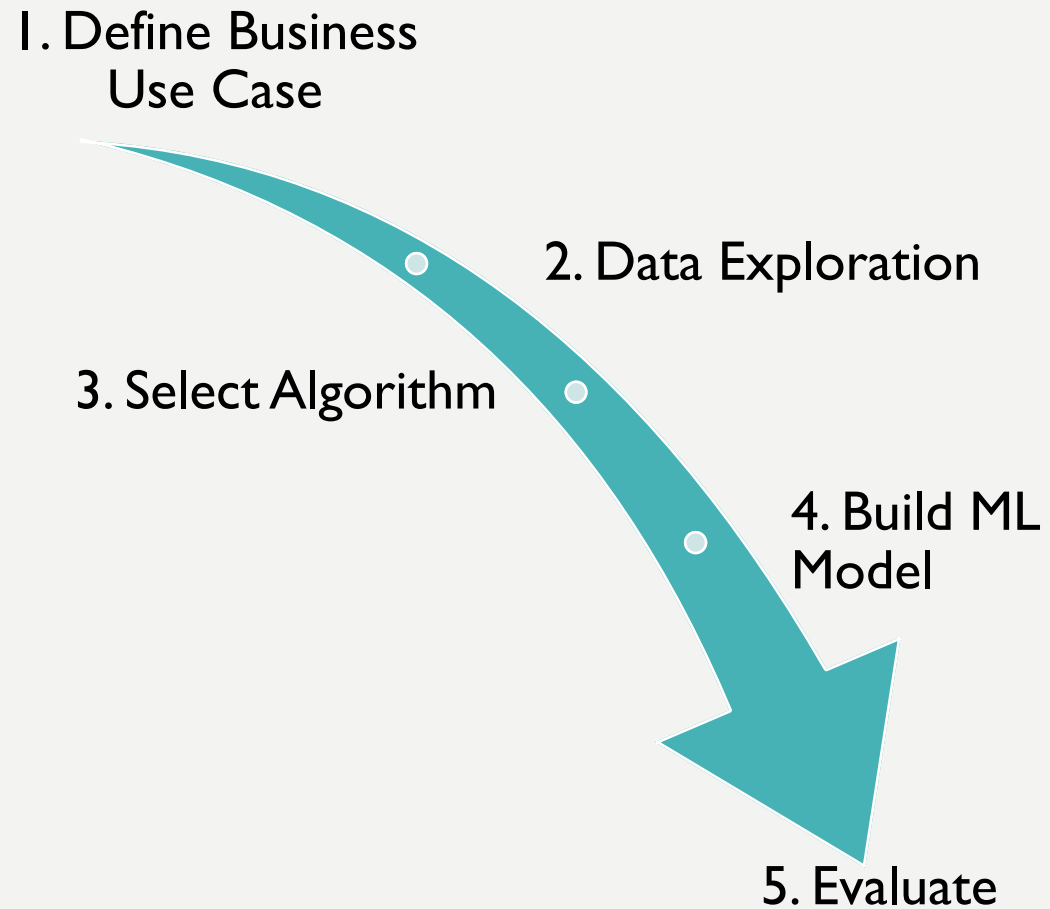
Cross
Validation

Train/Test
splitting of
data



- Regression
- Classification

ML MODEL DEVELOPMENT LIFE CYCLE

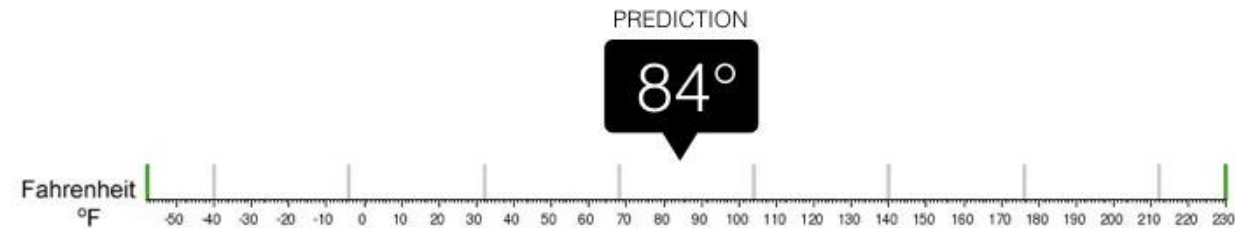


REGRESSION VS CLASSIFICATION



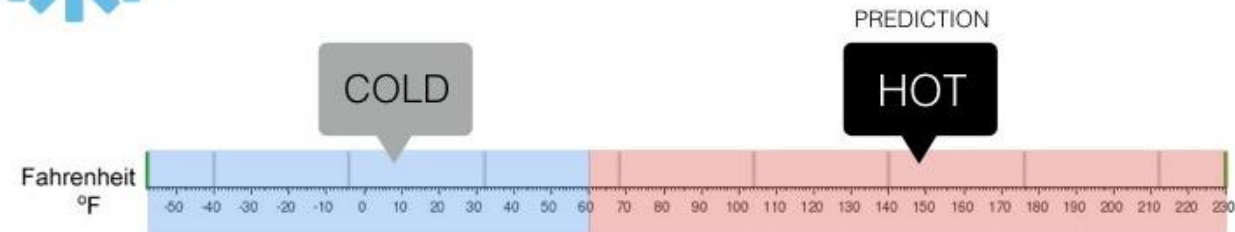
Regression

What is the temperature going to be tomorrow?



Classification

Will it be Cold or Hot tomorrow?



Linear Regression in Machine Learning



What is Regression?

- Function: a mathematical relationship enabling us to predict what values of one variable (Y) correspond to given values of another variable (X).
- Y : is referred to as the **dependent variable**, the **response variable** or the **predicted variable**.
- X : is referred to as **the independent variable**, the **explanatory variable** or the **predictor variable**.

Thus Regression analysis is a form of predictive modelling technique which investigates the relationship between a dependent and independent variable.

