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This session deals with

- Statistics

- Why Statistics?

- Inferential Statistics

- Statistics Learning

- What is Linear Regression

- Linear Regression functionality



# Statistics



- Statistics plays very important role in Data Science
- All the required algorithms for analysis are available in statistics
- Data representation in graphical and numeric is handled in statistics

## Why Statistics?

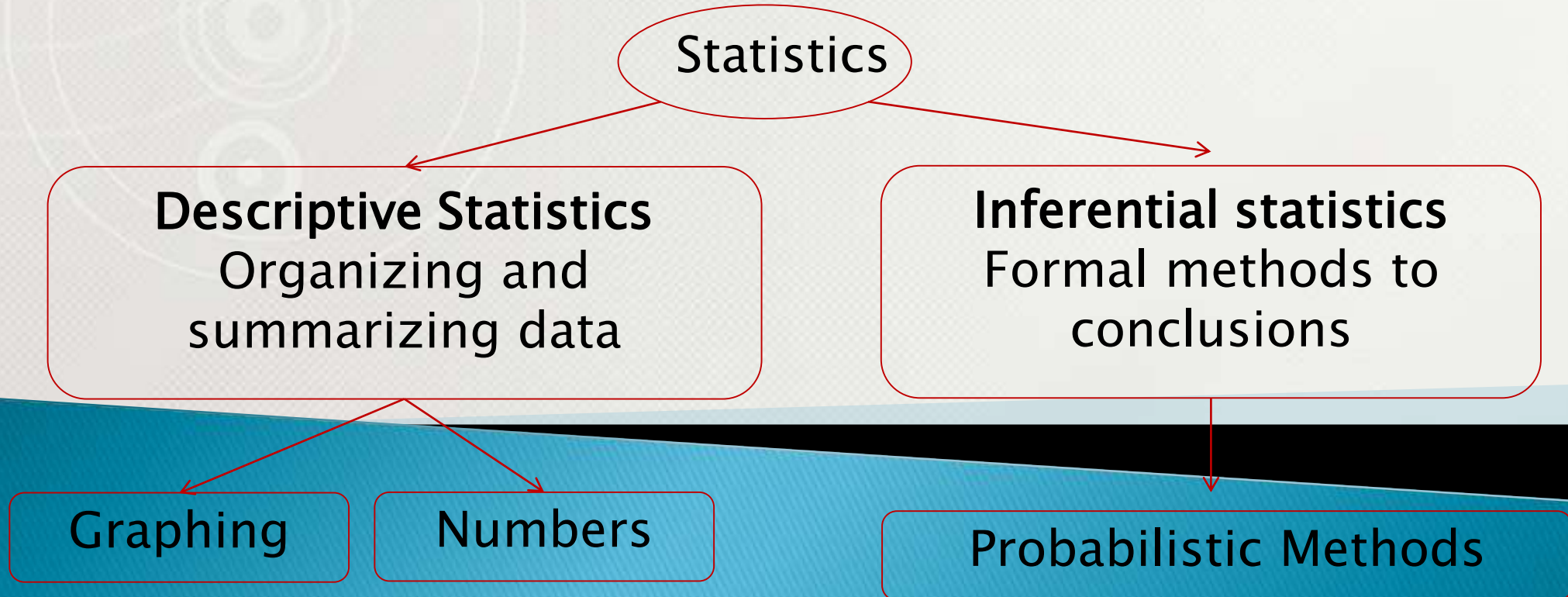


- Everyone is using statistics unknowingly.
- Ex: to buy a mobile... we may compare...  
cost      Performance      Warranty      Etc...
- We can find statistics in (almost everywhere):  
News paper      Television      Sports      Stock Market
- Sources provide sample information
  - We can understand and take decision
- It is the body of methods for making **wise decisions** in the face of uncertainty on the basis of numerical data and calculated risks.



# Statistics

- The science of Numbers that deals with the collection, analysis, interpretation and presentation of data.



