



# Data Science and Artificial Intelligence

## Machine Learning



**Regression**

**Lecture No. 08**



**By- SIDDHARTH SABHARWAL SIR**



# Recap of Previous Lecture



Topic

Multicollinearity

Topic

VIF, Covariance matrix

Topic

Assumption in LR

Topic

Topic



# Topics to be Covered



Topic

Advantage & disadvantage of LR

Topic

Time & Space Complexity of LR.

Topic

Topic

Topic



# About the Faculty

- AIR 1 GATE 2021, 2023 (ECE).
- AIR 3 ESE 2015 ECE.
- M.Tech from IIT Delhi in VLSI.
- Published 2 papers in field of AI-ML.
- Paper 1 : Feature Selection through Minimization of the VC dimension.
- Paper 2 : Learning a hyperplane regressor through a tight bound on the VC dimension.



By- SIDDHARTH SABHARWAL SIR



The background of the book cover is a photograph of a large, dark, jagged rock formation rising from the ocean. The sky is a pale, hazy blue. The text is overlaid on the rock formation.

FAILURE IS  
A BRUISE, NOT  
A TATTOO.

JON SINCLAIR



## Basics of Machine Learning



$$VIF = \frac{1}{1 - R^2}$$

VIF = Large : Multicollinearity exist

VIF = Small : No multicollinearity

Correlation matrix

→ diagonal elements = 1

→ Rest other are close to  $\pm 1$  → Multi

Other elements  $\approx 0$  → No Multi





## Assumptions

→ Homoscedasticity ⇒

• Noise in data  
Shd be independent of  $x$ .

→ No Multicollinearity

→ datapoints Independent

→  $y$  and  $x$  shd have linear Relation



## Considering data of P Dimensions

### Lets Practice

Based on the data provided below, answer questions from (7-10). We consider a function we wish to minimize.

$J(w) = \frac{1}{10} \sum_i (y^{(i)} - w_1 x^{(i)} - w_0)^2$  where the constants  $x^{(i)}$ ,  $y^{(i)}$  are provided in the table below

$i$	$x^{(i)}$	$y^{(i)}$
1	0	1
2	1	3
3	2	5
4	3	8
5	4	9

Dataset

7) The dimension of  $w$  is \_\_\_\_\_.





## Linear Regression

### Considering data of P Dimensions

### Lets Practice

8) Start with the initial guess of  $[w_0, w_1] = [0, 0]$ . Take the value of learning rate = 1. The value of  $w_0$  after 4 iterations of gradient descent will be \_\_\_\_\_.





# Linear Regression



## Space and Time Complexity of Linear Regression

### Time Complexity $\Rightarrow$

Training time Complexity

$\rightarrow$  In LR during training we want to find model, Relationship, Coefficients

$$\beta = (X^T X)^{-1} (X^T Y)$$

$$\begin{bmatrix} A \end{bmatrix}_{m \times n} \begin{bmatrix} B \end{bmatrix}_{n \times p} = \begin{bmatrix} R \end{bmatrix}_{m \times p}$$



$$\begin{array}{c} 1 \\ 2 \\ 3 \\ \vdots \\ m \end{array} \begin{bmatrix} 1 & 2 & 3 & \dots & n \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \end{bmatrix} \begin{array}{c} 1 \\ 2 \\ 3 \\ \vdots \\ n \end{array} \begin{bmatrix} 1 & 2 & 3 & \dots & p \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \end{bmatrix} = \begin{array}{c} 1 \\ 2 \\ 3 \\ \vdots \\ m \end{array} \begin{bmatrix} 1 & 2 & \dots & p \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \end{bmatrix}$$

$m \times n$        $n \times p$        $m \times p$

- To generate each term we multiply  $n$  values
- Total  $m \times p$  values
- ⇒ Total No of mult ⇒  $m \times n \times p$

$p$





# Linear Regression



## Space and Time Complexity of Linear Regression

$$A = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}_{3 \times 3}$$

$$\rightarrow \left. \begin{aligned} M_{11} &= e i - f h \\ M_{12} &= d i - f g \end{aligned} \right\}$$

• for each minor we need 2 mult

$$A^{-1} = \frac{1}{|A|} [Adj]$$

$$Adj = [Cof]^T$$

$$\rightarrow C_{ij} = (-1)^{i+j} \underbrace{M_{ij}}_{\text{Minor}} \rightarrow \text{for each cofactor we need 3 mult}$$

(det of matrix A)

• Cofactor matrix will have  $3 \times 3$  in size  $\Rightarrow 9$  Cofactors  
 $\Rightarrow 3 \times 3 \times 3$  } No of mult for Cof. matrix  
 $\Rightarrow 3^3$

•  $3 \times 3$  matrix inverse  
 $\Rightarrow$  we need  $3 \times 3 \times 3$   
No of mult  $\Rightarrow 3^3$





# Linear Regression



## Space and Time Complexity of Linear Regression

So To find inverse of  $k \times k$  matrix  $\rightarrow k^3$  No of multiplication  
Square