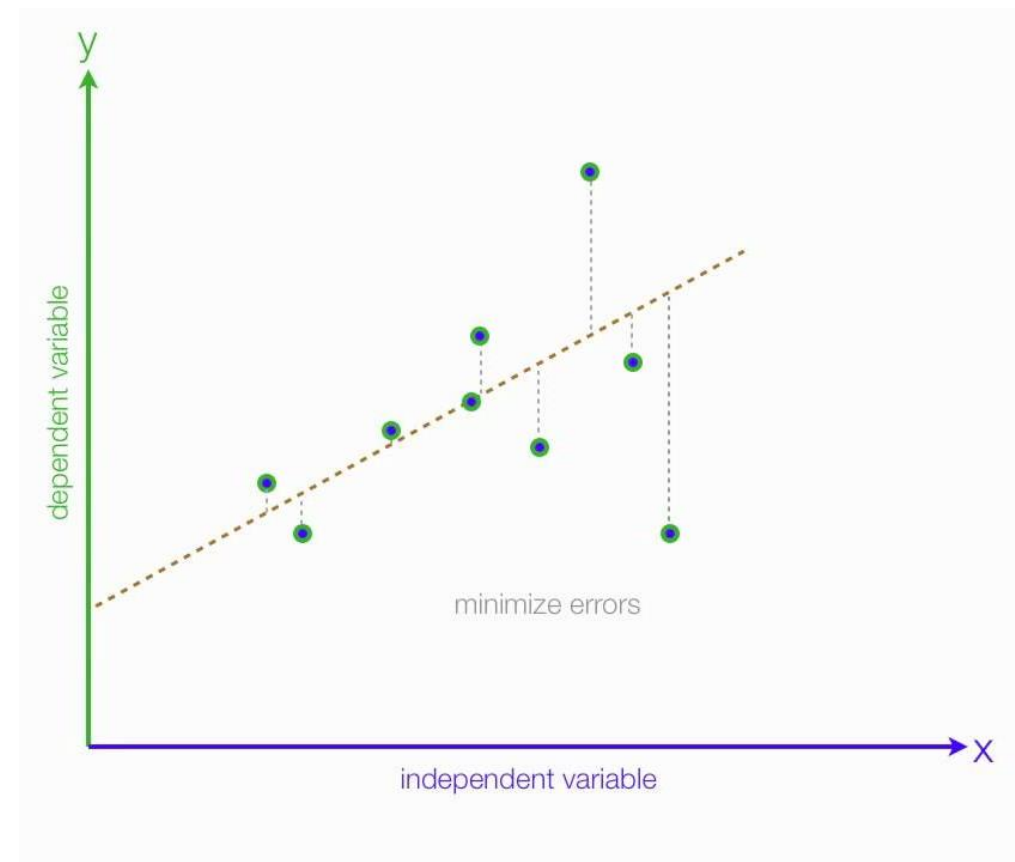
REGRESSION

- **Linear Regression**
 - Linear Regression with one variable
 - Linear Regression with multiple variable
- Non-Linear Regression/Polynomial Regression
 - Non-Linear Regression with one variable
 - Non-Linear Regression with multiple variables
- SGD
- Ridge
- Lasso
- Elastic Net

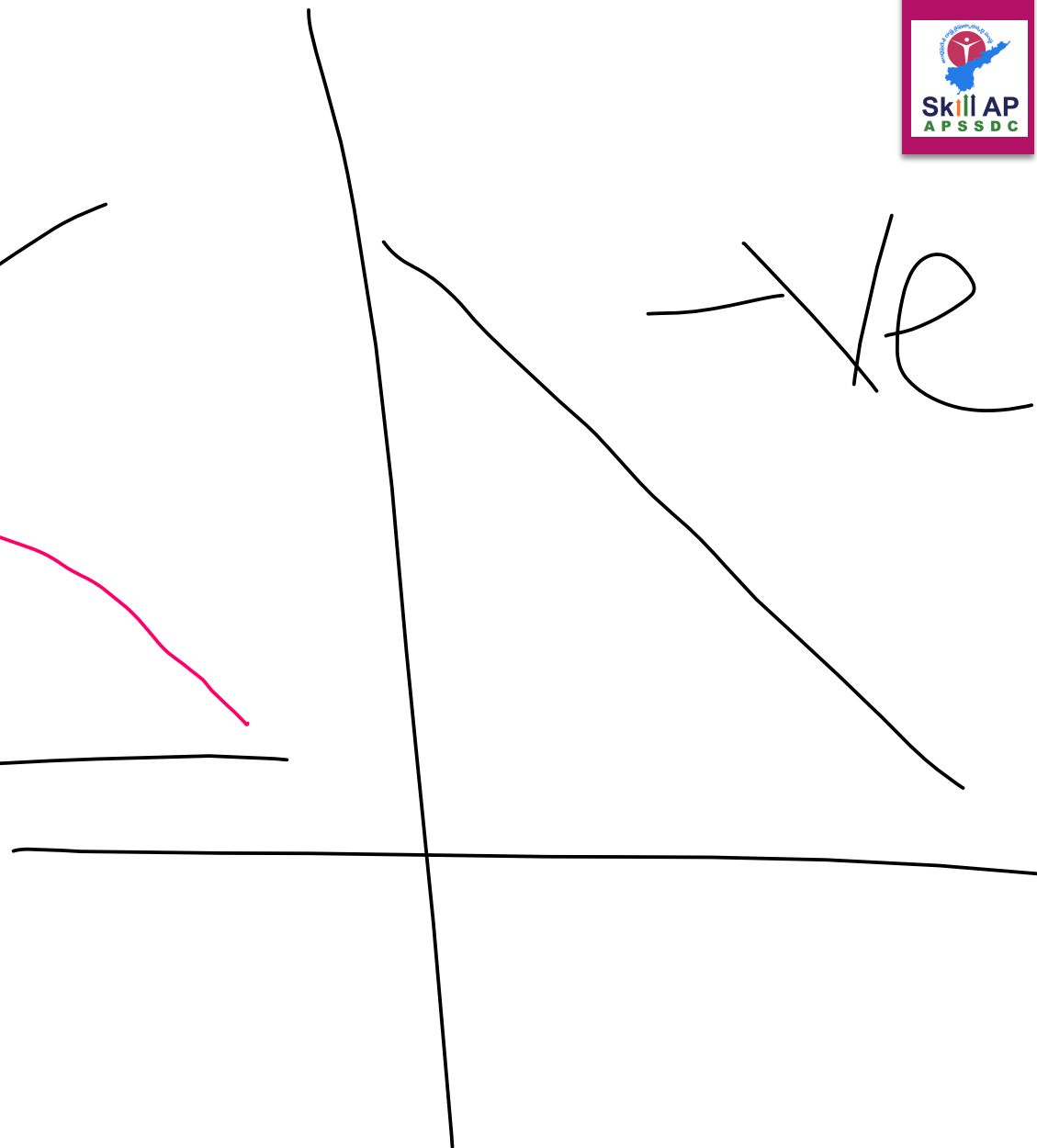
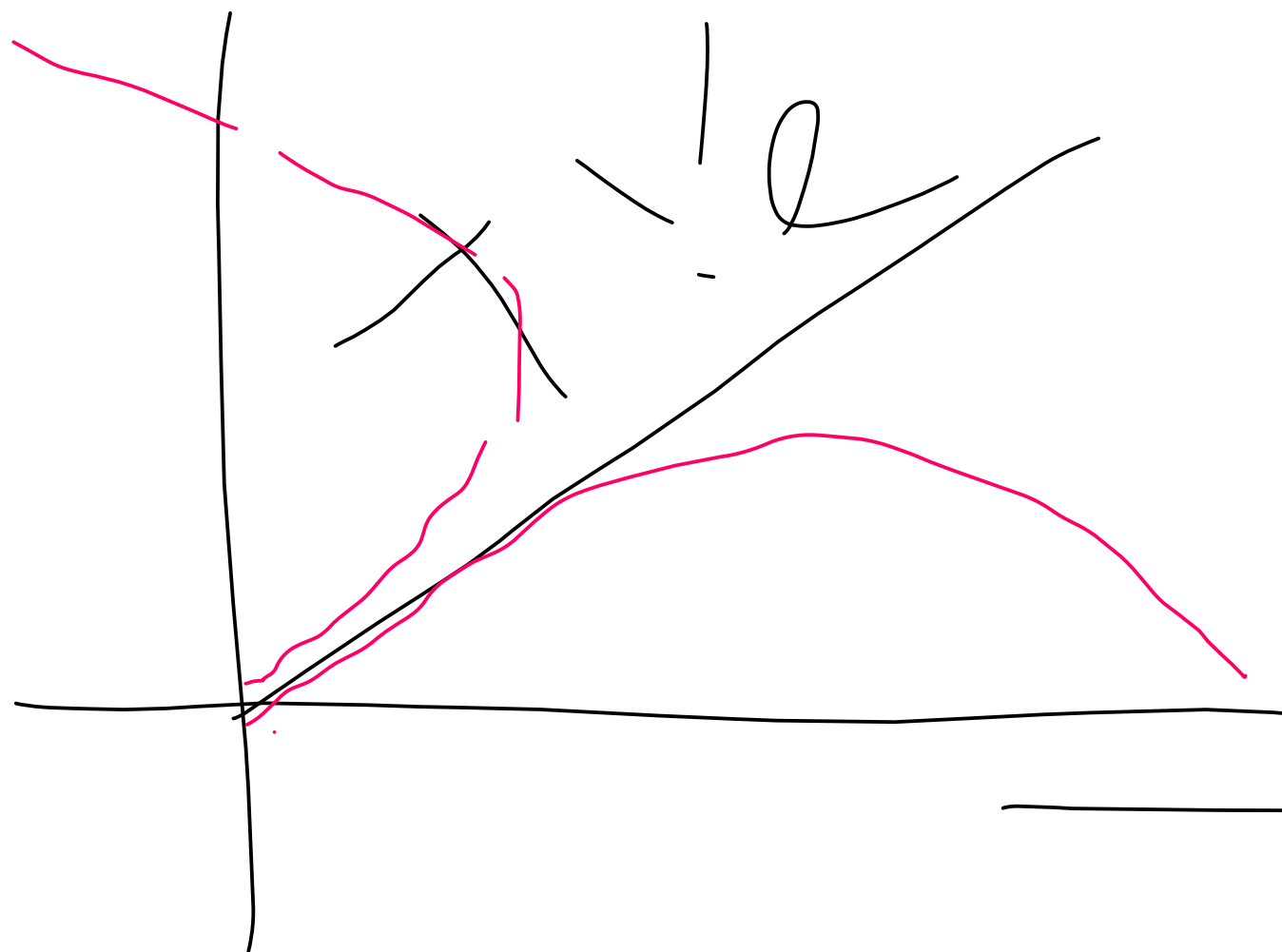
Contd.

