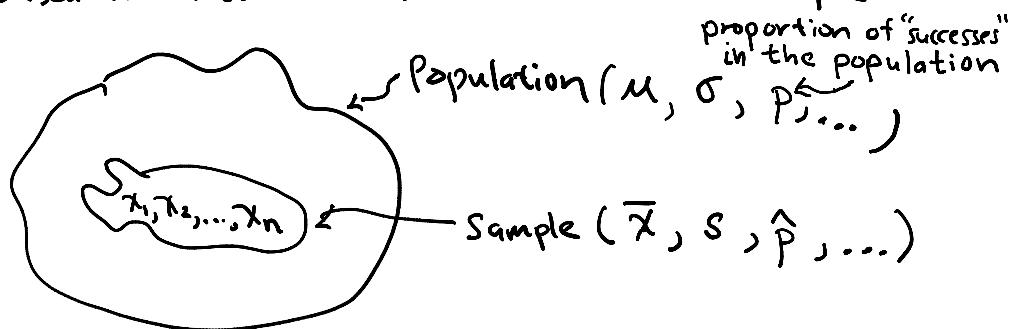


Chapter 6 (continued)

02/22/2023

Sampling Distributions and Estimators

One of the goals of statistical analysis is to make inferences about population characteristics based on information contained in a sample.



NOTE: Values of population mean (μ), standard deviation (σ), and proportion (p) are considered fixed values but unknown.

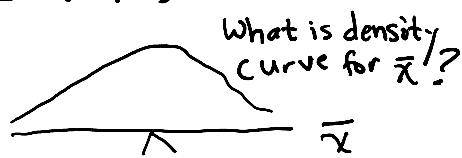
Consider using the sample mean $\bar{x} = \frac{\sum_{i=1}^n x_i}{n}$ to estimate the population mean μ .

\bar{x} is often called a statistic. It is a quantity computed from the sample. μ is called a population parameter.

We use statistics to estimate parameters.

The sample mean, \bar{x} , varies from sample-to-sample and thus has a sampling distribution.

Def : The sampling distribution of a statistic is a probability or density function that summarizes all of the possible values of the statistic.



Ex : x = blood platelet size (fl) of patient with non-cardiac chest pain.

NOTE: $1m^3 = 1 \times 10^{18} fl$

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

What is the sampling distribution of \bar{x} ?

$$n=5$$

From random Sample 1, compute \bar{x} .

" " Sample 2, compute \bar{x} .

⋮

From random sample 500, compute \bar{x} .

} Construct the histogram of the sampling distribution of \bar{x} .

Repeat above process with $n=10, n=20, n=30$.

Comment on the results. See Canvas Handout #3.

Ex : x = blood transfusion recipient stricken with Viral hepatitis. 02/27/2023

$$= \begin{cases} 1, & \text{if Yes} \\ 0, & \text{if No} \end{cases}$$

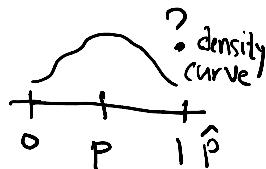
p = population proportion of blood recipients stricken with viral hepatitis.

\hat{p} = sample proportion " " " "

$$\begin{aligned}
 p &= \text{population proportion} && \text{viral hepatitis.} \\
 \hat{p} &= \text{sample proportion} && " " " " \\
 &= \frac{\text{number of Yes's in sample}}{n} \\
 &= \frac{\sum_{i=1}^n x_i}{n}
 \end{aligned}$$

What is the sampling distribution of \hat{p} ?

Suppose $p = 0.07$.



$n = 10$

From random sample 1, compute \hat{p}

⋮

From random sample 500, compute \hat{p} .

Repeat process for $n = 25, n = 50, n = 100$.

Describe sampling distribution of \hat{p} .

See Handout #3 in Canvas.

General Properties of Sampling Distributions

1. Sampling distribution of a statistic is often centered at the value of the population parameter.
2. The variability of the sampling distribution of the statistic decreases as n increases.
3. As n increases, sampling distribution of many statistics become more and more bell-shaped.

The Sampling Distribution of \bar{X} and the Central Limit Theorem (CLT)

Result : If X_1, X_2, \dots, X_n is a random sample from a population with mean μ and standard deviation σ , then

$$\begin{aligned} M_{\bar{X}} &= \text{mean of the sampling distribution of } \bar{X} \\ &= \mu \end{aligned}$$

$$\begin{aligned} \sigma_{\bar{X}} &= \text{standard deviation of the sampling distribution of } \bar{X} \\ &= \frac{\sigma}{\sqrt{n}} \\ &= \text{standard error of } \bar{X}, \text{ where } \bar{X} \text{ is used to estimate } \mu. \end{aligned}$$