N: data points  
D: No of dimension

$$X = \begin{bmatrix} \end{bmatrix}_{N \times P}$$
$$Y = \begin{bmatrix} \end{bmatrix}_{N \times 1}$$

$$D+1=P$$

$$\beta = \begin{bmatrix} \beta_0 \\ \vdots \\ \beta_D \end{bmatrix}_{P \times 1}$$

$$(X^T X) \Rightarrow (X^T)_{P \times N} (X)_{N \times P} = (X^T X)_{P \times P}$$

$\rightarrow$  No of Mult  $\Rightarrow P \times N \times P = NP^2$

$$(X^T X)_{P \times P} \rightarrow (X^T X)^{-1}$$

$\rightarrow$  No of Mult  $\Rightarrow P^3$

$$X^T Y \Rightarrow (X^T)_{P \times N} (Y)_{N \times 1} \Rightarrow (X^T Y)_{P \times 1}$$

$\rightarrow$  No of mult  $\Rightarrow N \times P \times 1$



$$\beta = (X^T X)^{-1}_{p \times p} (X^T Y)_{p \times 1}$$

$\underbrace{\hspace{10em}}_{\text{No of mult} = p^2}$

Train time  
Complexity.

© Total No of mult  $\Rightarrow (Np^2 + p^3 + Np + p^2)$

• So Time Complexity for Training  $\Rightarrow$  Order of  
 $O[p^3 + Np^2 + Np + p^2]$

\* These are exact number of operations  $\rightarrow$  No

\* So order of means  $\Rightarrow$  Range of No of Calculations.



# Linear Regression

Training  
Data  
 $x, y$

Training

model  
 $\beta$ 's

$$\beta = \begin{bmatrix} \beta_0 \\ \vdots \\ \beta_D \end{bmatrix}_{(D+1) \times 1}$$

Space  
Complexity



After training we only need to store model  
 $\Rightarrow$  i.e store the  $\beta$  values,  $\beta$  matrix  
 $\Rightarrow$  So we need to store  $(D+1)$  values

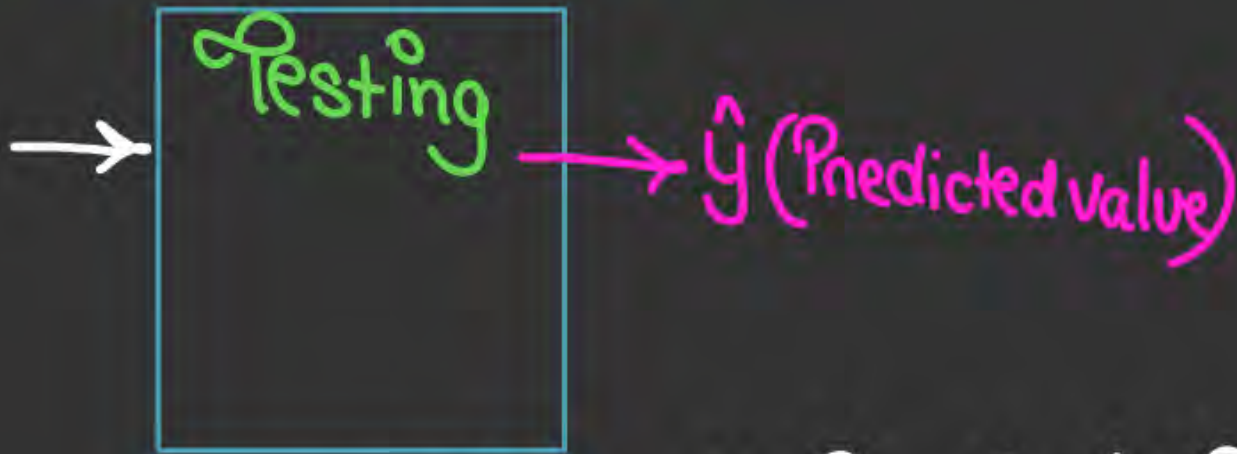
Space Complexity of  
memory used to store model  
 $\Rightarrow$  Order of  $(D+1)$   
 $O(D+1)$



# Linear Regression

Test time  
Complexity

new  
data  
point



model eq  $y = \beta_0 + \beta_1 x^1 + \beta_2 x^2 + \dots + \beta_D x^D$

Now to find  $\hat{y}$  for test point, simply put values of  $x$ 's in eq of model

$$\hat{y} = \beta_0 + \beta_1 x_t^1 + \beta_2 x_t^2 + \dots + \beta_D x_t^D \leftarrow \hat{y} \text{ value}$$