A typical Linear Regression model can be represented in the form :

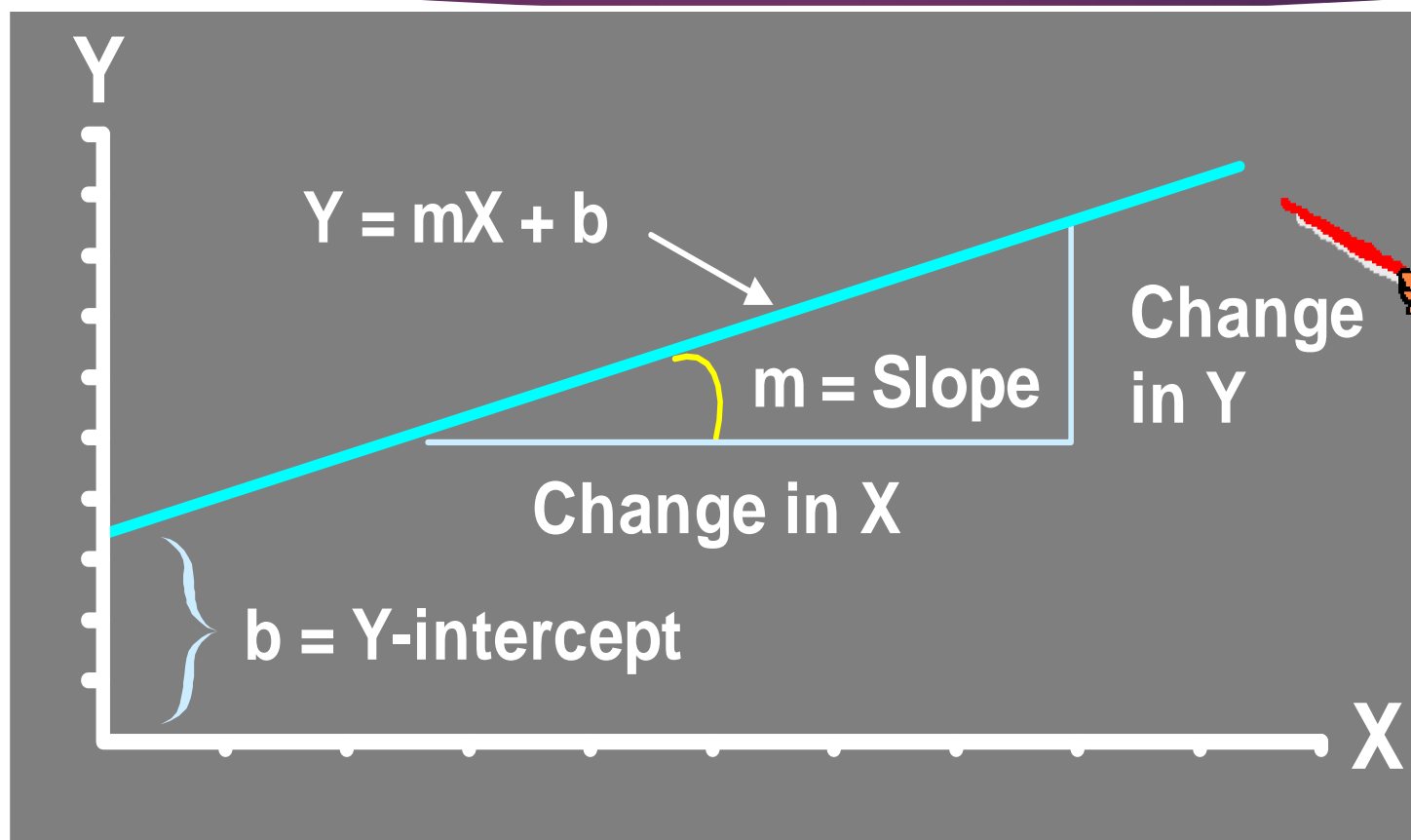
$y = b_1x + b_0$ where b_1 is slope and b_0 is the intercept.