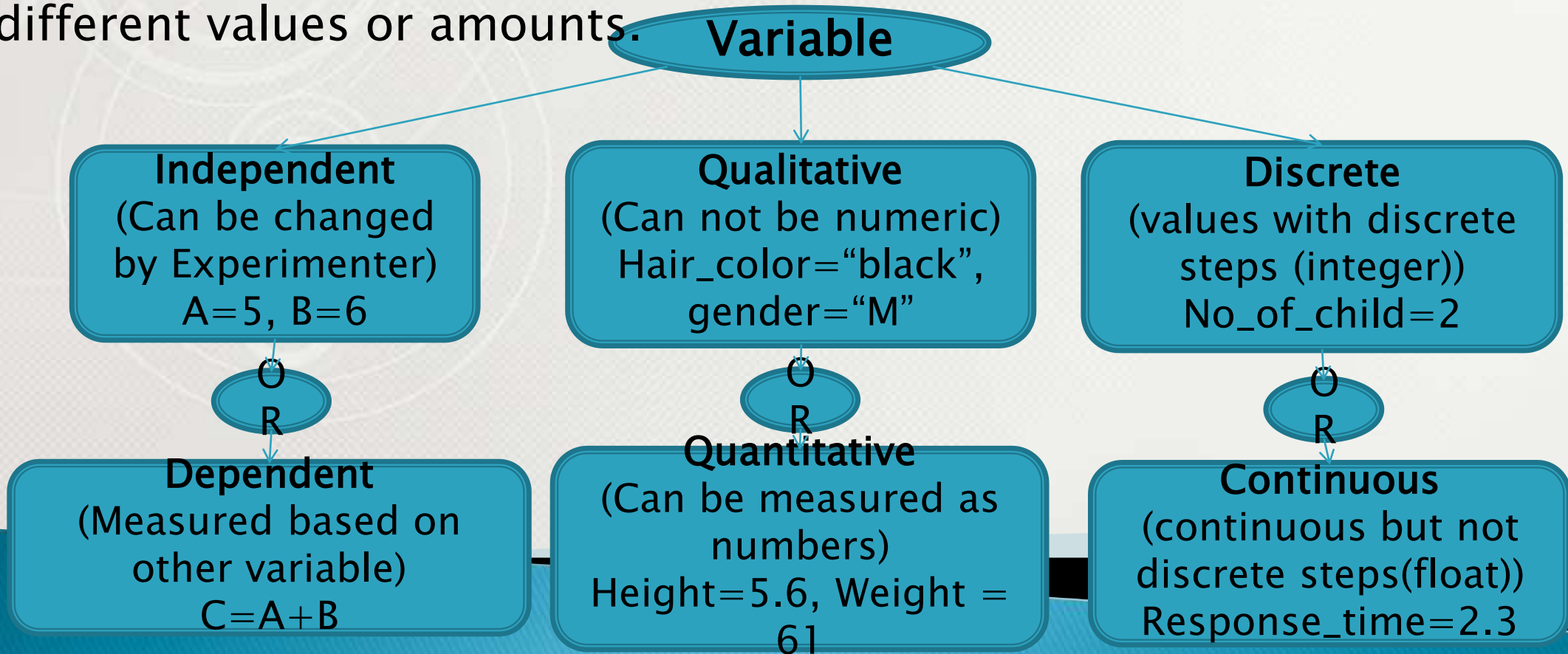
# Descriptive Statistics



- Deals with Organizing and summarizing data.
- used when we know all of the values in order to describe a phenomenon
- There are 3 steps
  - Describing Data
  - Summarize / Analyze
  - Visualize



- Variables : properties of some event, object, or person that can take on different values or amounts.



- Used for conducting analysis on one variable at a time or univariate Analysis

# Inferential Statistics

Inferential statistics use a random sample of data taken from a population to describe and make inferences about the population.

