

* Covariance

eg

| X | Y |
|-----|--------|
| Age | weight |
| 14 | 40 |
| 15 | 45 |
| 18 | 61 |
| 20 | 68 |
| 25 | 74 |

If $\begin{matrix} X \uparrow & Y \uparrow \\ X \downarrow & Y \downarrow \end{matrix} \left. \vphantom{\begin{matrix} X \uparrow \\ X \downarrow \end{matrix}} \right\} +ve \text{ covariance}$

If $\begin{matrix} X \uparrow & Y \downarrow \\ X \downarrow & Y \uparrow \end{matrix} \left. \vphantom{\begin{matrix} X \uparrow \\ X \downarrow \end{matrix}} \right\} -ve \text{ covariance}$

Covariance

population

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

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

Eg-

Economic growth
(%)

nitly (soy).
index

2.1

8

2.5

12

3.6

10

4.0

14

calculation

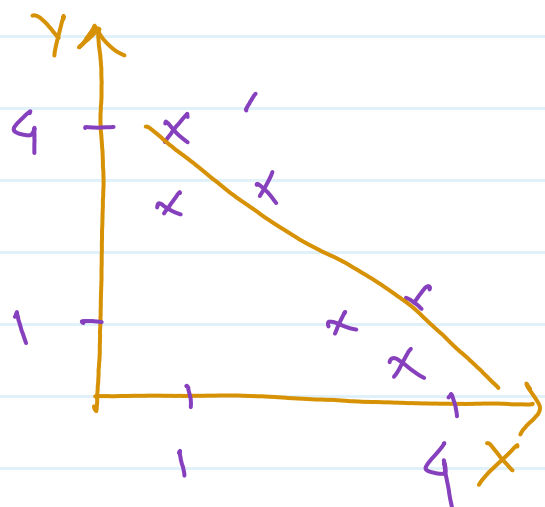
| x | y | \bar{x} | \bar{y} | $(x - \bar{x})$ | $(y - \bar{y})$ |
|-----|-----|-----------|-----------|-----------------|-----------------|
| 2.1 | 8 | | | -1 | -3 |
| 2.5 | 12 | 3.05 | 11 | -0.6 | 1 |
| 3.6 | 10 | | | 0.5 | -1 |
| 4.0 | 14 | | | 0.9 | 3 |

$$\text{Cov}(x, y) = \frac{(-1)(-3) + (-0.6)(1) + (0.5)(-1) + (0.9)(3)}{4-1}$$

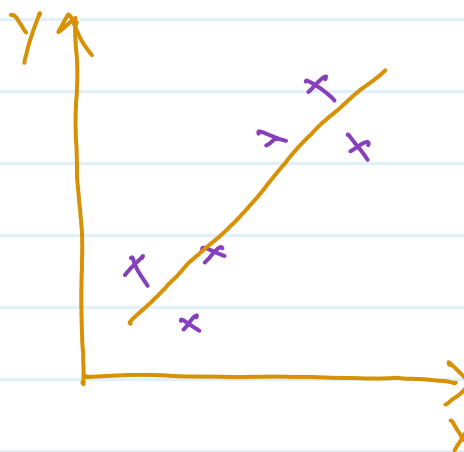
$$= \frac{4.6}{3} = \underline{\underline{1.533}}$$

Covariance -1000 to $+1000$

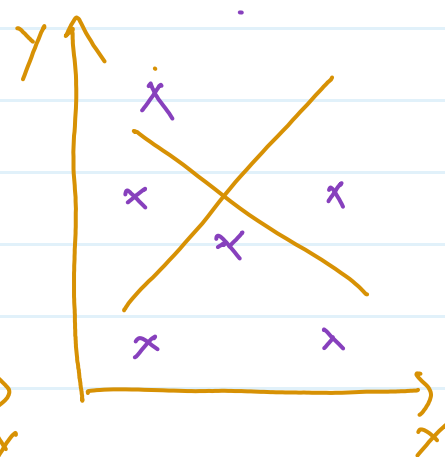
Zero means no covariance b/w x and y



-ve



+ve



Zero

Diff. b/w covariance and co-relation
in covariance we can find
relation b/w x and y but can't
say about its strength

In co-relation we can find relation
b/w x and y as well its strength
which is calculate by -1 to $+1$

* Pearson correlation coefficient

Eg.

| X growth | Y Nifty50 |
|-------------|--------------|
| 2.1 | 8 |
| 2.5 | 12 |
| 3.6 | 10 |
| 4.0 | 14 |

$$\Rightarrow \rho_{(X,Y)} = \frac{\text{Cov}(X,Y)}{\sigma_X * \sigma_Y}$$

standard devi.