Result : If the population from which the random sample is obtained is $N(\mu, \sigma^2)$, then

$$\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right). \quad \begin{array}{l} \text{Notation:} \\ M_{\bar{X}} = \mu \\ \sigma_{\bar{X}}^2 = \frac{\sigma^2}{n} \end{array}$$

$$Z = \frac{\bar{X} - M_{\bar{X}}}{\sigma_{\bar{X}}} = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \quad \text{where } Z \sim N(0, 1).$$

Big Result : Central Limit Theorem (CLT)

Suppose the shape of the population is unknown. When n is "large", then sampling distribution of \bar{X} is approximately normally distributed, i.e.,

$$\bar{X} \stackrel{\text{approx}}{\sim} N\left(\mu, \frac{\sigma^2}{n}\right) \text{ when } n \text{ is "large".}$$

Demonstrate the CLT using java applet.

Rule of Thumb for CLT involving \bar{X}

CLT for \bar{X} holds when $n > 30$, where the less symmetric the population is, the larger n will have to be to ensure normality for the sampling distribution of \bar{X} .

If n is sufficiently large, then

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

and

$$Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$$

where $Z \stackrel{\text{approx}}{\sim} N(0, 1)$.

Ex: X = weight of a newborn manatee.

$$\text{suppose } \mu = 60 \text{ lbs}$$

$$\sigma = 5 \text{ lbs}$$

- i.) Suppose a random sample of 36 newborn manatees is obtained. Find probability the sample mean weight is between 59 lbs and 61 lbs.

$$\begin{aligned}
 n &= 36 \\
 P(59 \leq \bar{x} \leq 61) &= P\left(\frac{59-60}{5/\sqrt{36}} \leq \frac{\bar{x}-\mu}{\sigma/\sqrt{n}} \leq \frac{61-60}{5/\sqrt{36}}\right) \\
 &= P(-1.2 \leq Z \leq 1.2) \\
 &= P(Z \leq 1.2) - P(Z < -1.2) \\
 &= 0.8849 - 0.1151 \\
 &= 0.7698
 \end{aligned}$$

ii) $n=64$

03/01/2023

$$\begin{aligned}
 P(59 \leq \bar{x} \leq 61) &= P\left(\frac{59-60}{5/\sqrt{64}} \leq \frac{\bar{x}-\mu}{\sigma/\sqrt{n}} \leq \frac{61-60}{5/\sqrt{64}}\right) \\
 &= P(-1.6 \leq Z \leq 1.6) \\
 &= 0.9452 - 0.0548 \\
 &= 0.8904
 \end{aligned}$$

Remark : If individual x -values are normally distributed, then above probabilities are exact.

If population of weights is not normally distributed, above probabilities are approximations of the true values.

$$\begin{aligned}
 \text{iii) } n &= 64 \\
 P(\bar{x} \geq 62) &\stackrel{\text{using CLT}}{=} P\left(Z = \frac{62-60}{5/\sqrt{64}}\right) = P(Z \geq 3.2) \\
 &= P(Z \leq -3.2) \\
 &= 0.0007
 \end{aligned}$$

$$\text{Remarks: 1. } Z = \frac{\bar{X} - M}{\sigma/\sqrt{n}}$$

$$2. \bar{X} = M + Z \cdot \frac{\sigma}{\sqrt{n}}$$

The Sampling Distribution of the Sample Proportion (\hat{p})

Ex: Smoking status (Yes, No)

p = population proportion of smokers

\hat{p} = sample proportion of smokers

n = sample size

x_i = response of i th person in sample = $\begin{cases} 1, & \text{if "Yes"} \\ 0, & \text{if "No"} \end{cases}$

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

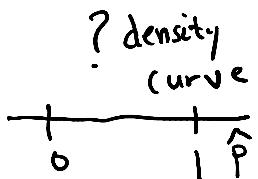
Remark : $0 \leq \hat{p} \leq 1$

$$0 \leq p \leq 1$$

Result : If x_1, x_2, \dots, x_n is a random sample from a population where p = population proportion of "successes", then

$$M_{\hat{p}} = p$$

$$\sigma_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$$



$$D. \quad n \quad \dots \quad n+1 - \hat{p}$$

Big Result : CLT for \hat{p}

If $np \geq 5$ and $n(1-p) \geq 5$, then the CLT for \hat{p} results in

$$\hat{p} \xrightarrow{\text{approx}} N(p, \frac{p(1-p)}{n}).$$

NOTE: $M_{\hat{p}} = p$, $\sigma_{\hat{p}}^2 = \frac{p(1-p)}{n}$

$$Z = \frac{\hat{p} - p}{\sqrt{\frac{p(1-p)}{n}}} = \frac{\hat{p} - M_{\hat{p}}}{\sigma_{\hat{p}}}$$

where $Z \xrightarrow{\text{approx}} N(0,1)$.