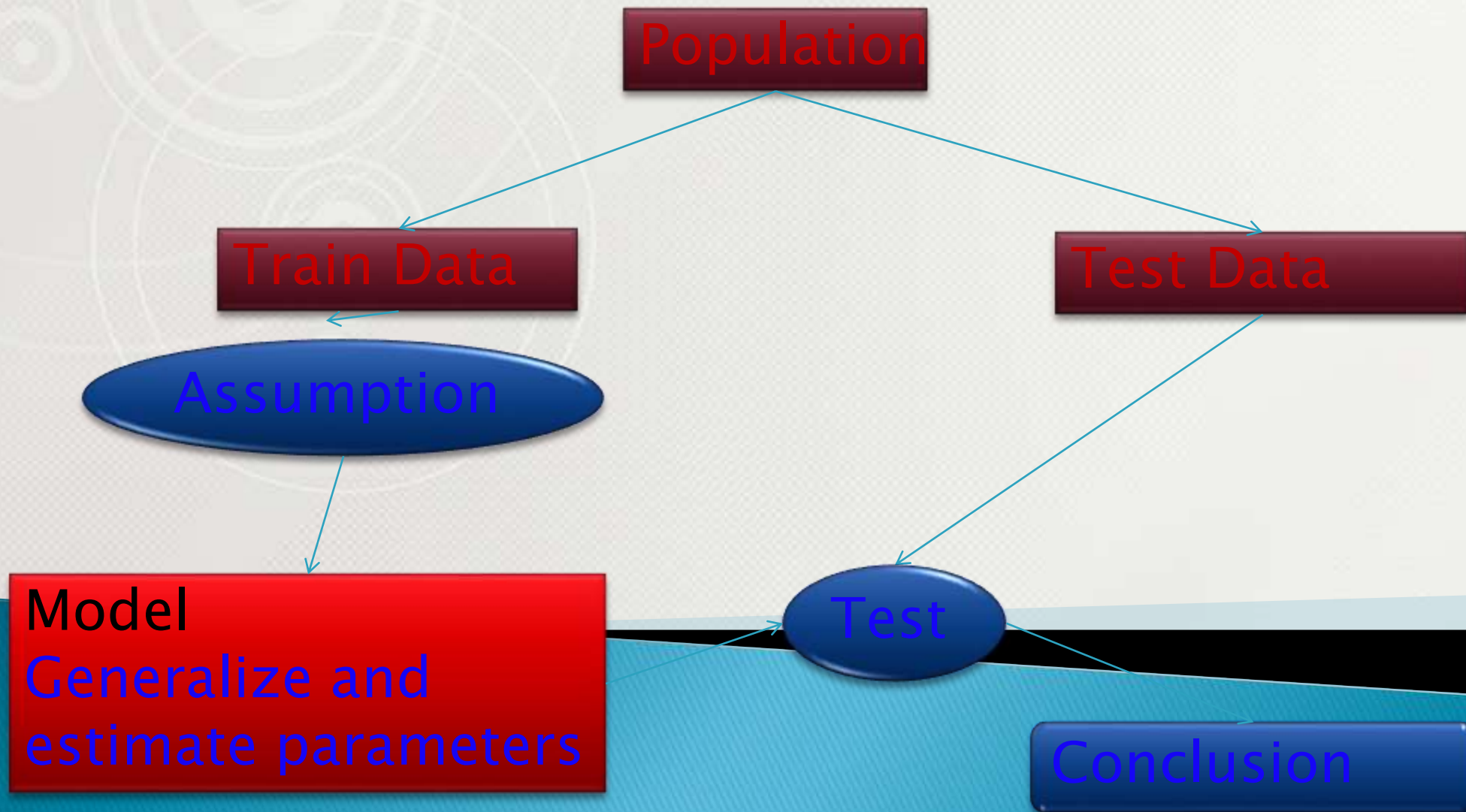
**Population :** Any group of data, which includes all the data interested in,

**Sample:** A smaller set of data, which are used to represent the larger population

The methods of inferential statistics

- (1) the estimation of parameter(s)
- (2) testing of statistical hypotheses







# Statistical Learning

Supervised

Unsupervised

Input Variable

Predictors /  
Independent  
Variables / Features

Output Variable

Response /  
Dependent  
Variables

Fit a model that relates to response to the predictors, for predicting the response for future observations.

Linear  
Regression

Logistic  
Regression

Etc...

Predictors

Cluster Analysis

No response variable to supervise so is called unsupervised learning.



In the context of Statistical learning, there are two types of data:

**Independent variables:** Data that can be controlled directly.

**Dependent variables:** Data that cannot be controlled directly. Dependent variables need to be predicted or estimated. To predict output will use model.