$$\sigma_X = \sqrt{\frac{(-1)^2 + (-0.6)^2 + (0.9)^2 + (0.5)^2}{4-1}}$$

$$\sigma_X = \sqrt{0.8060} = 0.89$$

$$\sigma_Y = \sqrt{\frac{(-3)^2 + (1)^2 + (3)^2 + (-1)^2}{4-1}}$$

$$\sigma_Y = \sqrt{6.66} = \underline{2.58}$$

$$\rho_{(x,y)} = \frac{1.533}{(0.89)(2.58)}$$

$$\Rightarrow 0.66$$

$$= \underline{\underline{66\%}}$$

$\overline{x \uparrow \quad y \uparrow}$
 $x \downarrow \quad y \downarrow \quad \left. \vphantom{\overline{x \uparrow \quad y \uparrow}} \right\} +ve$

$x \uparrow \quad y \downarrow$
 $x \downarrow \quad y \uparrow \quad \left. \vphantom{x \uparrow \quad y \downarrow} \right\} -ve$

BMI

weight height
 x_1 x_2
 └──────────┘
 feature column



Independent column
 variable

BMI

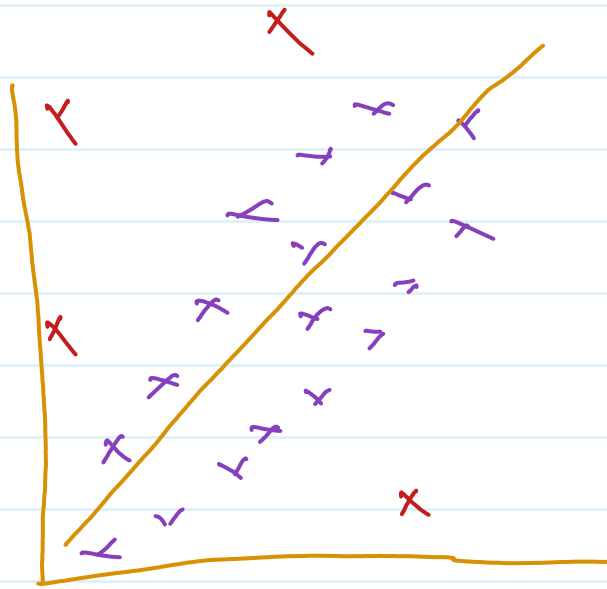
y
 └──┘
 Target column



Dependent column
 variable

Disadvantages of P.C.C. is,

It is able to capture linear property but is not able to capture non-linear property.

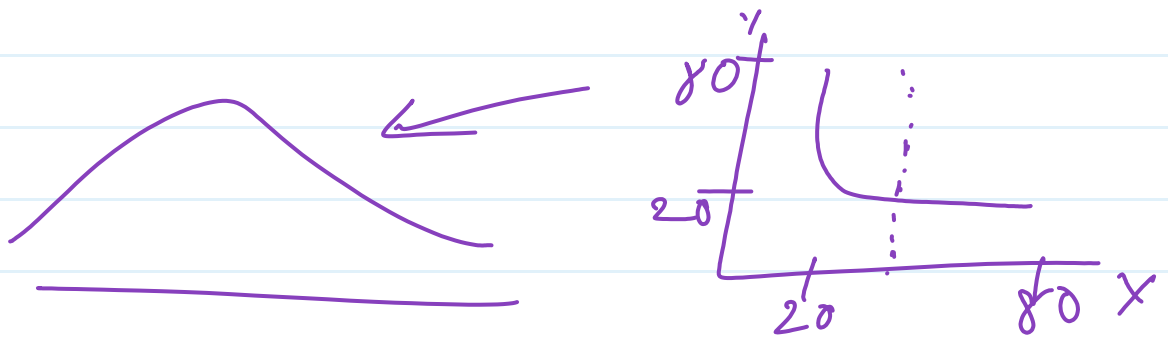


* Spearman Rank correlation

* Power law Dist.

80-20
pareto Dist.

- 80% Run — 20% Batter
- 80% work — 20% employee.
- 80% family income — 20% parent



To convert pareto dist. into Gaussian Dist. by "box-Cox transformation"

$$y_i = \frac{x_i^\alpha - 1}{\alpha}$$

α is weight of Dist

if $\lambda \neq 0$, Box-cox transformation

if $\lambda = 0$, Log transformation