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Linear Equations



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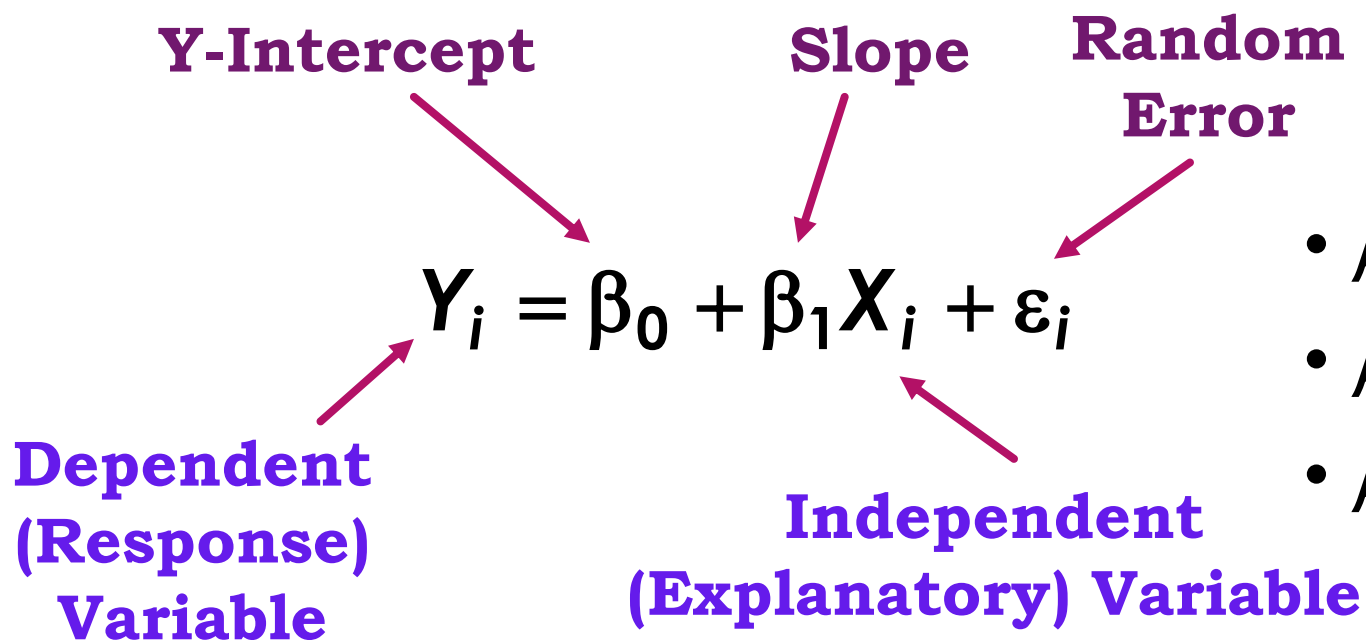
Linear Regression Model

