

= COVARIANCE =

$$\text{Cov}(X, Y) = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{N} \rightarrow \text{population covariance}$$

$$\text{Cov}(X, Y) = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{n-1} \rightarrow \text{sample covariance}$$

Another formula:

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

Significance:

- Covariance +ve \Rightarrow X, Y increase/decrease together
- Covariance -ve \Rightarrow As X increases, Y decreases and vice versa
- Covariance = 0 \Rightarrow X and Y are uncorrelated

Units:

(X units)(Y units)

EXAMPLE:

	X = -1	X = 0	X = 1	P(Y=E)
Y = -1	1/15	2/15	2/15	1/3
Y = 0	2/15	1/15	2/15	1/3
Y = 1	2/15	2/15	1/15	1/3
P(X=E)	5/15 = 1/3	1/3	1/3	

$$E(X) = E(Y) = 0$$

$$\begin{aligned} \text{Cov}(X, Y) &= E[(X - E(X))(Y - E(Y))] \\ &= (-1)(-1)\left(\frac{1}{15}\right) + (-1)(1)\left(\frac{2}{15}\right) \\ &\quad + (1)(-1)\left(\frac{2}{15}\right) + (1)(1)\left(\frac{1}{15}\right) \end{aligned}$$

$$= \frac{1}{15} - \frac{2}{15} - \frac{2}{15} + \frac{1}{15}$$

$$= -\frac{2}{15}$$

Properties

- $\text{Cov}(X, X) = \text{Var}(X)$
- $\text{Cov}(X, Y) = E(XY) - E(X)E(Y)$
- Covariance is symmetric
 $\text{Cov}(X, Y) = \text{Cov}(Y, X)$

Covariance Matrix

$$\Sigma = \begin{bmatrix} \sigma_x^2 & \sigma_{xy} & \sigma_{xz} \\ \sigma_{xy} & \sigma_y^2 & \sigma_{yz} \\ \sigma_{xz} & \sigma_{yz} & \sigma_z^2 \end{bmatrix}$$

multivariate normal

$$\phi(x) = \frac{1}{\sqrt{(2\pi)^n |\Sigma|}} e^{-\frac{1}{2}(\bar{x} - \bar{\mu})^T \Sigma^{-1}(\bar{x} - \bar{\mu})}$$

$$\sum E[(X - E(X))(Y - E(Y))]$$

$$\sum_{i=1}^n X_i \cdot P(X = x_i)$$

④ linear qty.

$$\text{Cov}(X, aY + bZ) = a \text{Cov}(X, Y) + b \text{Cov}(X, Z)$$

$$\text{Cov}(aX + bY, Z) = a \text{Cov}(X, Z) + b \text{Cov}(Y, Z)$$

= COVARIANCE & INDEPENDENCE =

- X and Y are independent $\Rightarrow X$ and Y are uncorrelated
($\text{Cov}(X, Y) = 0$)
- Converse need not be true.
Uncorrelated X, Y may still be dependent.
- Being independent is a much stronger notion of being unrelated compared to being uncorrelated

= CORRELATION COEFFICIENT =

Correlation coefficient \rightarrow normalised version of covariance $\rightarrow \rho / r$

$$\rho(X, Y) = \frac{\text{Cov}(X, Y)}{\text{SD}(X) \text{SD}(Y)}$$

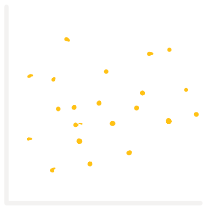
Properties

- $-\text{SD}(X) \text{SD}(Y) \leq \text{Cov}(X, Y) \leq \text{SD}(X) \text{SD}(Y)$
- $E \left[\left(\frac{X - E(X)}{\text{SD}(X)} + \frac{Y - E(Y)}{\text{SD}(Y)} \right)^2 \right] \geq 0 \rightarrow$ lower bound
- $E \left[\left(\frac{X - E(X)}{\text{SD}(X)} - \frac{Y - E(Y)}{\text{SD}(Y)} \right)^2 \right] \geq 0 \rightarrow$ upper bound
- $-1 \leq \rho(X, Y) \leq 1$
- Summarises trend b/w two random variables
- Dimensionless qty

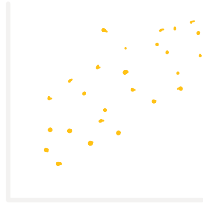
Results:

- ① $\rho(X, Y)$ close to 0
 $\Rightarrow X, Y$ close to being uncorrelated; no clear trend
- ② $\rho(X, Y) = 1 / \rho(X, Y) = -1$
 \Rightarrow There exists $a \neq 0, b$ such that $Y = aX + b$ with probability 1.
 Y is a linear function of X .
- ③ $|\rho(X, Y)| \rightarrow 1$
 $\Rightarrow X, Y$ have strong correlation; increase in X likely to match up with increase in Y .

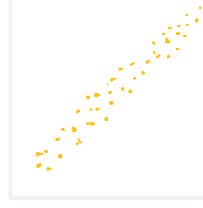
= SCATTER PLOTS =



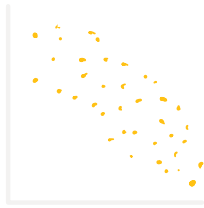
$\rho \approx 0$



$\rho \rightarrow +ve$



$\rho \approx +1$



$\rho \rightarrow -ve$



$\rho \approx -1$