

Indep $\Rightarrow \text{Cov}(X, Y) = 0$

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$$\text{Cov}(X, X) = \text{Var} X$$

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

$$\text{Cov}\left(\sum_i a_i X_i + b, \sum_j c_j Y_j + d\right) =$$

$$= \sum_{i,j} a_i c_j \text{Cov}(X_i, Y_j)$$

$$\text{Var} \sum_i X_i = \sum_i \text{Var} X_i + 2 \sum_{i < j} \text{Cov}(X_i, X_j)$$

$$|E[XY]| \leq \sqrt{E X^2 \cdot E Y^2}$$

$$-1 \leq \rho(X, Y) = \frac{\text{Cov}(X, Y)}{\text{SD} X \cdot \text{SD} Y} \leq 1$$

Correlation

$$X_i = \begin{cases} 1 & \text{if } A_i \\ 0 & \text{if } A_i^c \end{cases} \quad \text{indicator of } A_i$$

$$X = \sum_{i=1}^n X_i = \# \text{ of the } A_i \text{ that occur}$$

$$\mathbb{E} X = \sum_{i=1}^n \mathbb{E} X_i$$

$$\boxed{\mathbb{E} X_i = X_i} - \\ \parallel - \mathbb{E} X_i \cdot \mathbb{E} X$$

$$\text{Var } X = \sum_{i=1}^n \text{Var } X_i + 2 \sum_{1 \leq i < j \leq n} \text{Cov}(X_i, X_j)$$

$$\mathbb{E} X_i = P(A_i); \text{Var } X_i = P(A_i)(1 - P(A_i))$$

Conditional expectations

$$\mathbb{E}(X | Y = y) = \sum_i x_i \cdot P(x_i | y)$$

$\mathbb{E}(X | Y)$ is a fct of Y

is a r.v.

Tower rule: $\mathbb{E} \mathbb{E}(X | Y) = \mathbb{E} X$

$$\mathbb{E} \mathbb{E} g(X|Y) = \mathbb{E} g(X)$$

$$\text{Var} \mathbb{E}(X|Y) + \mathbb{E} \text{Var}(X|Y) = \text{Var} X$$

$$\mathbb{E}(R(X) \cdot g(Y) | Y) = g(Y) \mathbb{E}(R(X) | Y)$$

$$\mathbb{E} \sum_{i=1}^{N \leftarrow \text{Rand.}} X_i = \mathbb{E} \mathbb{E} \left(\sum_{i=1}^N X_i \mid N \right) =$$
$$= \mathbb{E} \left[\sum_{i=1}^N \mathbb{E}(X_i | N) \right]$$

$$M(t) = \mathbb{E} e^{tX}$$

$$M(0) = 1$$

$$M^{[n]}(0) = \mathbb{E} X^n$$

\uparrow
nth derivative

hoping $M(t) < \infty$
by some interval
of t 's around
0.

Unique

If X, Y are indep., then

$$M_{X+Y}(t) = M_X(t) M_Y(t)$$

⑦ Markov's ineq.

$X \geq 0$ r.v.

$\forall a > 0$

$$P(X \geq a) \leq \frac{E X}{a}$$

Chebyshev's: If $\text{Var } X \leq \sigma^2 < \infty$

$$P(|X - \mu| \geq b) \leq \frac{\sigma^2}{b^2}$$

\downarrow
 $E X$

$$\begin{aligned} E(g(Y) \cdot R(X)) &= E(E(g(Y) \cdot R(X) | Y)) \\ &= E(g(Y) \cdot E(R(X) | Y)) \\ &= E(g(Y) R(X)) \end{aligned}$$

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Thm WLLN: If X_1, X_2, \dots iid $\forall \epsilon > 0$.

$$P\left(\left|\frac{X_1 + \dots + X_n}{n} - \mu\right| \geq \epsilon\right) \xrightarrow{n \rightarrow \infty} 0$$

\downarrow
 $E X_i$

Thm: CLT : X_1, \dots iid $E X_i = \mu$ $\text{Var} X_i = \sigma^2$

$$P\left(a < \frac{X_1 + \dots + X_n - n\mu}{\sqrt{n}\sigma} \leq b\right) \xrightarrow{n \rightarrow \infty} \Phi(b) - \Phi(a)$$

iid = independent, identically distributed

DeMoivre-Laplace:

$X \sim \text{Binom}(n, p)$

fixed $n \rightarrow \infty$

$$P\left(a < \frac{X - np}{\sqrt{np(1-p)}} \leq b\right) \xrightarrow{n \rightarrow \infty} \Phi(b) - \Phi(a)$$

Beispiel:

$$X \sim \text{Binom}(n, p)$$

$$\stackrel{||}{=} \sum_{i=1}^n X_i$$

\downarrow $\begin{cases} 1 & \text{if trial} \\ & \text{succeeds} \\ 0 & \text{else} \end{cases}$