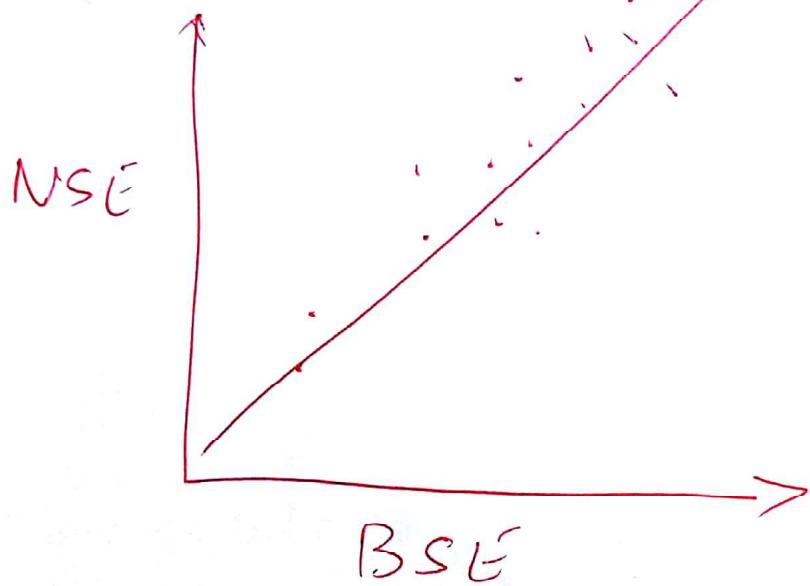


Covariance and correlation

→ They form a part of bivariate relationship analysis.



- Look at the data points
 - ↳ Pattern looks like linear relationship
- When one rises, other rises in the same way.
- In this case, relationship is strong.

In other words, there is a strong positive relationship.

We can quantify this relationship via covariance i.e. how they co-vary.

Note: Covariance is only one measure to quantify this idea.

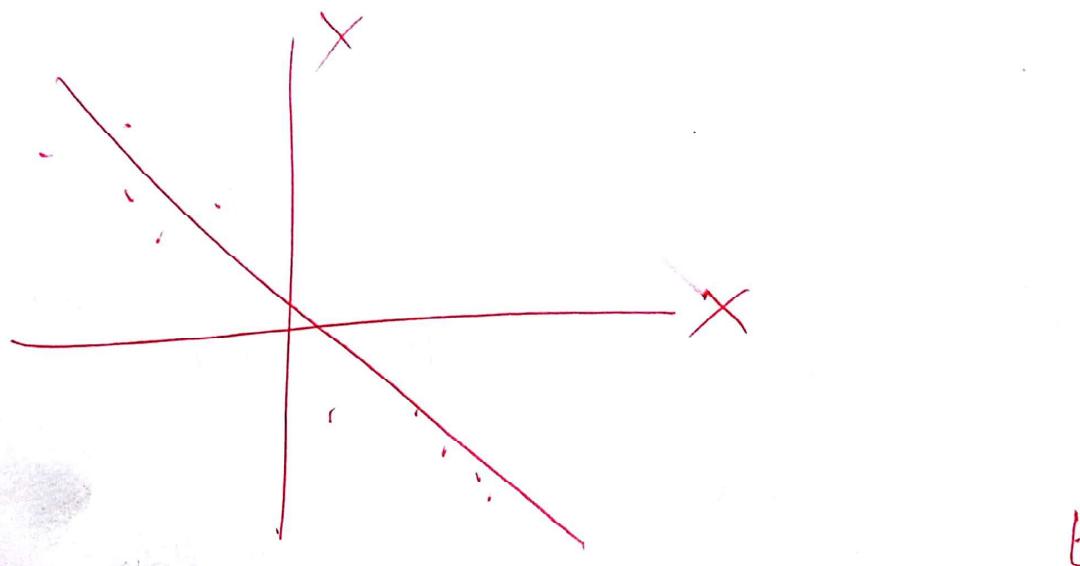
Others include: correlation.

Covariance: A descriptive measure of the relationships bw two variables.

