

Read chapter 1.

Sample mean: X_1, X_2, \dots, X_n a random sample from a popn.

$$\bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n} = \frac{\sum_{i=1}^n X_i}{n}$$

Sample variance:

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

\bar{X}, S^2 are both sample statistics,
random variables.

μ, σ^2 — are both popn parameters,
numbers.

Ex: a random sample of size 6,

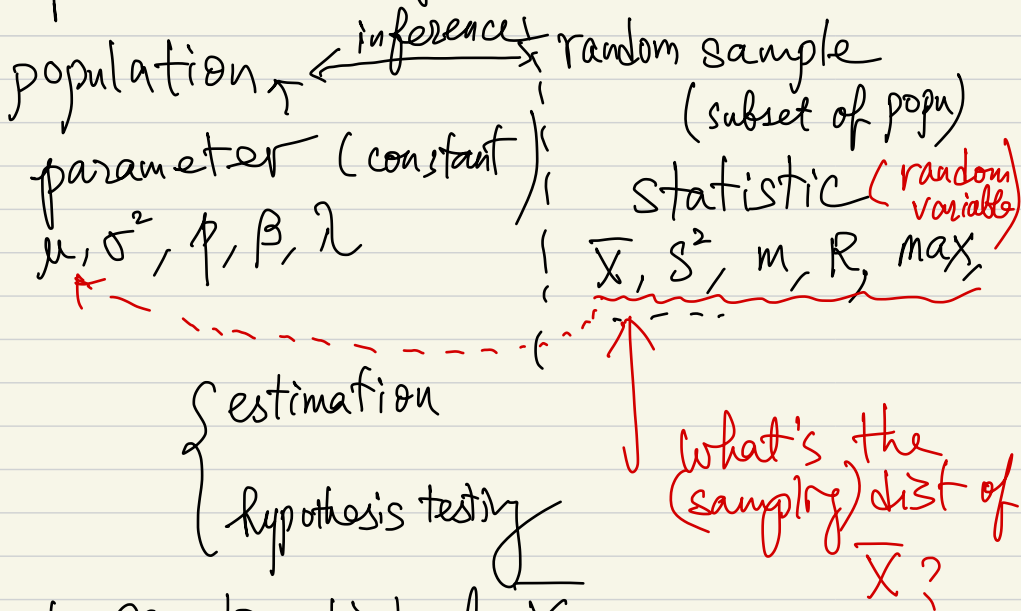
4, 3.2, 5, 3.9, 4.2, 3.7

$$\bar{X} = 4$$

$$S^2 = \frac{0.8^2 + 1^2 + 0.1^2 + 0.2^2 + 0.3^2}{5} = \underline{\hspace{2cm}}$$

S^2 ✓, O^2 ✗

chapter 8 Sampling dist



- The sampling dist of \bar{X} .

$$\text{Ex: } X_1, X_2, X_3 \sim N(70, 10^2)$$

$$\begin{aligned} \textcircled{1} & P(\bar{X} > 66.67) \\ &= P\left(\frac{X_1 + X_2 + X_3}{3} > 66.67\right) \\ &= P(\underbrace{X_1 + X_2 + X_3}_{\substack{\uparrow \\ N(210, 300)}} > 200) \end{aligned}$$

$$= P\left(Z > \frac{200 - 210}{\sqrt{300}}\right) = P\left(Z > -\frac{1}{\sqrt{3}}\right)$$

$$= P(Z > -0.58) = 0.7190$$

Thm: Let X_1, X_2, \dots, X_n be a random sample of size n from $N(\mu, \sigma^2)$

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

$$\sum_{i=1}^n X_i \sim N(n\mu, n\sigma^2)$$

$$\bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n} = \frac{X_1}{n} + \frac{X_2}{n} + \dots + \frac{X_n}{n}$$

$$E(\bar{X}) = \frac{1}{n}\mu + \frac{1}{n}\mu + \dots + \frac{1}{n}\mu = \mu.$$

$$\text{Var}(\bar{X}) = \frac{\sigma^2}{n^2} + \frac{\sigma^2}{n^2} + \dots + \frac{\sigma^2}{n^2} = \frac{\sigma^2}{n}$$

Q: What if we take away the normal assumption in the underlying dist?

We still have $E(\bar{X}) = \mu$, $Var(\bar{X}) = \frac{\sigma^2}{n}$, But not the Normal part.