For Testing we  
need D number  
of mult

Test time Complexity =  $O(D)$

Train time  
Comp:  $O(p^3 + Np^2 + Np + p^2)$

Space  
Complexity  $O(D+1)$

Test time  
Complexity  
 $O(D)$



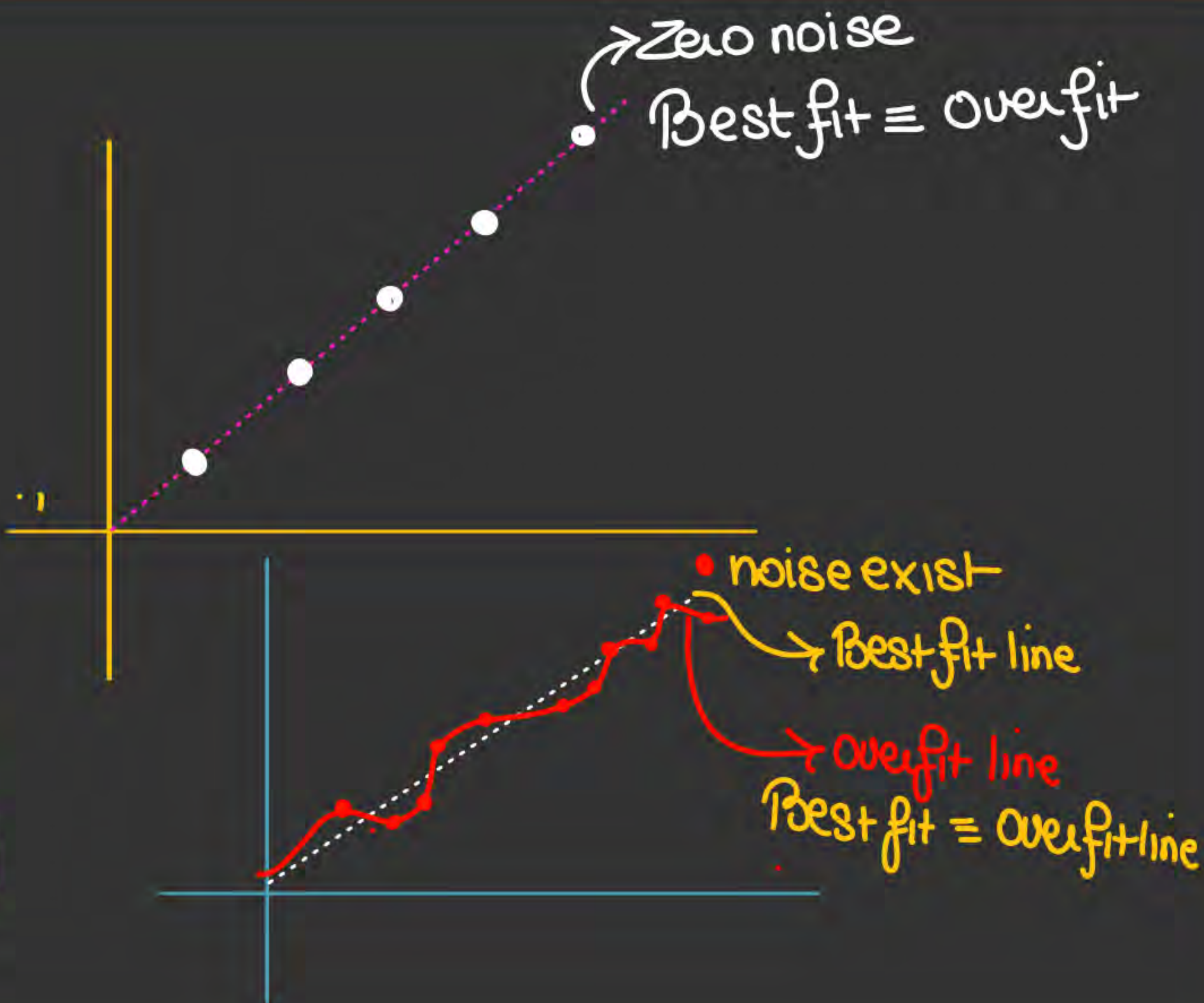
## Disadvantage

1. underfit on the Non linear data
  2. affected by outlier
  3. If dimensions are dependent LR model Perform poorly.
  4. Data shd have Homo-Scedasticity, then only LR algo works.
  5. The LR algorithm Tend to overfit
-

→ data Ko Yaad  
Rattu Tota



→ Badhiya data diya  
not noisy → data  
Pattern





## General Points $\Rightarrow$

- So model that try to provide Best fit  $\Rightarrow$  less/not affected by noisy/outlier points
- model that try to overfit data  $\Rightarrow$  are affected by noisy/outlier points.

