

Lecture 20

- Variance of the Sample Mean

- (1) Sample with Replacement: $N \gg n$
- (2) Sample without Replacement: N is small

Case (1): With Replacement

Question: $\text{Var}(\hat{X})$?

where $\hat{X} = \frac{1}{n} \sum_{i=1}^n X_i$
 \uparrow
 also a RV

$$\text{Var}(\hat{X}) = E[(\hat{X})^2] - E^2[\hat{X}]$$

$$E[\hat{X}] = E\left[\frac{1}{n} \sum_{i=1}^n X_i\right] = \frac{1}{n} \sum_{i=1}^n E[X_i] = \frac{1}{n} \cdot \sum_{i=1}^n \bar{X} = \frac{1}{n} \cdot n \cdot \bar{X} = \bar{X}$$

True Mean

How about $E[(\hat{X})^2]$?

$$\hat{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

$$(\hat{X})^2 = \left(\frac{1}{n} \sum_{i=1}^n X_i\right) \cdot \left(\frac{1}{n} \sum_{j=1}^n X_j\right) = \frac{1}{n^2} \sum_{i=1}^n \sum_{j=1}^n X_i \cdot X_j$$

$$E(\hat{X})^2 = \frac{1}{n^2} \sum_{i=1}^n \sum_{j=1}^n E[X_i X_j], \text{ where}$$

$$E[X_i X_j] = \begin{cases} E[X^2], & \text{if } i=j \\ E[X_i] E[X_j], & \text{if } i \neq j \end{cases}$$

$\underbrace{E[X_i] E[X_j]}_{E[X]^2} \rightarrow$ assume that X_i and X_j are uncorrelated

$$E(\hat{X})^2 = \frac{1}{n^2} \sum_{i=1}^n \sum_{j=1}^n E[X_i X_j] = \frac{1}{n^2} \left\{ n E[X^2] + \underbrace{(n^2 - n) E^2[X]}_{n^2 E^2[X] - n E^2[X]} \right\}$$

$$= \frac{1}{n^2} \left\{ n \text{Var}[X] + n^2 E^2[X] \right\} = \frac{\text{Var}[X]}{n} + E^2[X]$$

$$\text{Var}(\hat{X}) = E[(\hat{X})^2] - E^2[\hat{X}]$$

$$= \frac{\text{Var}[X]}{n} + E^2[X] - E^2[X]$$

$$\boxed{\text{Var}(\hat{X}) = \frac{\text{Var}[X]}{n}}$$

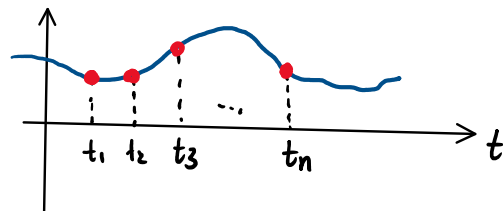
$$n \uparrow \Rightarrow \text{Var}(\hat{X}) \downarrow$$

Example :

Assume that

True Mean = 10

True Variance = 9



Question: How many samples we should take to estimate the true mean value such that $\rightarrow n?$

$$\text{Std}(\hat{X}) = 1\% \text{ of the true mean value ?}$$

$$\downarrow$$

$$\text{Var}(\hat{X}) = (1\% \text{ of } 10)^2 = (0.01 \times 10)^2 = 0.01 = \frac{\text{Var}[X]}{n} = \frac{9}{n}$$

$$\text{Thus } n = \frac{9}{0.01} = 900$$

(2) Sample without Replacement: N is small

$$\text{Var}(\hat{X}) = \frac{\text{Var}[X]}{n} \left(\frac{N-n}{N-1} \right)$$

N: Population Size

n: Sample Size

Example 2:

N = 100 transistors

Estimate the mean value of the current gain β

$$\begin{cases} \text{True population variance} = 25 \\ \text{True population mean} = 120 \end{cases}$$

Question: How many samples we should take to estimate the true mean value such that

Sample mean has a STD that is 1% of the True Mean?

$$\text{Var}(\hat{\beta}) = (1\% \text{ of } 120)^2 = (0.01 \times 120)^2 = 1.44$$

$$1.44 = \frac{\text{Var}(\beta)}{n} \left(\frac{N-n}{N-1} \right) = \frac{25}{n} \cdot \left(\frac{100-n}{100-1} \right) = \frac{25}{n} \cdot \frac{100-n}{99}$$

$$1.44 \times 99n = 25(100-n) = 2500 - 25n$$

$$(1.44 \times 99 + 25)n = 2500$$

$$n = \frac{2500}{1.44 \times 99 + 25} = 15$$

$$\gg 2500 / (1.44 \times 99 + 25)$$

ans =

(14.9200) \rightarrow round it up

Sum of many independent, identical RV's

Example: X_1, X_2, \dots, X_k are all $N(0, \sigma)$

↑ ↑
Mean standard deviation

$$Z = \sum_{i=1}^k X_i$$

Question: PDF of Z ?

$$\begin{aligned} \text{Answer: } \phi_Z(u) &= \phi_{X_1}(u) \cdot \phi_{X_2}(u) \cdot \dots \cdot \phi_{X_k}(u) = e^{-\frac{u^2 \sigma^2}{2}} \cdot e^{-\frac{u^2 \sigma^2}{2}} \dots e^{-\frac{u^2 \sigma^2}{2}} \\ &= e^{-u^2 \frac{k\sigma^2}{2}} \rightarrow N(0, \sqrt{k\sigma^2}) \end{aligned}$$

Central Limit Theorem

X_1, X_2, \dots, X_n are independent RV's, having the same distribution (Mean = m
Variance = σ^2)

$$Y = \frac{1}{\sqrt{n}} \sum_{k=1}^n (X_k - m), \quad \text{where } E[Y] = 0, \text{ Var}[Y] = \sigma^2$$

Then Y has a PDF $f_Y(y)$ that approaches a Gaussian distribution as n increases, regardless of the PDF of X_k 's.

http://www.ece.uah.edu/~dwpan/course/ee385/code/clt_hist.m

```
function [m, v] = clt_hist(n)
x = random('exp', 1, 1, 1000000);
mean(x)
var(x)
hist(x, 100);
for i = 2: n
    y = random('exp', 1, 1, 1000000);
    x = x - mean(x) + (y - mean(y));
end
```

```
S = x/sqrt(n);
m = mean(S);
v = var(S);
figure; hist(S, 100);
```