



Data Science and Artificial Intelligence

Machine Learning

Regression

Lecture No. 03



By- SIDDHARTH SABHARWAL SIR



Recap of Previous Lecture



Topic	Mean ✓	Covariance
Topic	Variance ✓	
Topic	<div> <div>1D data</div> <div>→</div> <div>$y = mx + c$</div> </div>	
Topic	m, c	
Topic		

Topics to be Covered



Topic

data with more than 1D

Topic

we will see 8+ line for data more than 1D

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.



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Inner
Motivation



***PUSH YOURSELF,
BECAUSE NO ONE ELSE
IS GOING TO DO IT
FOR YOU.***



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1. What is the Loss Function

$$\rightarrow RSS = \sum_{i=1}^N (y_i - y_i')^2$$

\rightarrow we minimize loss $f_{x,n}$.



3. Direct formulae for M and C.

$$(y = mx + c)$$

St. line for 1D data

No of Parameters \Rightarrow
2 Parameters m, c

dimension.

1D data

x	y
x_1	y_1
x_2	y_2
x_3	y_3
\vdots	\vdots

label

$$m = \frac{\text{Cov}(x, y)}{\text{Var } x}$$

$$c = \bar{y} - m\bar{x}$$



4. Covariance :

$$\text{Cov}(x, y) = \underbrace{\sum_{i=1}^N (x_i - \bar{x})(y_i - \bar{y})}_N$$

5. Variance :

$$\text{Var}(x) = \frac{\sum_{i=1}^N (x_i - \bar{x})^2}{N}$$



Linear Regression



6. So the data that we were using has 1
number of dimensions and the straight line
obtained by liner regression has 2
number of parameters.

What is Correlation Coefficient

→ X, Y are 2 variables

$$\rho_{xy} = \frac{\text{Cov}(x, y)}{\text{Std. } x \text{ Std. } y} = \frac{\text{Cov}(x, y)}{\sigma_x \sigma_y}$$

① Value -1 to 1

② Value = ± 1 , then X, Y are highly Correlated

③ Value = 0, X, Y are un Correlated



Representing the two equations in Matrix format

Eq 1

$$\begin{aligned}aw + bz &= c \\ dw + ez &= f\end{aligned}$$

→ In matrix form

$$\begin{bmatrix} a & b \\ d & e \end{bmatrix} \begin{bmatrix} w \\ z \end{bmatrix} = \begin{bmatrix} c \\ f \end{bmatrix}$$

$$\begin{bmatrix} aw + bz \\ dw + ez \end{bmatrix} = \begin{bmatrix} c \\ f \end{bmatrix}$$





Representing the two equations in Matrix format

1D data ✓

N Points

X	Y
x_1	y_1
x_2	y_2
x_3	y_3
\vdots	\vdots

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

Let f_{xn} is $y = mx + c$ for any x , $y'_i = \underline{mx_i + c}$

$$d = \sum_{i=1}^N (y_i - mx_i - c)^2$$

$$d = \sum_{i=1}^N (y_i - mx_i - c)^2$$

No Need to Remember

Steps hein
No need to Remember

$$\downarrow$$

$$\frac{\partial L}{\partial c} \Rightarrow \cancel{2} \sum_{i=1}^N (y_i - mx_i - c) \cancel{(-1)} = 0$$

$$\Rightarrow \sum_{i=1}^N y_i - \sum_{i=1}^N m x_i - \sum_{i=1}^N c = 0$$

$$\Rightarrow \left\{ m \sum_{i=1}^N x_i + c \sum_{i=1}^N 1 = \sum_{i=1}^N y_i \right\}$$

①

$$\frac{\partial L}{\partial m} = \cancel{2} \sum_{i=1}^N (y_i - mx_i - c) \cancel{(-x_i)} = 0$$

$$\sum_{i=1}^N (y_i x_i - m x_i^2 - c x_i) = 0$$

$$\left\{ m \sum_{i=1}^N x_i^2 + c \sum_{i=1}^N x_i = \sum_{i=1}^N y_i x_i \right\}$$

②

$$\Rightarrow \left\{ m \sum_{i=1}^N x_i + c \sum_{i=1}^N 1 = \sum_{i=1}^N y_i \right\}$$

$$\left\{ m \sum_{i=1}^N x_i^2 + c \sum_{i=1}^N x_i = \sum_{i=1}^N y_i x_i \right\}$$

$$\begin{bmatrix} \sum_{i=1}^N 1 & \sum_{i=1}^N x_i \\ \sum_{i=1}^N x_i & \sum_{i=1}^N x_i^2 \end{bmatrix} \begin{bmatrix} c \\ m \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^N y_i \\ \sum_{i=1}^N y_i x_i \end{bmatrix}$$

$X^T X$ $X^T Y$

\Rightarrow (No need to Remember)



Representing the two equations in Matrix format

$y = mx + c$ → function that give Predicted Value

$y = mx + c$ → using matrix must.

