

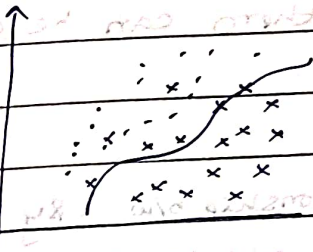
## Linear Regression, MSE and polynomial regression

classmate

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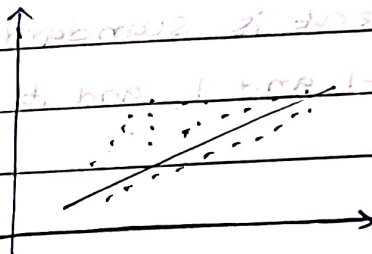
## \* classification vs Regression

classification: Predicts a discrete value/class label.



Ex: spam filtering, Image classification.

Regression: Type of ML that predicts a continuous value.



Ex: a house's [Area, age] (x) vs its Price (y)

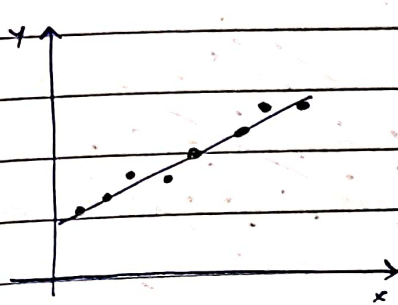
\* Covariance from PCA

→ Covariance tells us about the amount of dependency between two variables.

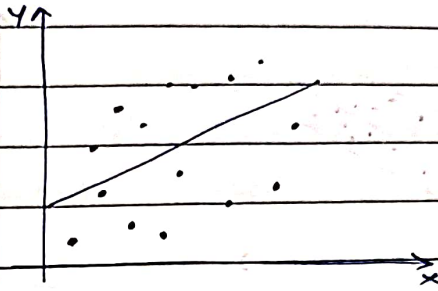
$$\text{Cov}(x, y) = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$$

 $\text{Cov}(x, y) > 0 \rightarrow x$  and  $y$  are positively related $\text{Cov}(x, y) < 0 \rightarrow x$  and  $y$  are inversely related $\text{Cov}(x, y) = 0 \rightarrow x$  and  $y$  are independent

# \* Scatter Plots of Data with various correlation coefficients



$$r = +1$$

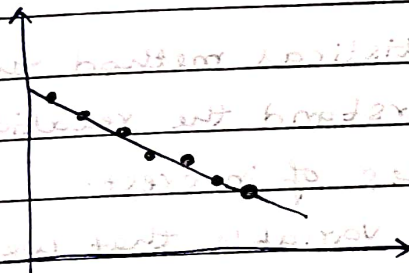


$$r = +0.3$$

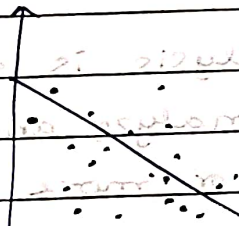
- relation is there
- increasing only for some values
- relationship is not obvious here; and not strong.

y value is constant no matter how much x changes.

no relation b/w x & y



$$r = -1$$

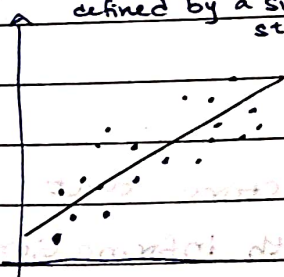


$$r = -0.6$$

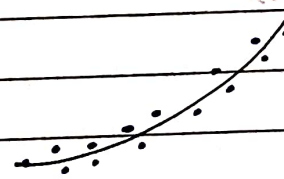
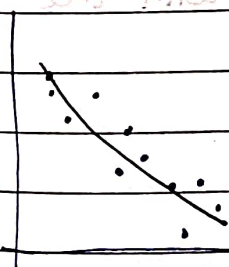
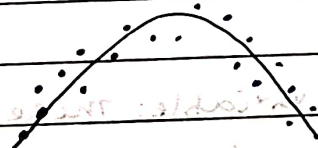
## \* linear correlation

Relationship:

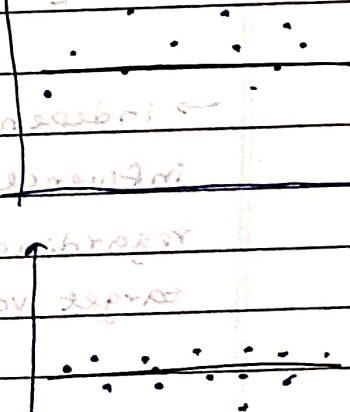
Linear defined by a simple st line.



Curvilinear

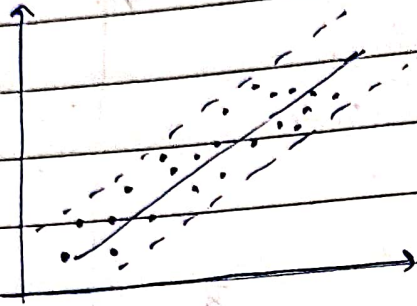


No

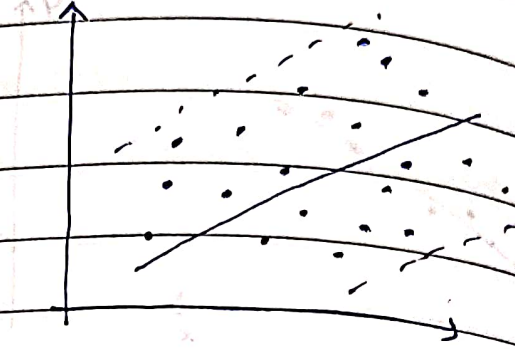




strong relationships



weak relationships



## \* Regression Analysis

→ The two variables  $(x_i, y_i)$  are treated as equals correlation.

→ Regression analysis is a statistical method that helps us to analyze and understand the relationship between two or more variables of interest.

→ Dependent Variable: This is the variable that we are trying to forecast ( $y$ )

TV	Radio	Newspaper	( $y$ ) Sales
230.1	37.8	69.2	22.1
44.5	39.3	45.1	10.4
17.2	45.9	69.3	12

→ Independent Variable: These are the factors that influence the analysis and provide us with information regarding the relationship of the variables with the target variable ( $x$ ).

## \* Correlation Coefficient

→ Assuming a linear relationship between the variables, the relative strength between them can be observed.

→ Normalization & correlation

↓  
compute relationship b/w  $x$  &  $y$

but your values will be standardized  
(dependency range  $[-1, 1]$ )

→ Pearson's correlation coefficient is standardized covariance ranging between  $-1$  and  $1$ , and it is unitless.

$$r = \frac{\text{cov}(x, y)}{\sqrt{\text{var}_x} \sqrt{\text{var}_y}}$$

- $r = 1$  Perfect +ve linear correlation
- $r = -1$  Perfect -ve linear correlation
- $r = 0$  No linear correlation

(we assumed in PCA;

all variables when they are related

⇒ consider it as linear relationship.



## \* Linear Regression

→ Linear regression is a predictive model used for finding the linear relationship b/w a dependent variable and one or more independent variables.

$$Y = mx + b$$

↓  
dependent  
variable

→  
independent  
variable

→ Y-intercept

Slope

(gradient, determines  
changes in Y, per unit  
change in X)

NOTE:

Correlation does not imply causation  
 $x$  is increasing because  $y$  is increasing.  $X \rightarrow Y$  NOT TRUE

Even if two variables are correlated, it does not necessarily mean that changes in one variable cause changes in the other.

Ex: In summers; Swimming & ice-creams