Relationship Between Variables Is a Linear Function

Y-Intercept **Slope** **Random Error**

$$Y_i = \beta_0 + \beta_1 X_i + \varepsilon_i$$

Dependent (Response) Variable **Independent (Explanatory) Variable**

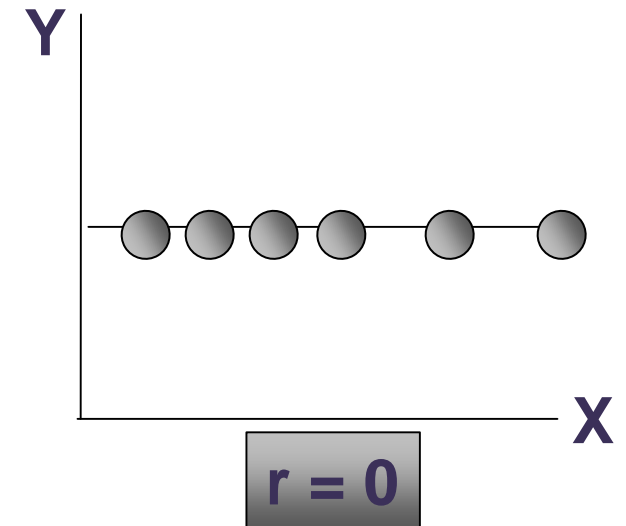
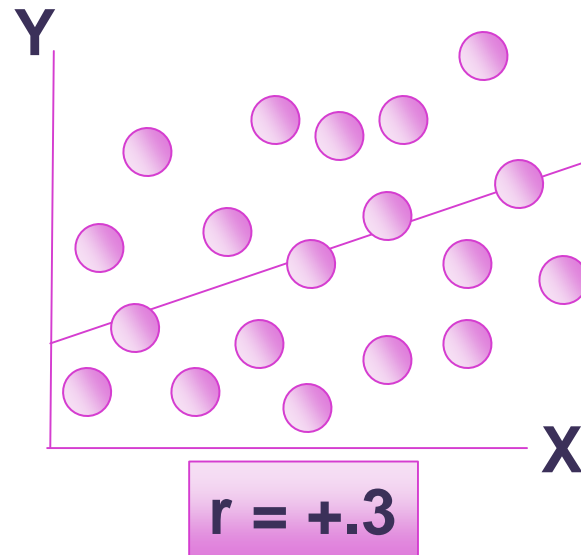
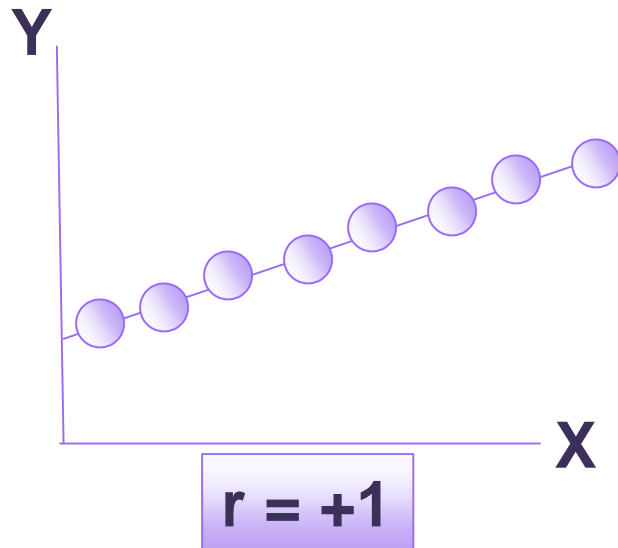
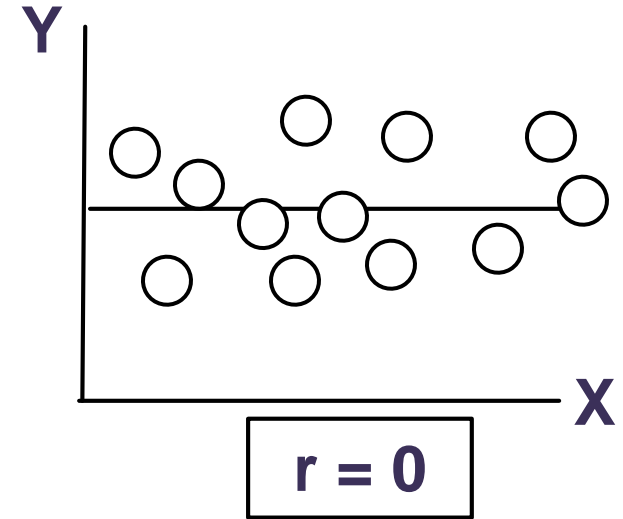
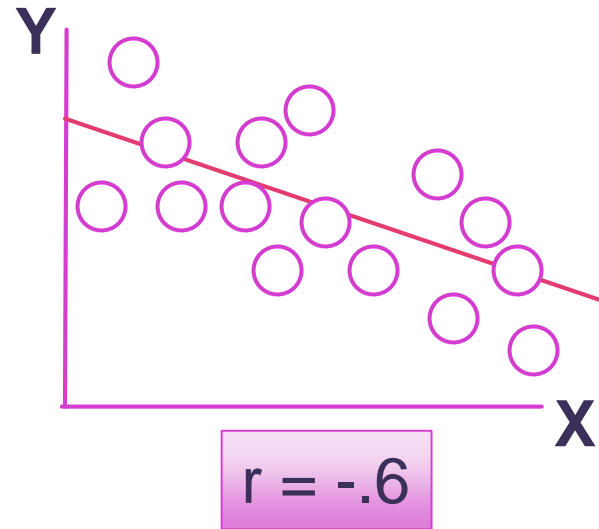
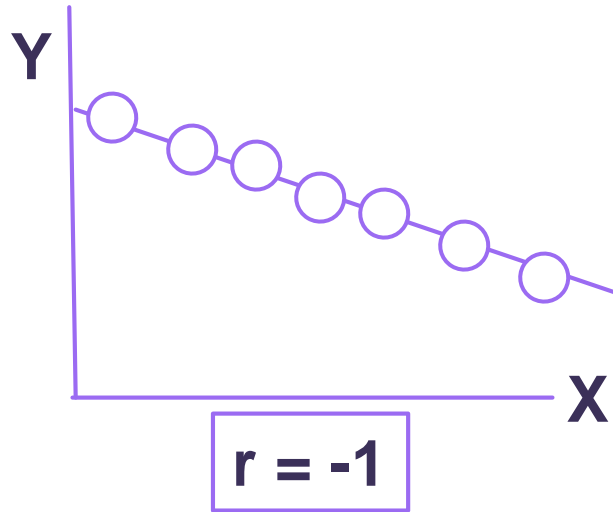


- $\beta_1 > 0 \Rightarrow$ Positive Association
- $\beta_1 < 0 \Rightarrow$ Negative Association
- $\beta_1 = 0 \Rightarrow$ No Association

Correlation

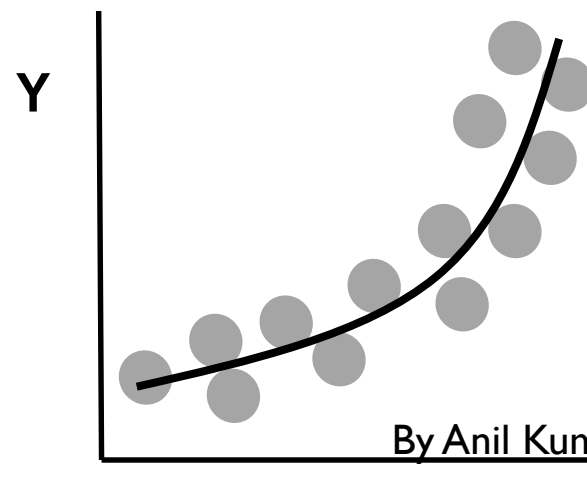
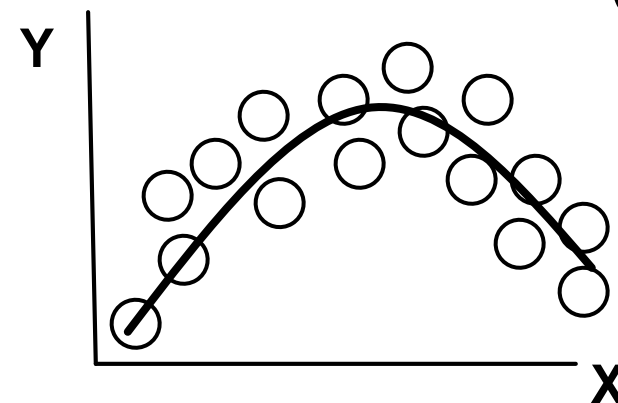
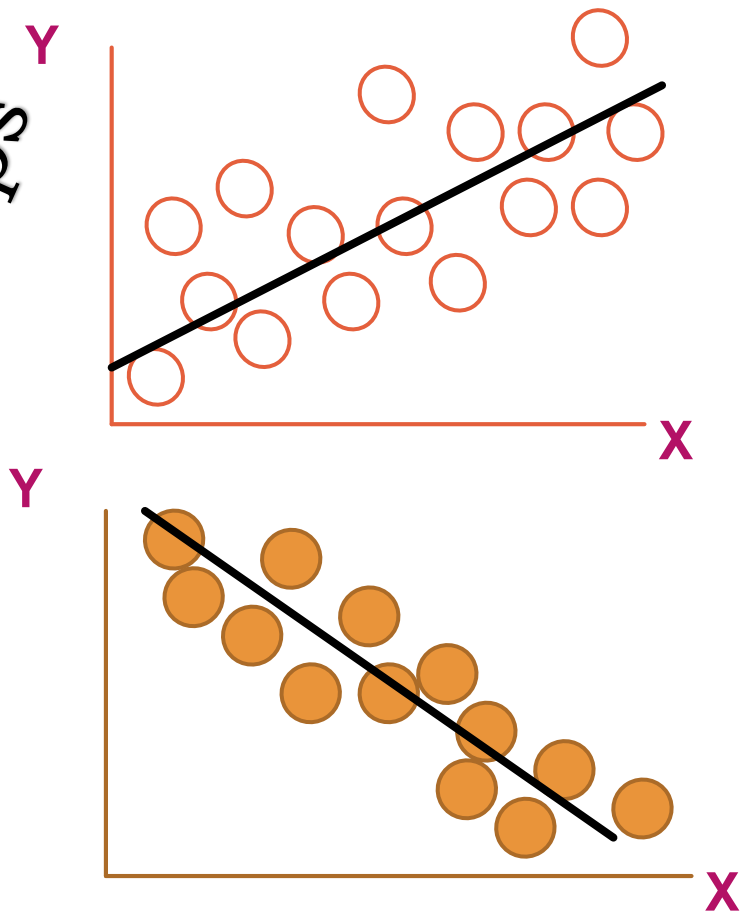
- ▶ Measures the relative strength of the *linear* relationship between two variables Unit-less
- ▶ Ranges between -1 and 1
- ▶ The closer to -1 , the stronger the negative linear relationship
- ▶ The closer to 1 , the stronger the positive linear relationship
- ▶ The closer to 0 , the weaker any positive linear relationship