Data

$\begin{bmatrix} x \\ x_1 \\ x_2 \\ x_3 \\ \vdots \end{bmatrix}$ $\begin{bmatrix} y \\ y_1 \\ y_2 \\ y_3 \\ \vdots \end{bmatrix}$ → actual values.

$y_1' = mx_1 + c \Rightarrow \begin{bmatrix} 1 & x_1 \end{bmatrix} \begin{bmatrix} c \\ m \end{bmatrix}$
 $y_2' = mx_2 + c \Rightarrow \begin{bmatrix} 1 & x_2 \end{bmatrix} \begin{bmatrix} c \\ m \end{bmatrix}$

1D
data

$$\begin{bmatrix} x \\ x_1 \\ x_2 \\ x_3 \\ \vdots \end{bmatrix}$$

$$\begin{bmatrix} y \\ y_1 \\ y_2 \\ y_3 \\ \vdots \end{bmatrix}$$

$$X = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ 1 & x_3 \\ \vdots & \vdots \end{bmatrix}$$

Column
of 1

$$Y =$$

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \end{bmatrix}$$

Let us have 4 point
1D data

$$\begin{bmatrix} x \\ x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} \quad \begin{bmatrix} y \\ y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix}$$

$$X = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ 1 & x_3 \\ 1 & x_4 \end{bmatrix}$$

$$Y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix}$$

$$X^T = \begin{bmatrix} 1 & 1 & 1 & 1 \\ \underline{x_1} & \underline{x_2} & \underline{x_3} & \underline{x_4} \end{bmatrix}$$

$$X^T X \Rightarrow$$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ x_1 & x_2 & x_3 & x_4 \end{bmatrix} \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ 1 & x_3 \\ 1 & x_4 \end{bmatrix}$$

$$\begin{bmatrix} \sum_{i=1}^4 1 & \sum_{i=1}^4 x_i \\ \sum_{i=1}^4 x_i & \sum_{i=1}^4 x_i^2 \end{bmatrix}$$

Let us have 4 Point
1D data

$$\begin{bmatrix} x & y \\ x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \\ x_4 & y_4 \end{bmatrix}$$

$$X = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ 1 & x_3 \\ 1 & x_4 \end{bmatrix}$$

$$Y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix}$$

$$X^T = \begin{bmatrix} 1 & 1 & 1 & 1 \\ \underline{x_1} & \underline{x_2} & \underline{x_3} & \underline{x_4} \end{bmatrix}$$

$$X^T Y \Rightarrow$$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ x_1 & x_2 & x_3 & x_4 \end{bmatrix}$$

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix}$$

$$\begin{bmatrix} \sum_{i=1}^4 y_i \\ \sum_{i=1}^4 x_i y_i \end{bmatrix}$$

Conclusion

Bas itna hi
dimag mein
Rakho

\Rightarrow

1D data

x y
 x_1 y_1
 x_2 y_2
 x_3 y_3
 \vdots

$$X = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ 1 & x_3 \\ \vdots & \vdots \end{bmatrix}$$

$$Y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \end{bmatrix}$$


model \rightarrow

$$y = \beta_0 + \beta_1 x$$

2 Parameters β_0, β_1

$$\beta = \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix}$$

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

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}_{2 \times 2}$$


$$A^{-1} = \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \frac{1}{ad-bc}$$



Linear Regression

A set of observations of independent variable (x) and the corresponding dependent variable (y) is given below.