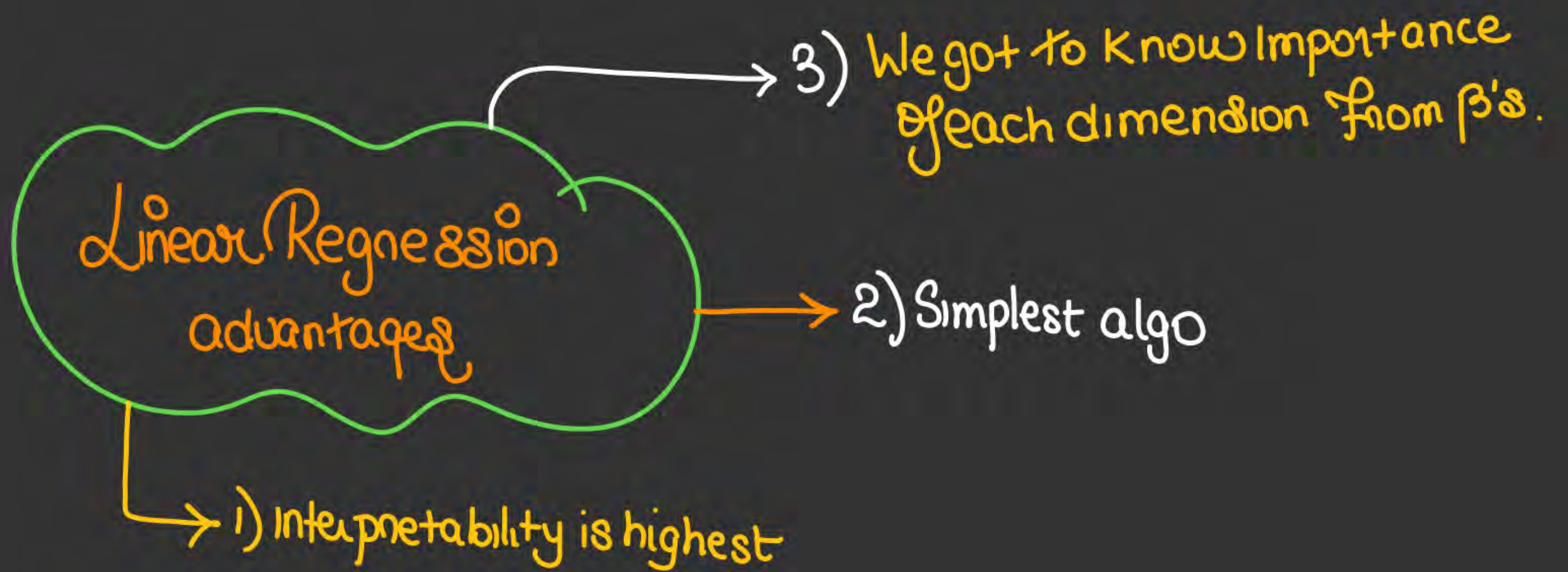
## Linear Regression

only one Task  
 $\Rightarrow$  To (min L)

$$L = \sum_{i=1}^N (y_i - \hat{y}_i)^2$$

$\rightarrow$  So algo has only one Task  $\Rightarrow$  min gap b/w  $y$  and  $\hat{y}$   
 $\Rightarrow$  Want to equate  $y$  and  $\hat{y} \Rightarrow$  Tendency to overfit.









# Linear Regression

## Why Linear Regression is Important

- ✓ The interpretability of linear regression is a notable strength.
- ✓ The model's equation provides clear coefficients that elucidate the impact of each independent variable on the dependent variable, facilitating a deeper understanding of the underlying dynamics.
- ✓ Its simplicity is a virtue, as linear regression is transparent, easy to implement, and serves as a foundational concept for more complex algorithms.





## Linear Regression

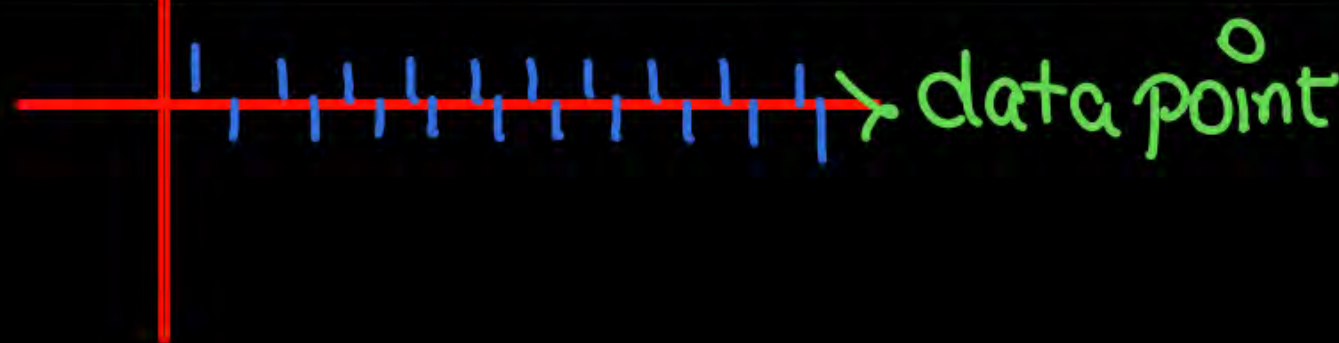
$y$  and  $\hat{y}$

• Overfit  $\Rightarrow$  Residual Plot  $\Rightarrow$  O

What is the purpose of the residual plot in multiple linear regression analysis?

- A) To visualize the relationship between independent variables.
- ☒ B) To check for homoscedasticity and the presence of outliers.
- C) To calculate the correlation coefficient ( $r$ ).
- D) To assess multicollinearity.