Ex: $X_1, X_2, \dots, X_n \sim N(80, 8^2)$.

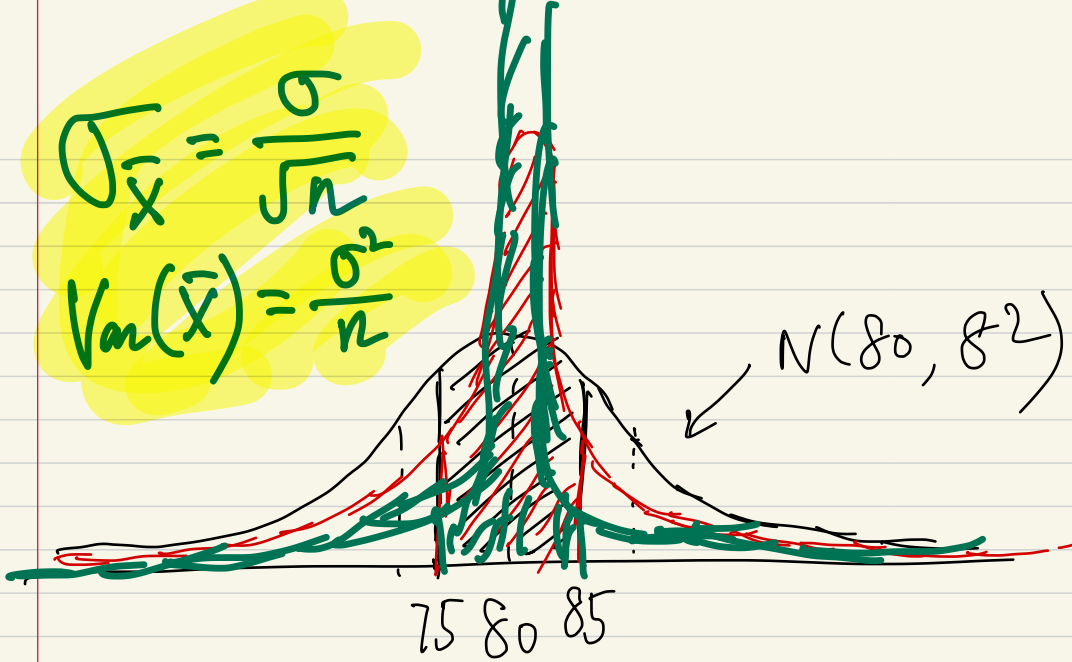
$$\begin{aligned} \textcircled{1} P(|X_1 - 80| < 5) &= P\left(\frac{|X_1 - 80|}{8} < \frac{5}{8}\right) \\ &= P(|Z| < \underline{0.625}) = 1 - 2 * \underset{5286}{0.2643} \\ &= 0.4714 \end{aligned}$$

$\textcircled{2}$ if $n=4$, what is $P(|\bar{X} - 80| < 5) =$

$$\begin{aligned} \bar{X} \sim N\left(80, \frac{8^2}{4}\right) &= N(80, 4^2) \quad \begin{aligned} &= P(|Z| < \frac{5}{4}) \\ &= P(|Z| < 1.25) \\ &= \underline{0.7888} \end{aligned} \end{aligned}$$

$$\begin{aligned} \textcircled{3} \text{ if } n=16, \text{ what is } P(|\bar{X} - 80| < 5) &= P(|Z| < 2.5) \\ \bar{X} \sim N\left(80, \frac{8^2}{16}\right) &= N(80, 2^2) \quad = \underline{\underline{0.987}} \end{aligned}$$

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$
$$\text{Var}(\bar{x}) = \frac{\sigma^2}{n}$$



$$\bar{X}_{n=4} \sim N(80, 4^2)$$

$$\bar{X}_{n=16} \sim N(80, 2^2)$$

The Central Limit Theorem.

Let X_1, X_2, \dots, X_n be a random sample from any dist with mean μ and variance $\sigma^2 < \infty$.

Then the limit dist of $\frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$ is Z as $n \rightarrow \infty$.

In application, when n is large, we say the dist of $\frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$ is approx Z .

$$\Leftrightarrow \bar{X} \overset{\text{approx}}{\sim} N\left(\mu, \frac{\sigma^2}{n}\right)$$

$$\Leftrightarrow \sum_{i=1}^n X_i \overset{\text{approx}}{\sim} N(n\mu, n\sigma^2)$$

$n \geq 30$ is enough.

