

VARIANCE

STAT 510

DEFN THE VARIANCE OF  $X$ , WITH  $\mu = E[X]$  IS

$$\sigma^2 = E[(X-\mu)^2] = \begin{cases} \sum_x (x-\mu)^2 f(x) & X \text{ DISC} \\ \int_x f(x) dx & X \text{ CONT} \end{cases}$$

$\rightarrow \sigma_x^2 = V[X]$

THE STANDARD DEVIATION IS  $\sigma = \sqrt{\sigma^2}$

$$\sigma = \sqrt{\sigma_x^2} = SD[X]$$

Thm

THE VARIANCE HAS THREE PROPERTIES:

$$(1) \quad V[X] = E[X^2] - (E[X])^2$$

$$(2) \quad \text{GIVEN CONSTANTS } a, b : V[aX + b] = a^2 V[X]$$

(3)  $X_1, \dots, X_n$  INDEPENDENT,  $a_1, \dots, a_n$  CONSTANTS THEN

$$V\left[\sum_{i=1}^n a_i X_i\right] = \sum_{i=1}^n a_i^2 V[X_i]$$

# EXAMPLE

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

$$V[X] = \sum_{x=0}^n (x - np)^2 \binom{n}{x} p^x (1-p)^{n-x} = \dots$$

AGAIN  $X = \sum_{i=1}^n X_i$   $X_i \xrightarrow{\text{BERNOULLI}}$  NO  $\text{RECALL } E[X_i] = p$

$$E[X_i^2] = 0 \cdot (1-p) + 1^2 \cdot p = p$$

$$V[X_i] = E[X_i^2] - (E[X_i])^2 = p - p^2 = p(1-p)$$

$$V[X] = V\left[\sum_{i=1}^n X_i\right] = \sum_{i=1}^n V[X_i] = \sum_{i=1}^n p(1-p) = np(1-p) \checkmark$$

Thm

$X_1, X_2, \dots, X_n$  i.i.d. with  $E[X_i] = \mu$  AND  $V[X_i] = \sigma^2$

①  $E[\bar{X}_n] = \mu$  WHERE  $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$

②  $V[\bar{X}_n] = \sigma^2/n$

SAMPLE MEAN

③  $E[S_n^2] = \sigma^2$  WHERE  $S_n^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2$

SAMPLE  
VARIANCE

DEFN

X WITH MEAN  $\mu_X$  AND VARIANCE  $\sigma_X^2$

Y WITH MEAN  $\mu_Y$  AND VARIANCE  $\sigma_Y^2$

THE COVARIANCE OF X AND Y IS

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

THE CORRELATION OF X AND Y IS

$$\rho = \rho_{X, Y} = \rho(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{\sigma_X^2 \sigma_Y^2}}$$

# Thm

$$(1) \quad \text{Cov}[X, Y] = \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y]$$

$$(2) \quad -1 \leq \rho(X, Y) \leq 1$$

$$(3) \quad \text{IF } X, Y \text{ IND THEN } \text{Cov}[X, Y] = \rho = 0.$$

THE REVERSE IS NOT TRUE IN GENERAL.

Thm

$$V[X+Y] = V[X] + V[Y] + 2\text{Cov}[X, Y]$$

OR, IN GENERAL

$$V\left[\sum_{i=1}^n a_i X_i\right] = \sum_{i=1}^n \underline{a_i^2 V[X_i]} + 2 \sum_{i < j} \underline{a_i a_j \text{Cov}[X_i, X_j]}$$



# VARIANCE - COVARIANCE MATRIX

$$X = \begin{pmatrix} X_1 \\ \vdots \\ X_k \end{pmatrix}$$

$$\mu = E[X] = \begin{pmatrix} \mu_1 \\ \vdots \\ \mu_k \end{pmatrix} = \begin{pmatrix} E[X_1] \\ \vdots \\ E[X_k] \end{pmatrix}$$

$$\Sigma = V[X] =$$

$$\begin{bmatrix} V(X_1) & \text{Cov}(X_1, X_2) & \dots & \text{Cov}(X_1, X_k) \\ \text{Cov}(X_2, X_1) & V(X_2) & \dots & \text{Cov}(X_2, X_k) \\ \vdots & \vdots & \ddots & \vdots \\ \text{Cov}(X_k, X_1) & \dots & \dots & V(X_k) \end{bmatrix}$$