**A model is a transformation engine** that helps us to express dependent variables as a function of independent variables.

Parameters are ingredients added to the model for estimating the output.

# Linear Regression



**Linear:** arranged in or extending along a straight or nearly straight line.

Linear suggests that the relationship between dependent and independent variable can be **expressed in a straight line**.

$$y = mx + c$$

**y** is the dependent variable i.e. the variable that needs to be estimated and predicted.

**x** is the independent variable i.e. the variable that is controllable. It is the input.

**m** is the slope. It determines what will be the angle of the line. It is the parameter denoted as  $\beta$ .

**c** is the intercept. A constant that determines the value of **y** when **x** is 0.



# Linear Regression



Linear regression models are not perfect.

It tries to approximate the relationship between dependent and independent variables in a straight line.

Approximation leads to errors. Some errors can be reduced. Some errors are inherent in the nature of the problem.

These errors cannot be eliminated. They are called as an **irreducible error**

The noise term in the true relationship that cannot fundamentally be reduced by any model.

The equation can be re-written as

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

**$\beta_0$  and  $\beta_1$**  are two unknown constants that represent the intercept and slope.  **$\epsilon$**  is the error term

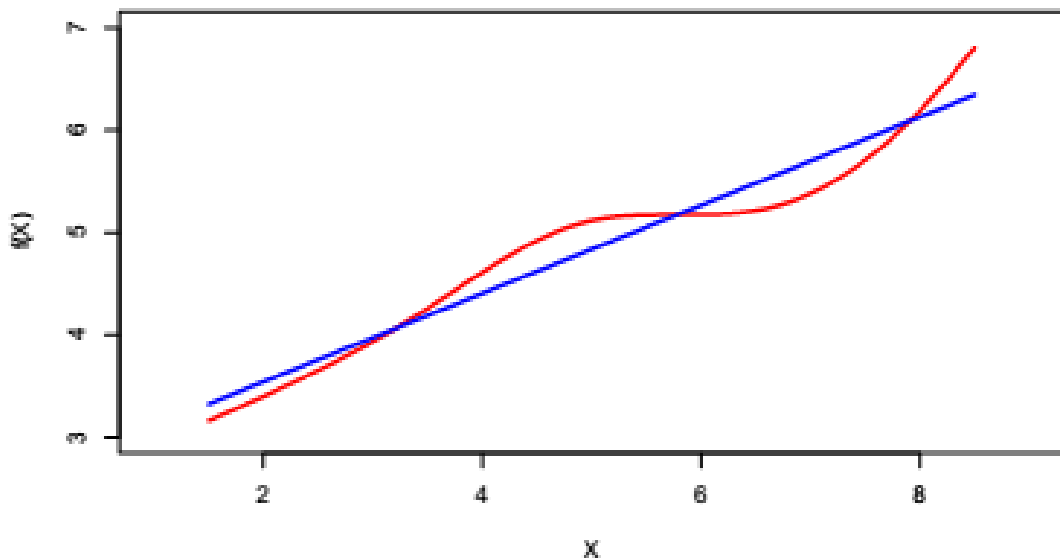
# Linear Regression

Tool for predicting an unknown value based on existing data.