$\wedge$  Residue

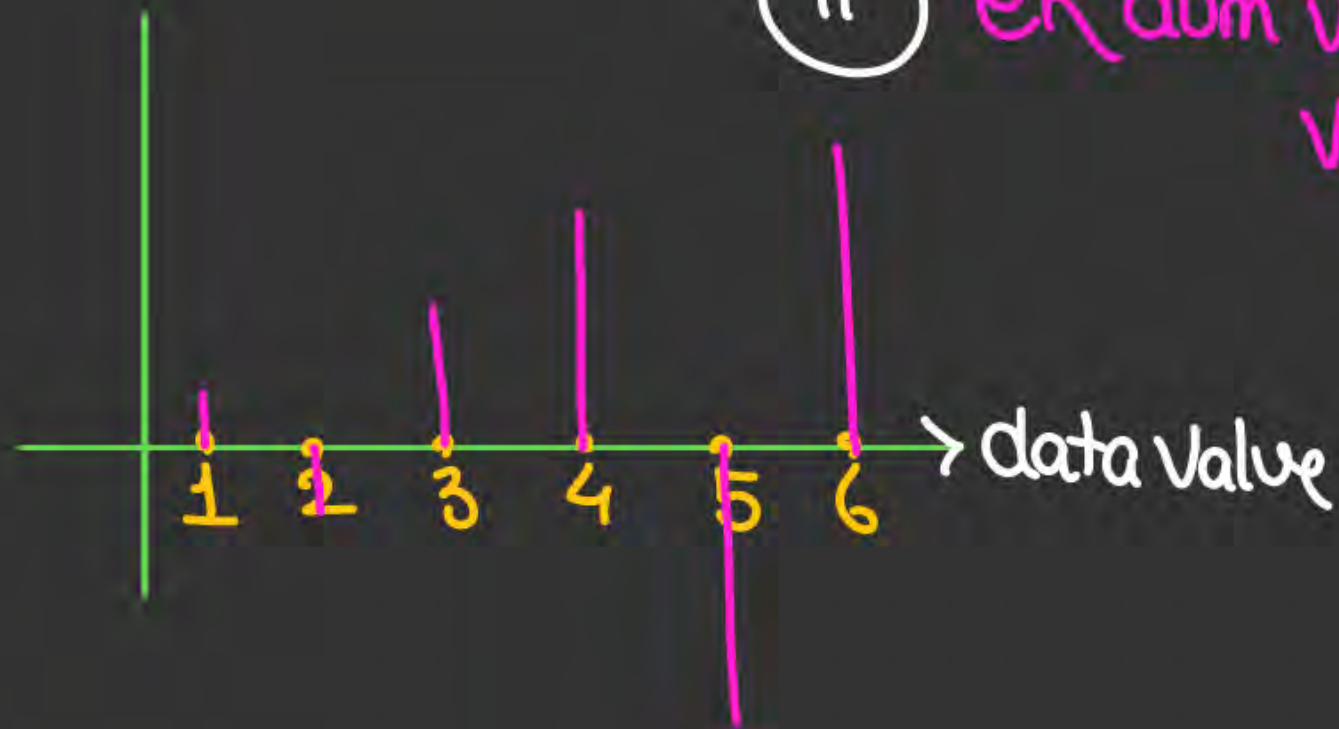


Residue  $\pm$ ve  
 $\Rightarrow$  Similar Nature  $\Rightarrow$   
Homoscedasticity



Residual Plot  $\Rightarrow$  (i)  $\pm$ ve equal small values  $\Rightarrow$  homo...

(ii) ek dum values inc,  
value of Residue inc with  $x \Rightarrow$  hetero...







## Linear Regression

What is the main purpose of the intercept term in a multiple linear regression model?

~~A)~~ It represents the slope of the regression line.

~~B)~~ It is used to control for multicollinearity.

☒ C) It represents the expected value of the dependent variable when all independent variables are zero.

~~D)~~ It is not used in multiple linear regression.

$$y = \beta_0 + \beta_1 x^1 + \dots + \beta_D x^D$$

all ind. variable  $x^1, x^2, \dots, x^D = 0$

$$y = \beta_0$$





## Linear Regression

Question 15: What is the purpose of the coefficient of determination (R-squared) in simple linear regression?

• In LR,  $R^2$  find  
goodness of model

- ☒ A. To determine the slope of the regression line
- ☐ B. To measure the strength of the linear relationship
- ☐ C. To calculate the p-value of the regression
- ☐ D. To identify outliers in the dataset





## Linear Regression

Question 2: What does the coefficient of determination (R-squared) measure in multiple linear regression?



- ~~A.~~ The correlation between predictor variables
- ☒ B. The percentage of variance in the dependent variable explained by the model
- ~~C.~~ The significance of the intercept term
- ~~D.~~ The number of predictor variables in the model

Question 12: In simple linear regression, which variable is considered the independent variable?

 $x$ 

A. The variable being predicted  $\rightarrow y$

B. The response variable  $\rightarrow y$

☒ C. The predictor variable

D. There is no independent variable in simple linear regression



negatively  
Correlated  $\Rightarrow$  if  $x \uparrow, R \downarrow$   
OR  $x \downarrow, R \uparrow$

Positively  
Correlated if  $x \uparrow, R \uparrow$   
 $x \downarrow, R \downarrow$

Question 20: Which of the following statements is true regarding the residual plot in simple linear regression?

- A. Residuals should exhibit a clear linear pattern.
- ☒ B. Residuals should be randomly scattered around the horizontal line.
- ☐ C. Residuals should be negatively correlated with the predictor variable.
- ☐ D. Residuals should have a positive correlation with the dependent variable.