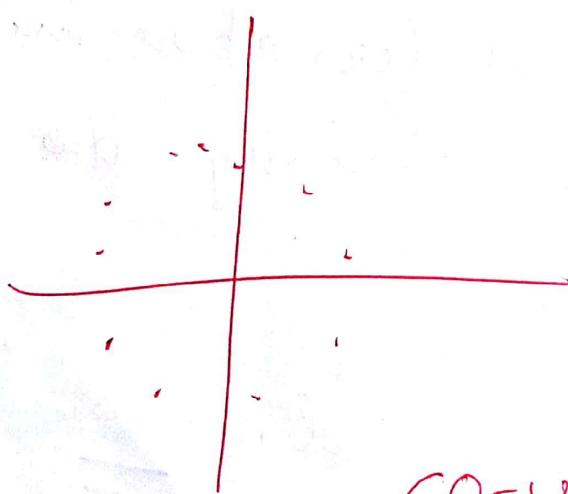
Emphasis is on direction.

the indicator increasing R/p
-ve " decreasing R/p

Note Covariance is used to identify linear relationship



-ve covariance



co-variance = 0

$$\text{cov}(X, Y) = E[(X - E[X])(Y - E[Y])]$$

Covariance:

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

for sample

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

For population:

Example: DC: workers

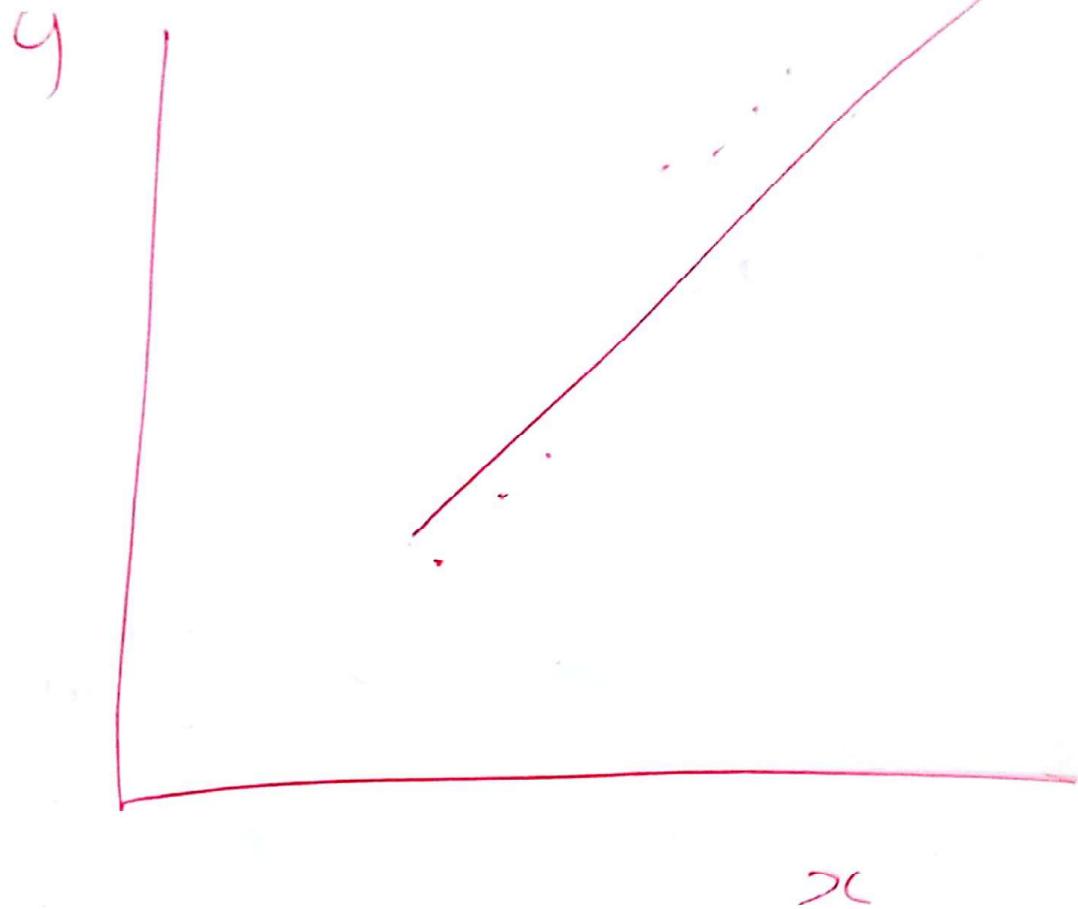
10 random samples from new buildings

y: no. of walls painted

		$\bar{x} = 21.3$	$\bar{y} = 41.2$	$\sum (x_i - \bar{x})(y_i - \bar{y})$
		x_i	y_i	
12		20	-9.3	18.8
30		60	8.7	-14.2
15		27	-6.3	8.8
24		50	2.7	-20.2
14		21	-7.3	-11.2
18		30	-3.3	19.8
28		61	6.7	12.8
26		54	4.7	-9.2
19		32	-2.3	15.8
27		57	5.7	
				$\sum = 962.4$