x	5	2	4	3
y	16	10	13	12

$y = mx + c$ find m, c

$\frac{\text{Cov}(x, y)}{\text{Var}x}$ Valid

$\bar{y} - m\bar{x}$

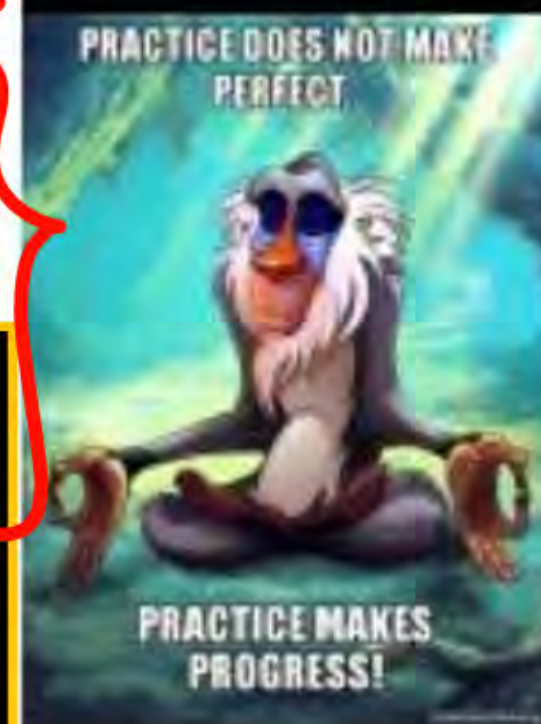
$$X = \begin{bmatrix} 1 & 5 \\ 1 & 2 \\ 1 & 4 \\ 1 & 3 \end{bmatrix}$$

$$Y = \begin{bmatrix} 16 \\ 10 \\ 13 \\ 12 \end{bmatrix}$$

$$X^T = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 5 & 2 & 4 & 3 \end{bmatrix}$$

$$X^T X = \begin{bmatrix} 4 & 14 \\ 14 & 54 \end{bmatrix}$$

$$X^T Y = \begin{bmatrix} 51 \\ 188 \end{bmatrix}$$



$$\beta = \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix} \begin{matrix} C \\ m \end{matrix}$$

$$X^T X (\beta) = (X^T Y)$$

$$\beta = \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix} = (X^T X)^{-1} (X^T Y) \Rightarrow$$

$$\Rightarrow \begin{bmatrix} 54 & -14 \\ -14 & 4 \end{bmatrix} \frac{1}{54 \times 4 - 14^2} \begin{bmatrix} 51 \\ 188 \end{bmatrix}$$

$$\Rightarrow \frac{1}{20} \begin{bmatrix} 54 & -14 \\ -14 & 4 \end{bmatrix} \begin{bmatrix} 51 \\ 188 \end{bmatrix} \Rightarrow$$

$$\frac{1}{20} \begin{bmatrix} 122 \\ 38 \end{bmatrix}$$

$$\beta_0 = C = 122/20$$

$$m = 38/20$$



Linear Regression

1D data ✓

$$y = (-0.4 + 2.4x)$$

$$y = 2 + 2.4(x - 1)$$

For a bivariate data set on (x, y) , if the means, standard deviations and correlation coefficient are

$$\bar{x} = 1.0, \bar{y} = 2.0, s_x = 3.0, s_y = 9.0, r = 0.8$$

Then the regression line of y on x is:

$$1. y = 1 + 2.4(x - 1)$$

$$2. y = 2 + 0.27(x - 1)$$

$$✓✓ y = 2 + 2.4(x - 1)$$

$$4. y = 1 + 0.27(x - 2)$$

$$\bar{x} = 1 \quad \sigma_x = 3$$

$$\bar{y} = 2 \quad \sigma_y = 9$$

$$r_{xy} = 0.8$$

$$\begin{aligned} \beta_0 &= \bar{y} - \beta_1 \bar{x} \\ &= 2 - 2.4(1) \\ &= -0.4 \end{aligned}$$

$$y = \beta_0 + \beta_1 x$$

$$r_{xy} = \frac{\text{Cov}(x, y)}{\sigma_x \sigma_y}$$

$$0.8 = \frac{\text{Cov}(x, y)}{3 \times 9}$$

$$\text{Cov}(x, y) = 9 \times 3 \times 0.8 = 21.6$$

$$\beta_1 = \frac{\text{Cov}(x, y)}{\text{Var}(x)} = \frac{21.6}{\sigma_x^2} = \frac{21.6}{9} = 2.4$$





Linear Regression

In the regression model $(y = a + bx)$ where $\bar{x} = 2.50$, $\bar{y} = 5.50$ and $a = 1.50$ (\bar{x} and \bar{y} denote mean of variables x and y and a is a constant), which one of the following values of parameter ' b ' of the model is correct?