$\rightarrow$  Heteroscedasticity



$X_1$	$X_2$	$X_3$	$X_n$	$Y$
a	b	c	d	$y_1$
e	f	g	h	$y_2$
i	j	k	l	$y_3$
⋮				⋮

→ -ve show if  
 $X_1$  inc  $Y$  dec

5. For a given  $N$  independent input variables ( $X_1, X_2, \dots, X_n$ ) and dependent (target) variable  $Y$  a linear regression is fitted for the best fit line using least square error on this data. The correlation coefficient for one of its variables (say  $X_1$ ) with  $Y$  is  $-0.97$ . Which of the following is true for  $X_1$ ?

- ☒ A) Relation between the  $X_1$  and  $Y$  is weak
- ☒ B) Relation between the  $X_1$  and  $Y$  is strong
- ☐ C) Relation between the  $X_1$  and  $Y$  is neutral
- ☐ D) Correlation does not imply relationship

$\rho \Rightarrow -1 \text{ to } 1$

$0 \approx \text{unCorrelated}$

$\pm 1 \Rightarrow \text{highly Correlated}$



$V_1, V_2 \rightarrow$  neither uncorrelated  
nor highly correlated

6. Given below characteristics which of the following option is the correct for Pearson correlation between  $V_1$  and  $V_2$ ? If you are given the two variables  $V_1$  and  $V_2$  and they are following below two characteristics. 1. If  $V_1$  increases then  $V_2$  also increases 2. If  $V_1$  decreases then  $V_2$  behavior is unknown?

- ☒ A) Pearson correlation will be close to 1
- ☒ B) Pearson correlation will be close to -1
- ☒ C) Pearson correlation will be close to 0  $\rightarrow$  Uncorrelated.
- ☒ D) None of these

- 2) In regression analysis, the variable that is being predicted is;
- a) the independent variable
  - b) **the dependent variable**
  - c) usually denoted by  $x$
  - d) usually denoted by  $r$



- 6) Least square method calculates the best-fitting line for the observed data by minimizing the sum of the squares of the \_\_\_\_\_ deviations.
- ☒ a) **Vertical**
  - b) Horizontal
  - c) Both of these
  - d) None of these

7) Which one is the least square method formula;

a)  $\min \sum (y_i - \hat{y}_i)^2$

h.w b)  $\min \sum (\hat{y}_i - y_i)$

✓ c)  $\min \sum (y_i - \hat{y}_i)^2$

d)  $\min \sum (y_i - \hat{y}_i)$



13) Below you are given a summary of the output from a simple linear regression analysis from a sample of size 15, SSR=100, SST = 152. The coefficient of determination is;

$$R^2 = 1 - \frac{RSS}{TSS} \Rightarrow 0.3421$$

10) A residual is defined as

a) The difference between the actual Y values and the mean of Y.

☒ b) The difference between the actual Y values and the predicted Y values.

c) The predicted value of Y for the average X value.

d) The square root of the slope.

+P.W

11) If the regression equation is equal to  $y=23.6-54.2x$ , then 23.6 is the \_\_\_\_\_ while -54.2 is the \_\_\_\_\_ of the regression line.

a) Slope, intercept

b) Slope, regression coefficient

☒ c) Intercept, slope

d) Radius, intercept



Q8. Suppose we have  $N$  independent variables ( $X_1, X_2 \dots X_n$ ) and  $Y$ 's dependent variable.



Now Imagine that you are applying linear [regression](#) by fitting the best-fit line using the least square error on this data. You found that the correlation coefficient for one of its variables (Say  $X_1$ ) with  $Y$  is  $-0.95$ .

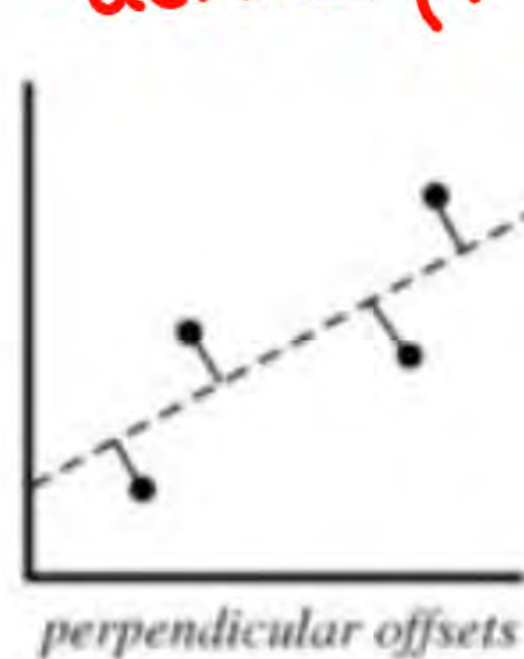
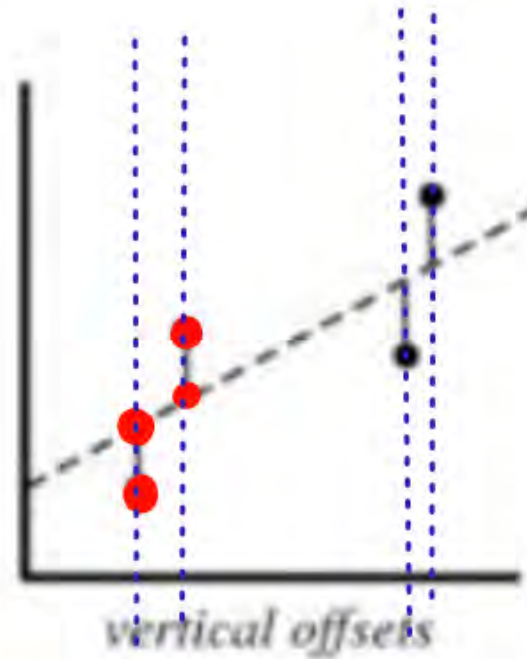
H.W

Which of the following is true for  $X_1$ ?