$$\text{Cov}(x,y) = S_{xy} = \frac{962.4}{10-1}$$

$$= 106.93$$



Positive linear relationship

Focus only on sign

Correlation

Covariance provides the direction

Correlation provide direction and strength.

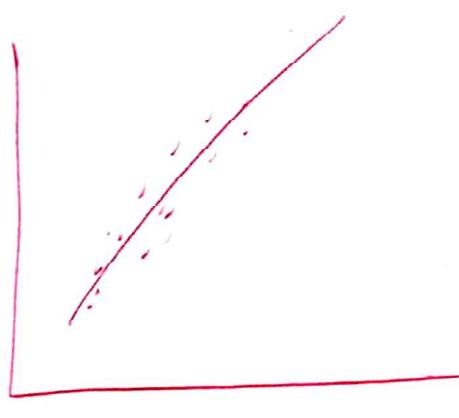
Covariance \rightarrow No upper and lower bound.

Correlation \rightarrow Between (-1, 1).

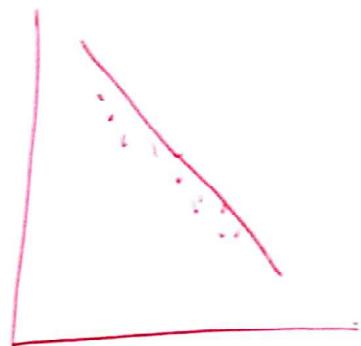
Covariance is dependent on scale.

Correlation is independent of scale.

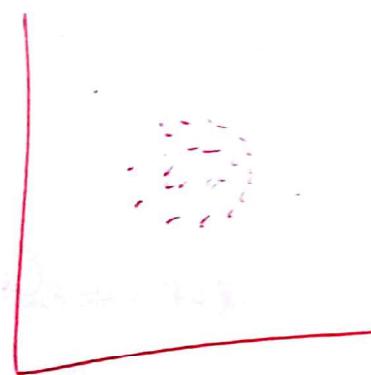
Drawback: Covariance and correlation are applicable to linear relationships.



Near +1



Close to -1



Close to 0

Correlation Coefficient
(ρ)

$$\rho_{xy} \text{ (or } \rho_{xy}) = \frac{\text{cov}(X, Y)}{\sigma_x \sigma_y}$$

$$\text{cov}(x, y) = E[(x - \mu)(y - \mu)]$$

From wikipedia -

If (x_i, y_i) (for $i=1 \dots n$) can be realized with probability p_i , then

$$\text{cov}(x, y) = \sum_{i=1}^n p_i (x_i - E[x])(y_i - E[y])$$

$$\text{cov}(x, a) = 0$$

$$\text{cov}(x, x) = \text{var}(x)$$

$$\text{cov}(x, y) = \text{cov}(y, x).$$

$$\text{cov}(ax + by) = ab \text{cov}(x, y)$$

$$\text{cov}(x, y) = E[xy] - E[x]E[y].$$

$$P_{xy} = \frac{\text{cov}(X, Y)}{\sigma_x \sigma_y}$$

~~BxSxy σx σy~~

$$= \frac{106.93}{6.48 \times 16.69} = 0.989.$$

Rule of thumb

$$\text{if } |r| > \frac{2}{\sqrt{n}}$$

then relation exists

Not Universal

Important point:
correlation does not imply
causation.

$$X = [x_1 \ x_2 \ x_3]$$

$\text{cov}(X) \rightarrow$ covariance matrix

$$\begin{matrix} & x_1 & x_2 & x_3 \end{matrix}$$

$$\begin{matrix} x_1 & \text{var}(x_1) & \text{cov}(x_1, x_2) & \text{cov}(x_1, x_3) \end{matrix}$$

$$\begin{matrix} & x_2 & \text{cov}(x_2, x_1) & \text{var}(x_2) & \text{cov}(x_2, x_3) \end{matrix}$$

$$\begin{matrix} x_3 & \text{cov}(x_3, x_1) & \text{cov}(x_3, x_2) & \text{var}(x_3) \end{matrix}$$

Suppose we have a random variable $X \sim N(\mu, \sigma^2)$.