Scatter Plots of Data with Various Correlation Coefficients



Linear Correlation

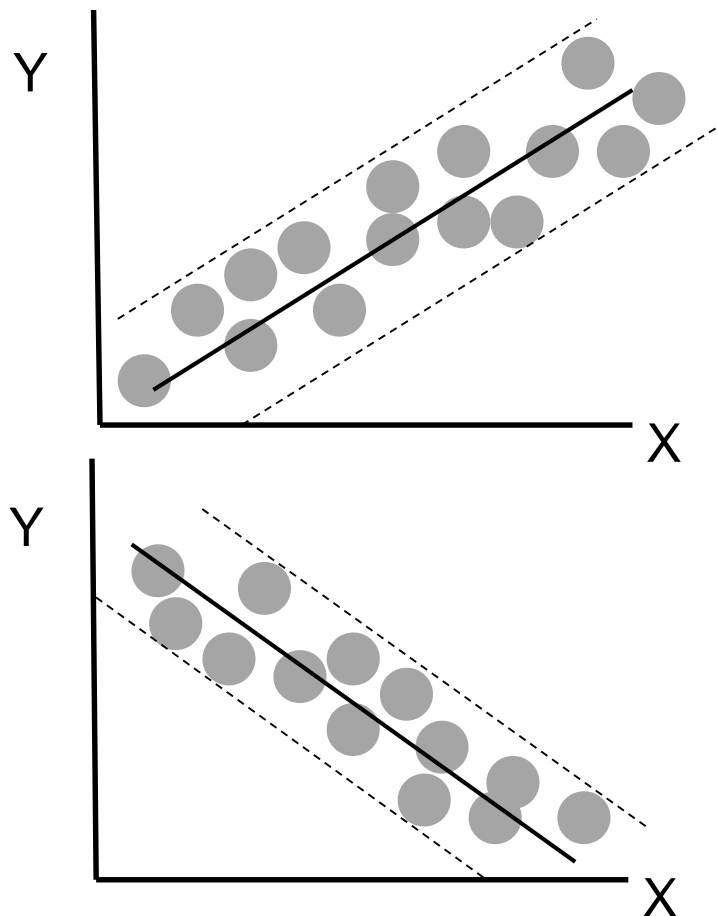
Linear relationships



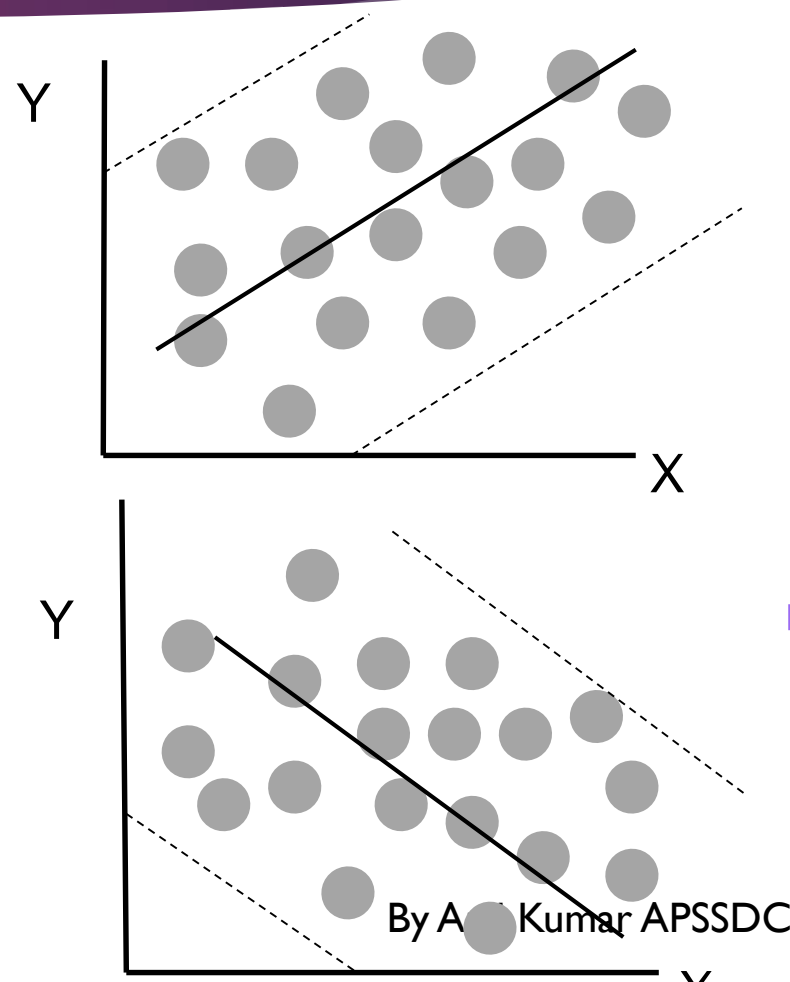
Curvilinear relationships

Linear Correlation

**Strong
relationships**



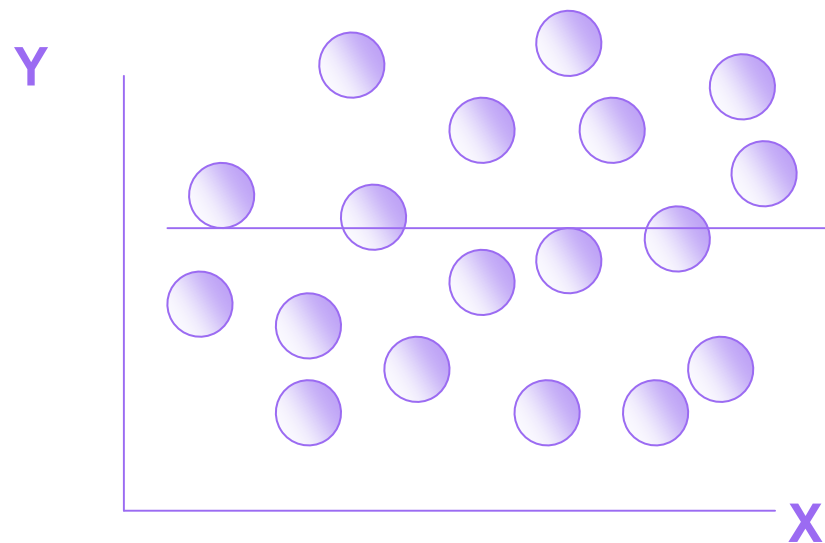
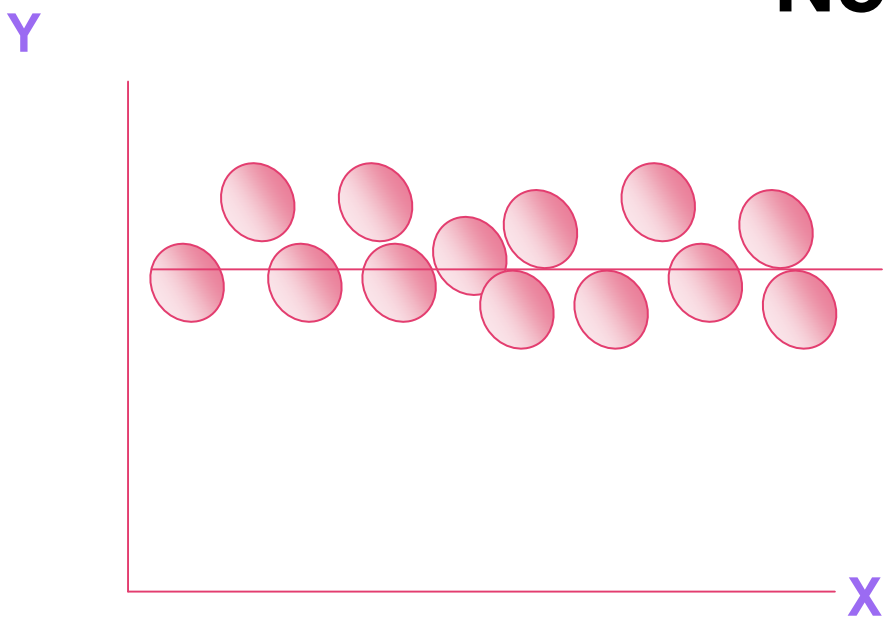
**Weak
relationships**



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Linear Correlation

No relationship



Steps in Regression Analysis

- Examine the scatterplot of the data.
 - I. Does the relationship look linear?
 - II. Are there points in locations they shouldn't be?
 - III. Do we need a transformation?
- Assuming a linear function looks appropriate, estimate the regression parameters.
 - I. How do we do this? (Method of Least Squares)
- If there is a significant linear relationship, estimate the response, Y , for the given values of X , and compute the residuals

Regression Analysis

- Thus we have the regression formula as :