Simple: $a = \bar{y} - b\bar{x}$
 $1.5 = 5.5 - b(2.5)$
 $b = 1.6$

1. 1.75

✓ 2. 1.60

3. 2.00

4. 2.50



Linear Regression

There is no value of x that can simultaneously satisfy both the given equations. Therefore, find the 'least squares error' solution to the two equations, i.e., find the value of x that minimizes the sum of squares of the errors in the two equations. _____

$$2x = 3 \checkmark$$

$$4x = 1$$

$$2x = 3$$

$$4x = 1$$

we can see that single x cannot satisfy both.

error

let x' is value of x that min the sum of square error

$$\begin{cases} \underline{2z=3} \\ \underline{4z=1} \end{cases}$$

Actual
 Actual

$2z'$
 $4z'$

$(z=z')$

$$\text{error} \Rightarrow (2z' - 3)$$

$$\text{error} \Rightarrow (4z' - 1)$$

$$L = (2z' - 3)^2 + (4z' - 1)^2$$

$$\frac{\partial L}{\partial z'} \Rightarrow 2(2z' - 3) \times 2 + 2(4z' - 1) \times 4 = 0$$

$$8z' - 12 + 32z' - 8 = 0$$

$$40z' = 20, z' = .5$$



Linear Regression



We can expect
one
Question from
here in
GATE exam


Q. $5z=1$
 $2z=5$
 $10z=3$

Find z that minimize the sum of square of error in above eq. \Rightarrow

\rightarrow let z' is the correct value
 $L \Rightarrow (5z'-1)^2 + (2z'-5)^2 + (10z'-3)^2$

$\min L \Rightarrow 2(5z'-1) \times 5 + 2(2z'-5) \times 2 + 2(10z'-3) \times 10 = 0$

$50z'-10 + 8z'-20 + 200z'-60 = 0$

$258z' = 90, z' = 90/258 = 15/43$ 



Linear Regression

Attribute/Features

Considering data of 2 Dimensions

Attributes,
Features,
Dimensions...

2 Dimension
2 Attributes.

label y

Income (LPA)	Age	Sale of I-Phone (in a month)
20	30	300
50	40	400
70	50	300

We have N Data points

Now the input data is 2 D (age and income)



Linear Regression

GAATE
2026
FOODENGE



How to write the 2 D inputs ??

Now I want a linear model :

$$y = \beta_0 + \beta_1(\text{Income}) + \beta_2(\text{Age})$$

→ 2D data

→ 2 Attributes data

Parameters ⇒ 3 Parameters.

1D
 $y = \beta_0 + \beta_1 x$



Linear Regression

Linear model will have _____ number of parameters

2D data Representation

data Point	x^1	x^2	label
1	x_1^1	x_1^2	y_1
2	x_2^1	x_2^2	y_2
3	x_3^1	x_3^2	y_3
4	x_4^1	x_4^2	y_4
...			

dimension \rightarrow Super script
datapoint \rightarrow Subscript

model \Rightarrow

$$\Rightarrow (y = \beta_0 + \beta_1 x^1 + \beta_2 x^2)$$



Linear Regression



Linear model will have _____ number of parameters

2D data Representation

data Point	x^1	x^2	y
1	x_1^1	x_1^2	y_1
2	x_2^1	x_2^2	y_2
3	x_3^1	x_3^2	y_3
4	x_4^1	x_4^2	y_4
...			

$$X = \begin{bmatrix} 1 & x_1^1 & x_1^2 \\ 1 & x_2^1 & x_2^2 \\ 1 & x_3^1 & x_3^2 \\ \vdots & \vdots & \vdots \end{bmatrix}_{N \times 3}$$
$$Y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \end{bmatrix}$$
$$\beta = \begin{bmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \end{bmatrix} \quad [X^T]_{3 \times N}$$

$$(X^T X) \beta = (X^T Y) \quad \text{Valid}$$
$$\bullet \beta = (X^T X)^{-1} (X^T Y)$$



Linear Regression

The optimisation method and equation will be ...

3D \Rightarrow

dataPoint	x^1	x^2	x^3	y
1	x_1^1	x_1^2	x_1^3	y_1
2	x_2^1	x_2^2	x_2^3	y_2
3				
...				

$$X = \begin{bmatrix} 1 & x_1^1 & x_1^2 & x_1^3 \\ 1 & x_2^1 & x_2^2 & x_2^3 \\ \vdots & \vdots & \vdots & \vdots \end{bmatrix}_{N \times 4}$$
$$Y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \end{bmatrix}_{N \times 1}$$
$$\beta = \begin{bmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix}$$
$$\underline{\underline{\beta = (X^T X)^{-1} (X^T Y)}} \text{ (6)}$$

THANK - YOU