

Statistics and Sampling Distributions

In a typical statistical problem, we have a random variable whose underlying probability distribution is known, but its parameters are unknown; e.g.,

$$X \sim \text{Exp}(\mu)$$

$$X \sim N(\mu, \sigma^2)$$

$$X \sim \text{Poisson}(\lambda)$$

* We represent the parameters of these distributions by θ ; e.g.,

$$X \sim \text{Exp}(\theta)$$

$$X \sim N(\theta_1, \theta_2)$$

$$X \sim \text{Poisson}(\theta)$$

* We call θ a parameter of the distribution.

* The goal is to estimate the unknown θ based on the available data or observations.

* A statistic (singular) is a function of the observations which summarizes (reduces) the data.

* Let us assume that x_1, x_2, \dots, x_n ^{sample size} $\sim f(x; \theta)$ read as "parametrized by θ "

The sequence of random variables x_1, x_2, \dots, x_n is called a sample, where each of the random variables is called observation ①

- * Once the sample is realized, we denote the numerical values taken by each observation using lower-case Roman letters.

$$x_1, x_2, \dots, x_n$$



these are realized observations once the sample is drawn

- * Note that a statistic is always a function of the samples:

$$T = g(x_1, x_2, \dots, x_n)$$

e.g.,

$$\bar{X} = \frac{x_1 + x_2 + \dots + x_n}{n}$$

$$X_{(n)} = \max(x_1, x_2, \dots, x_n)$$

- * Estimator vs. Estimate:

While we call T an estimator of θ , we call its realization t an estimate of θ .

- * Once the sample is drawn, then t is called the realization of the statistics T, where

$$t = g(x_1, x_2, \dots, x_n)$$

Note: A sample is random if $x_1, x_2, \dots, x_n \stackrel{iid}{\sim} f(x; \theta)$

Order Statistics

* Let X_1, X_2, \dots, X_n be a random sample from the distribution $f(x; \theta)$.

Then, $X_{(1)}, X_{(2)}, \dots, X_{(n)}$ represents the order statistic associated to the random sample.

Q. Find the sampling distribution of maximum order statistics.

$$\text{Let } Y = \max(X_1, X_2, \dots, X_n)$$

$$F_Y(y) = P(Y \leq y)$$

$$= P(\max(X_1, X_2, \dots, X_n) \leq y)$$

$$= P(X_1 \leq y, X_2 \leq y, \dots, X_n \leq y)$$

$$\stackrel{\text{independence}}{=} P(X_1 \leq y) P(X_2 \leq y) \dots P(X_n \leq y)$$

$$= F_X(y) \cdot F_X(y) \dots F_X(y)$$

$$= (F_X(y))^n$$

∴

$$F_Y(y) = (F_X(y))^n$$

$$f_Y(y) = n f_X(y) (F_X(y))^{n-1}$$

Exp. $X_1, X_2, \dots, X_n \stackrel{\text{iid}}{\sim} \text{Uniform}(0, 1)$

$$f_Y(y) = ny^{n-1}$$

alternatively
↓

$$f_{X_{(n)}}(x) = nx^{n-1}$$