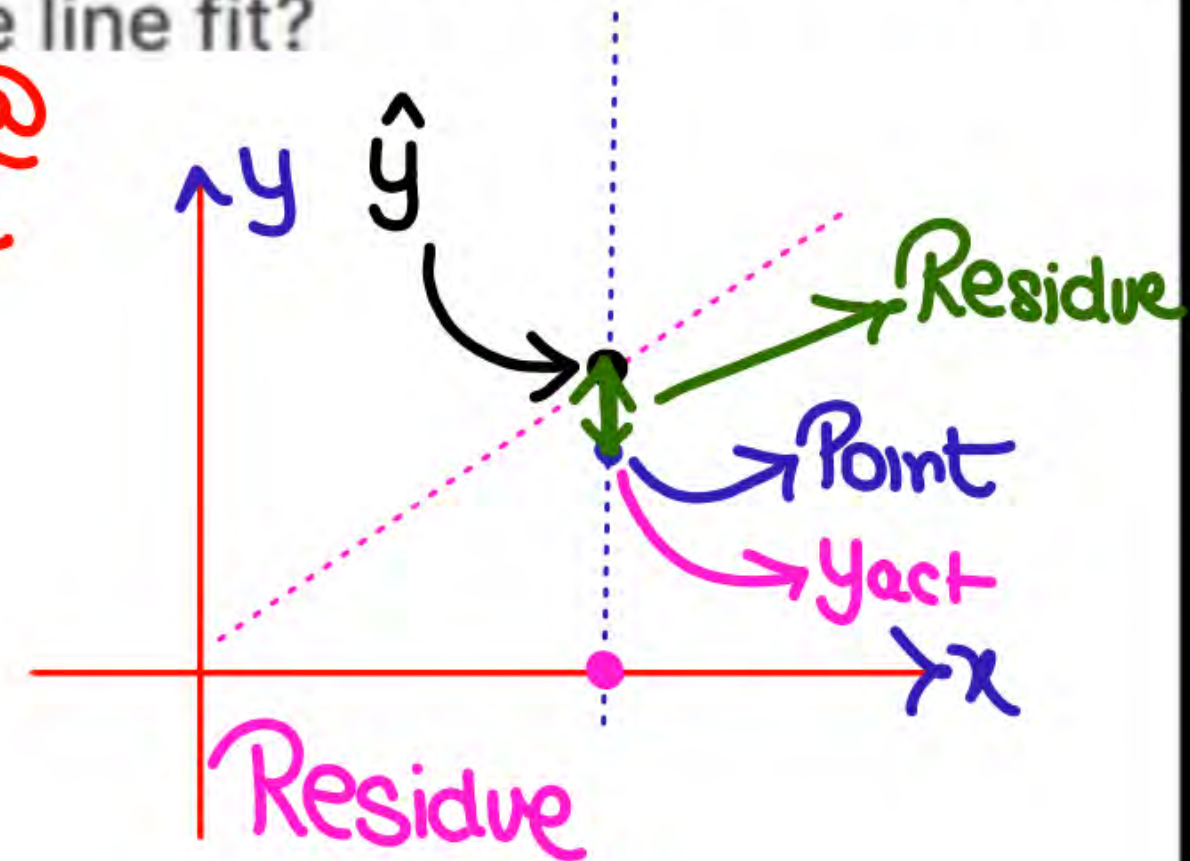
- A) Relation between the  $X_1$  and  $Y$  is weak
- B) Relation between the  $X_1$  and  $Y$  is strong
- C) Relation between the  $X_1$  and  $Y$  is neutral
- D) Correlation can't judge the relationship

**Solution: (B)**

Q11. Suppose the horizontal axis is an independent variable and the vertical axis is a dependent variable. Which of the following offsets do we use in linear regression's least square line fit?



actual & predicted @ same  $x$



~~A) Vertical offset.~~

- B) Perpendicular offset
- C) Both, depending on the situation
- D) None of above



Q12. True- False: Overfitting is more likely when you have a huge amount of data to train.