$$Y = MX + C + \text{error}(e).$$

Initially we calculate the value for slope and predict the values of Y for any given X values we have.

$$\text{Slope}(M) = \sum_{i=0}^{\text{len}(X)} \frac{(X_i - X_{\text{mean}}) * (Y_i - Y_{\text{mean}})}{(X_i - X_{\text{mean}})^2}$$

Thus we calculate the C value and find out the “Line of Regression”.

Regression Analysis

- Next our job is to reduce the distance between the actual value and the predicted value or in other words reduce the error between the actual and predicted value. Thus the line with least error will be the “**Best Fit Line**”.
- In order to check it out we calculate the “Coefficient of Determination”.

$$\text{Mean Squared value } (R^2) = \sum_{i=0}^{\text{len}(X)} \frac{(Y_{\text{pred}} - Y_{\text{mean}})^2}{(Y - Y_{\text{mean}})^2}$$

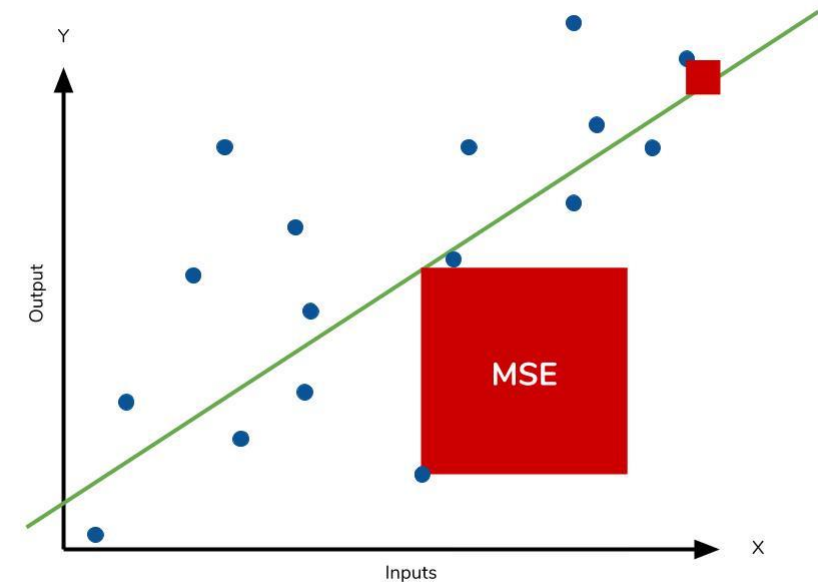
- Thus our ultimate aim is to reduce the error i.e. distance between the actual and predicted values.

Contd..