Linear Regression is Supervised Learning

Predictors  $X_1, X_2, X_3, \dots$

Response  $Y=f(x)$



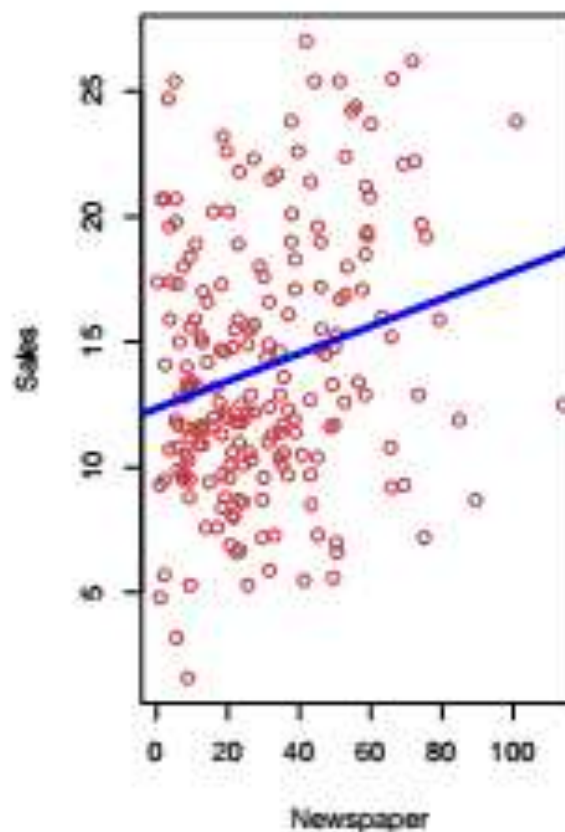
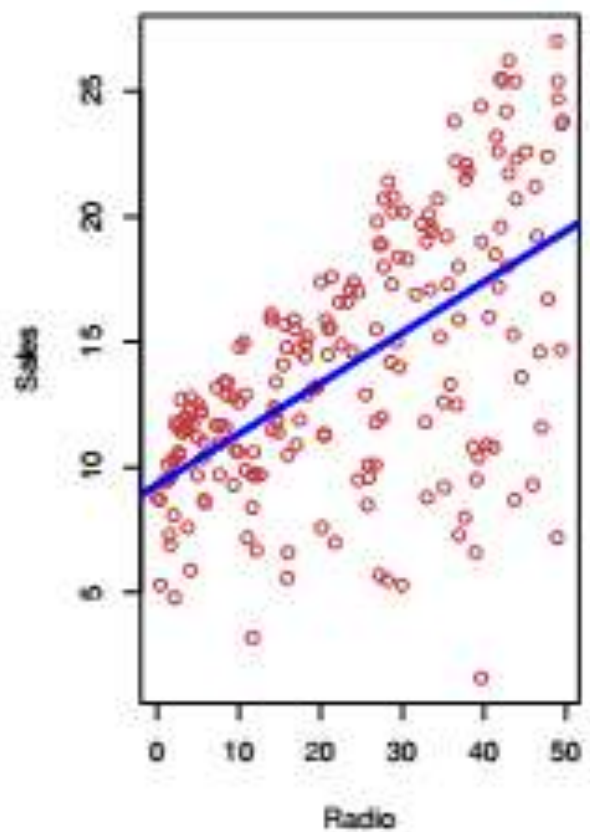
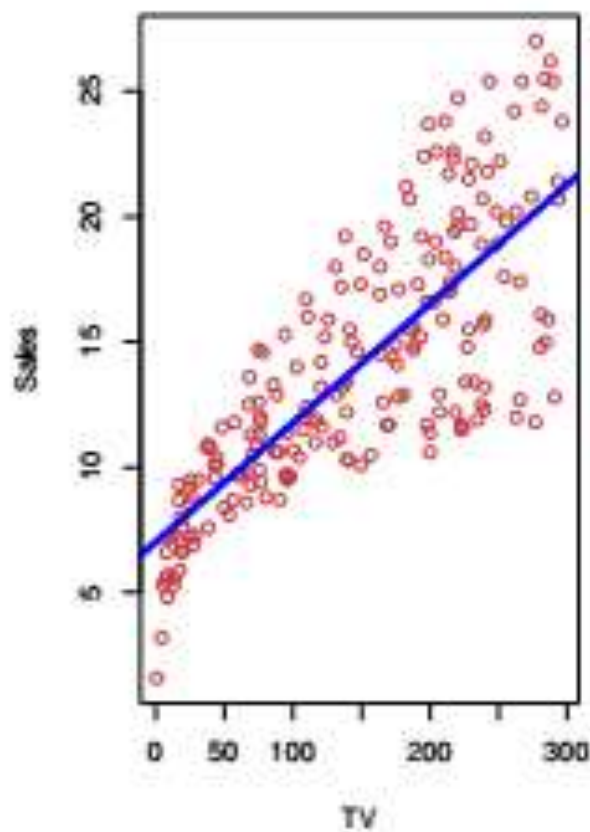
\_\_\_\_\_ Linear Regression  
 \_\_\_\_\_ Actual Regression



# Linear Regression



In Advertising data,  
Predictors: Budget for TV, Radio, News Paper  
Response: Sales





Linear regression answers...