A) TRUE

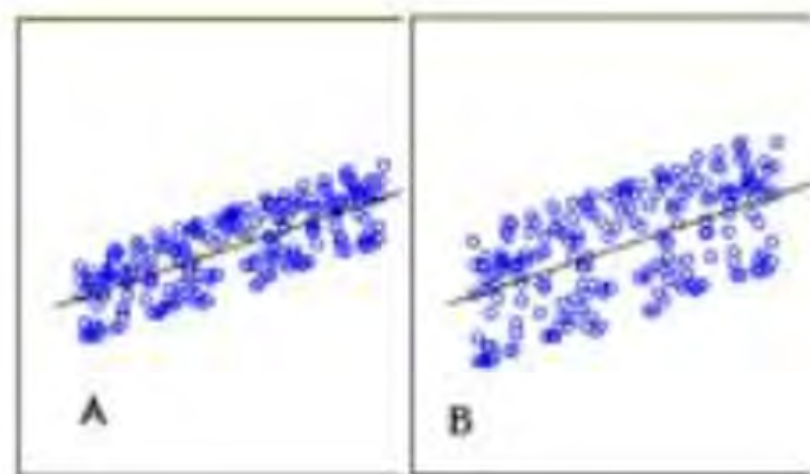
B) FALSE

HPW

**Solution: (B)**

Q14. Which of the following statement is true about the sum of residuals of A and B?

Below graphs show two fitted regression lines (A & B) on randomly generated data. Now, I want to find the sum of residuals in both cases, A and B.



- A) A has a higher sum of residuals than B
- B) A has a lower sum of residual than B
- C) Both have the same sum of residuals
- D) None of these

P.W



Q18. Which of the following statement is true about outliers in Linear regression?

P.W

- A) Linear regression is sensitive to outliers
- B) Linear regression is not sensitive to outliers
- C) Can't say
- D) None of these

Q19. Suppose you plotted a scatter plot between the residuals and predicted values in linear regression and found a relationship between them. Which of the following conclusion do you make about this situation?

HPW

- A) Since there is a relationship means our model is not good
- B) Since there is a relationship means our model is good
- C) Can't say
- D) None of these



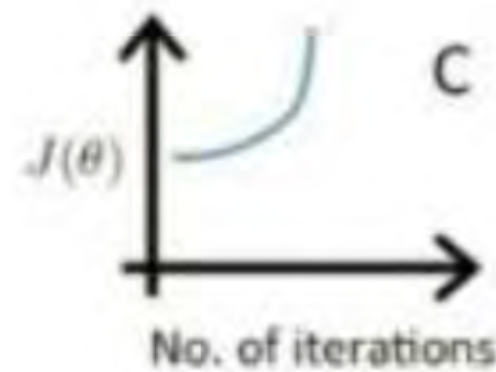
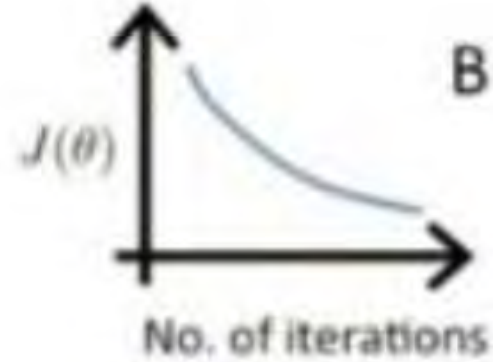
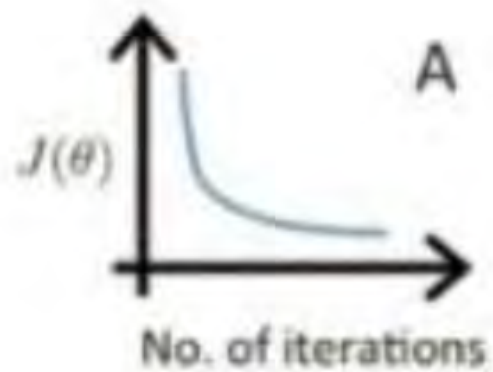
Suppose that you have a dataset  $D_1$  and you design a linear model of degree 3 polynomial and find that the training and testing error is "0" or, in other words, it perfectly fits the data.

Q20. What will happen when you fit a degree 4 polynomial in linear regression?

H.W

- A) There is a high chance that degree 4 polynomial will overfit the data
- B) There is a high chance that degree 4 polynomial will underfit the data
- C) Can't say
- D) None of these

Below are three graphs, A, B, and C, between the cost function and the number of iterations,  $l_1$ ,  $l_2$ , and  $l_3$ , respectively.



f.p.w

Q23. Suppose  $l_1$ ,  $l_2$ , and  $l_3$  are the three learning rates for A, B, and C, respectively. Which of the following is true about  $l_1$ ,  $l_2$ , and  $l_3$ ?

- A)  $l_2 < l_1 < l_3$
- B)  $l_1 > l_2 > l_3$
- C)  $l_1 = l_2 = l_3$
- D) None of these





# Linear Regression



## Considering data of P Dimensions

### Lets Practice

Based on the data provided below, answer questions from (7-10). We consider a function we wish to minimize.

$J(w) = \frac{1}{10} \sum_i (y^{(i)} - w_1 x^{(i)} - w_0)^2$  where the constants  $x^{(i)}, y^{(i)}$  are provided in the table below

$i$	$x^{(i)}$	$y^{(i)}$
1	0	1.4822
2	0.25	1.8165
3	0.50	1.9171
4	0.75	2.3930
5	1.00	2.5826

Dataset

7) The dimension of  $w$  is \_\_\_\_\_.

**THANK - YOU**