2. Mean Square Error

$$MSE = \frac{1}{n} \sum \left(y - \hat{y} \right)^2$$

The square of the difference between actual and predicted

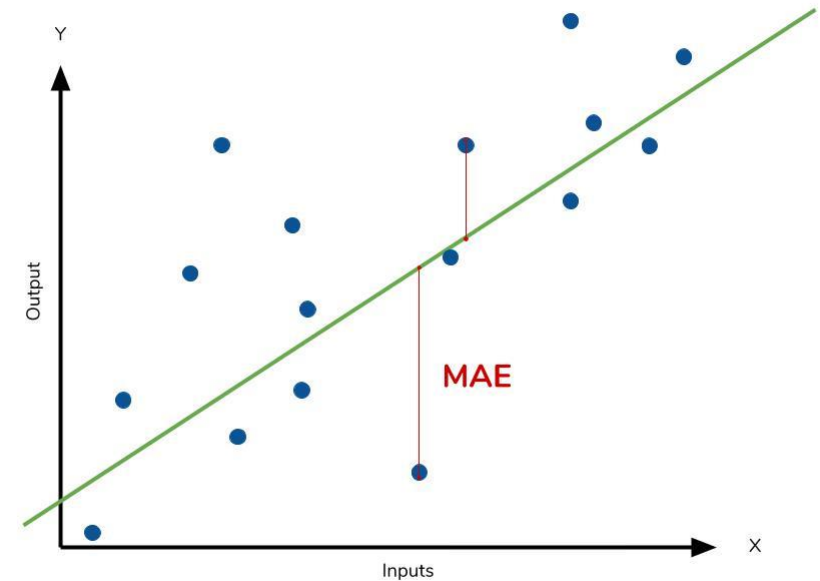


Evaluation Metrics

1. Mean Absolute Error

$$MAE = \frac{1}{n} \sum |y - \hat{y}|$$

Divide by the total number of data points (points to $\frac{1}{n}$)
 Actual output value (points to y)
 Predicted output value (points to \hat{y})
 Sum of (points to \sum)
 The absolute value of the residual (points to $|y - \hat{y}|$)

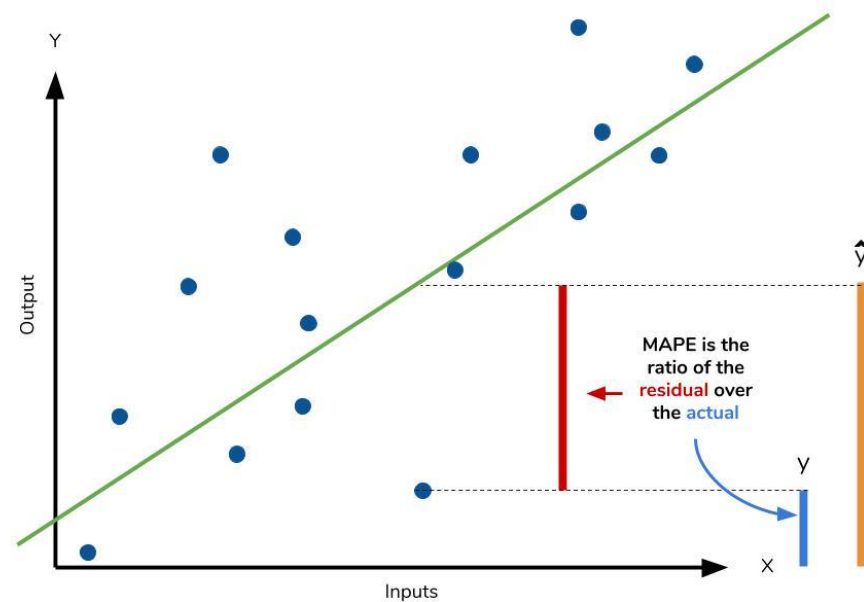


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3. Mean Absolute Percentage Error

$$MAPE = \frac{100\%}{n} \sum \left| \frac{\overbrace{y - \hat{y}}^{\text{The residual}}}{\underbrace{y}_{\text{Each residual is scaled against the actual value}}} \right|$$

Multiplying by 100% converts to percentage

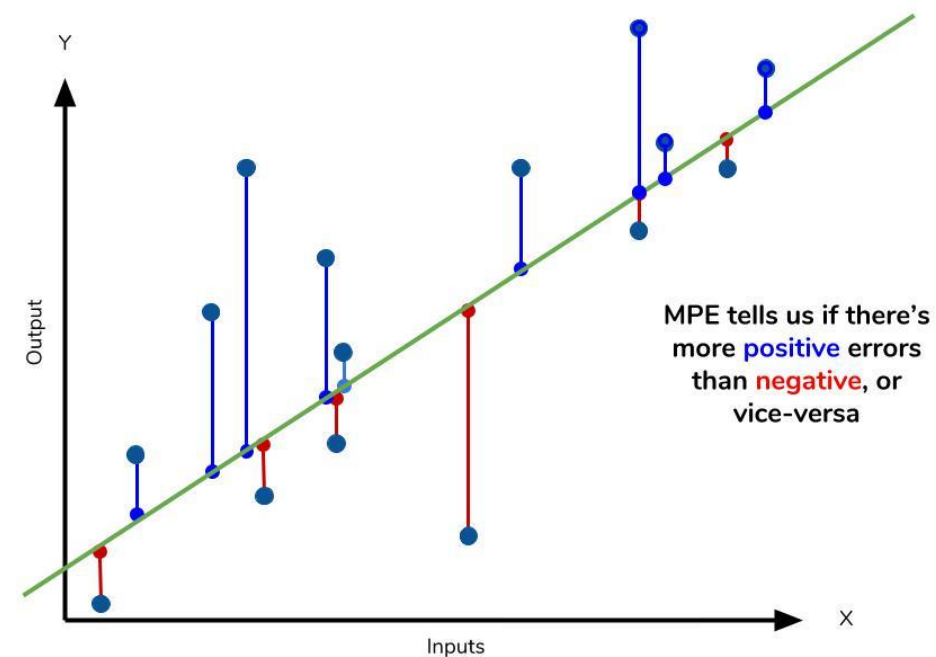


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Contd..

3. Mean Percentage Error

$$MPE = \frac{100\%}{n} \sum \left(\frac{y - \hat{y}}{y} \right)$$



Conclusion

Acroynm	Full Name	Residual Operation?	Robust To Outliers?
MAE	Mean Absolute Error	Absolute Value	Yes
MSE	Mean Squared Error	Square	No
RMSE	Root Mean Squared Error	Square	No
MAPE	Mean Absolute Percentage Error	Absolute Value	Yes
MPE	Mean Percentage Error	N/A	Yes