- Is there a relationship between advertising budget and sales?
- How strong is the relationship between advertising budget and sales?
- Which media contribute to sales?
- How accurately can we predict future sales?
- Is the relationship linear?
- Is there synergy among the advertising media?



Predicting a quantitative response  $Y$  on the basis of single predictor variable  $X$

Assumption: there is a linear relationship between  $x$  and  $y$

Model:

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

Diagram illustrating the components of the Simple Linear Regression model equation:

- $Y$ : Sales
- $\beta_0$ : intercept
- $\beta_1$ : slope
- $X$ : TV
- $\epsilon$ : Error term

The intercept and slope coefficients are collectively referred to as **Coefficients / parameters**.

Estimate:  $\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x$

$\hat{\phantom{x}}$  represents estimated term

# Linear Regression

To buy a car

Predictors/features: make, fuel Type, nDoor, engine Size

Response: Price

make	fuelType	nDoors	engineSize	price
alfa-romero	gas	two	130	13495
alfa-romero	gas	two	130	16500
alfa-romero	gas	two	152	16500
audi	gas	four	109	13950
audi	gas	four	136	17450
audi	gas	two	136	15250
audi	gas	four	136	17710
audi	gas	four	136	18920
audi	gas	four	131	23875



Predicting a quantitative response  $Y$  on the basis of single predictor variable  $X$ .

Assumption: there is a linear relationship between  $x$  and  $y$

Model:

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

Price ←  $Y$

←  $\beta_0$  intercept  
Coefficients / parameters

←  $\beta_1$  slope  
Coefficients / parameters

←  $X$  Enginesize

←  $\epsilon$  Error term

Estimate:  $\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x$

^ represents estimated term

Linear regression answers...

- Is price of car price related with engine size?
- How strong is the relationship?
- Is the relationship linear?
- Can we predict/estimate car price based on engine size?





Correlation is a measure of how much the two variables are related.

It is measured by a metric called as the **correlation coefficient**. Its value is between 0 and 1.

# Correlation



- **Correlation** – show whether and how strongly pairs of variables are related.
  - For Example: engine size, price are related; *A decent prediction of price can be made using engine size.*
  - **Correlation can tell you just how much of the variation in products' engine size is related to price.**



## Correlation

- Correlation works for **quantifiable data** (Numerical Data).
- It cannot be used for purely categorical data, such as gender, brands purchased, or favourite colour.
- Correlation is used to understand the relationship between variables such as:
  - Whether the relationship is +ve or -ve
  - the strength of the relationship.
- **Positive correlation:** is a relationship between two variables where **if one variable increases, the other one also increases and vice-versa**
  - Eg: family size , family expenditure will increase or decrease together.

# Correlation

- **Negative Correlation:** there is an inverse relationship between two variables – when one variable decreases, the other increases and vice-versa.
  - **Eg:** negative correlation, between price and demand for goods and services. As the price of goods and services increases, the quantity demanded falls.
- **Coefficient of Correlation (r):** Statistical correlation is measured is known as “coefficient of correlation (r)”. Its numerical value ranges from +1.0 to -1.0. It gives us the strength of relationship.



# Correlation



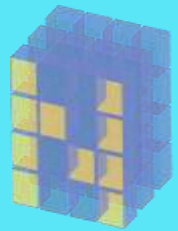
- In general,  $r > 0$  – positive relationship
- $r < 0$  – negative relationship
- $r = 0$  – no relationship (meaning the variables are independent and not related)
- when  $r = +1.0$  – describes a perfect +ve correlation
- when  $r = -1.0$  – describes a perfect -ve correlation
- closer the coefficients are to  $+1.0$  and  $-1.0$ , greater is the strength of the relationship between the variables.

# Pre-requisite for Implementing algorithms

## DATA SCIENCE



IDE



NumPy

Data Analysis



Pandas



Python Data Structures



# DATA VISUALIZATION

Data Science  
Case Studies



MATPLOTLIB

Data Visualization in Python

Seaborn

ANACONDA