To convert this normal distribution into standard normal, we follow the standardizing procedure i.e.

$$Z = \frac{X - \mu}{\sigma}$$

Here $Z \sim N(0, 1)$.

E.g. $X \sim N(162.2, \sigma = 6.8)$

Say what is the prob. that

an average girl is taller than 170.5

$$P(X > 170.5)$$

$$P\left(\frac{X - 162.5}{6.8} > \frac{170 - 162.5}{6.8}\right)$$

$$P(Z > 1.22)$$

Discussion on reading 2-table

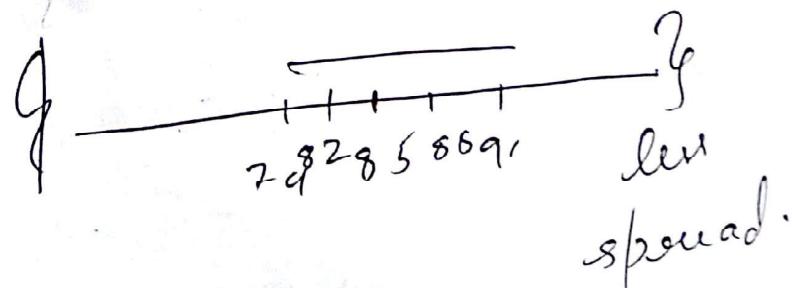
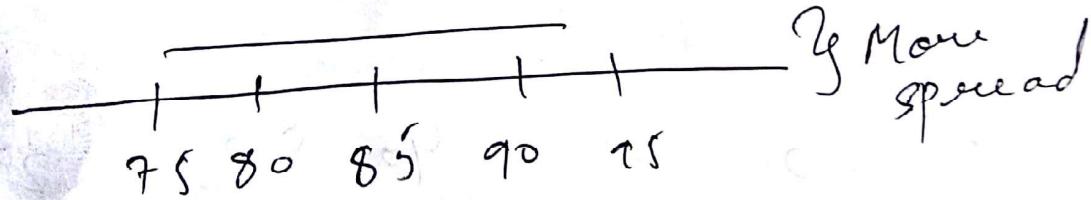
z-score

Suppose, we have the following data

85	88
95	79
75	91
80	85
90	82

$$\mu = 85$$

$$\mu = 85$$



z-score is a statistical measure that describes how far a given data point is from the mean.
↳ It describes the distance in term standard deviation.

- Q) Suppose a data sample has mean 50 and σ of 5. A value of 55 will have z-score of 1, and 40 will have z-score of -2.

$$Z = \frac{x - \mu}{\sigma}$$
$$= \frac{55 - 50}{5} = 1$$

$$Z = \frac{x - \mu}{\sigma} = \frac{40 - 50}{5} = -2$$

" Suppose in a test a person scored 630 marks. Mean marks is 500. S.D is 150.

$$z = \frac{x - \mu}{\sigma} = \frac{630 - 500}{150}$$

$$= \frac{130}{150} = 0.83$$

Q) ICSE board a person had got 87% with others having an average 80% with S.D. of 5.

CBSE board a person got 82% with class avg. 73 and s.d. of

$$z_1 = \frac{87 - 80}{5} ; z_2 = \frac{82 - 73}{8}$$

$$z_1 = 1.4 ; z_2 = 1.125$$

Suppose IQ of people follow normal distribution with $\mu = 100$ and $\sigma = 15$.

What % of people are i) stupid
ii) smart
iii) average.

stupid: 85 or less

smart: 140 or more

avg : ~~50~~-140

$$\Phi(\text{Stupid}) = \frac{85 - 100}{15} = -1.033$$

$$= 0.5 - 0.004$$

$$= 0.5 - 0.491$$

$$\begin{aligned}\text{smart} : \frac{140 - 100}{15} &= \frac{40}{15} = 2.66 \\ &= 0.9961\end{aligned}$$

$$1 - 0.9961$$

$$= 0.0039$$

$$= 0.4\%$$