

# Review

"Random sample"

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

$$E[\bar{x}] = \mu$$

"Statistic"

$$x_i \sim N(\mu, \sigma)$$

$$\text{Var}[\bar{x}] = \frac{\sigma^2}{n}$$

Sample mean & Normal RV's.

$$\bar{x} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

• Properties of statistics

• Parameters of a distribution

$$z = \frac{\bar{x} - \mu}{\left(\frac{\sigma^2}{n}\right)^{\frac{1}{2}}}$$

$$z \sim N(0, 1)$$

# How to Choose Parameters

$\theta$  parameter

$\hat{\theta} = T(X_1, \dots, X_n)$  estimate

$\bar{X} = \frac{1}{n} \sum X_i$  sample mean.

# Estimators

"Bias" of an estimator

$$E[\hat{\theta}] - \theta$$

Unbiased estimator has zero bias

# Unbiased Estimators

Example - Sample Mean.

# Example - Sample Variance $X \sim \sigma^2, \mu$

$$\text{Sample var } S^2 = \frac{1}{n-1} \sum_i (X_i - \bar{X})^2 = \frac{1}{n-1} \sum_i X_i^2 - 2 \sum_i X_i \bar{X} + \sum_i \bar{X}^2$$

$$E[S^2] = \frac{1}{n-1} \sum_{i=1}^n (E[X_i^2] - 2E[X_i \bar{X}] + E[\bar{X}^2])$$

$$\text{Var}[Y] = E[Y^2] - E[Y]^2$$

$$E[Y^2] = \text{Var}[Y] + E[Y]^2$$

- ①  $\text{Var}[X_i] + E[X_i]^2 = \sigma^2 + \mu^2$
- ②  $\text{Cov}[X_i, \bar{X}] + E[X_i]E[\bar{X}] = \mu^2$
- ③  $\text{Var}[\bar{X}] + E[\bar{X}]^2 = \frac{\sigma^2}{n} + \mu^2$

$$\begin{aligned}
 &= \frac{1}{n-1} \sum_{i=1}^n \left( \sigma^2 + \mu^2 - 2\mu^2 + \frac{\sigma^2}{n} + \mu^2 \right) \\
 &= \frac{n}{n-1} \left( \sigma^2 + \mu^2 - 2\mu^2 + \frac{\sigma^2}{n} + \mu^2 \right) \\
 &= \frac{n}{n-1} \left( \frac{n}{n} \sigma^2 - \frac{\sigma^2}{n} \right) = \left( \frac{n-1}{n-1} \right) \sigma^2 = \sigma^2
 \end{aligned}$$

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

$$S^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2$$

unbiased

$$E[S^2] - \sigma^2 = \sigma^2 - \sigma^2 = 0$$

Example - Bernoulli

$$P = \frac{1}{2}$$

$(x_1, \dots, x_n)$

$$\frac{1}{n} \sum_{i=1}^n x_i = \hat{p}$$

$$E\left[\frac{1}{n} \sum_{i=1}^n x_i\right] = \frac{1}{n} \sum_{i=1}^n E[x_i] = \frac{1}{n} \sum_{i=1}^n p = p$$



## Deciding Between Estimators.

$$\hat{\theta}_1 = T_1[x_1 \dots x_2]$$

$$\hat{\theta}_2 = T_2[x_1 \dots x_2]$$

unbiased - both

choose

$$\text{Var}(\hat{\theta}_1) < \text{Var}(\hat{\theta}_2)$$