

• Critical Limit Theorem. (중심극한정리)

↓
 1) 이치: 모집단이 정규분포일지, 그렇지 않을지 모르지만, 모집단 평균이 같을 때.

↓
 ; If the population distribution is $N(\mu, \sigma^2)$

; X_1, X_2, \dots, X_n are iid $N(\mu, \sigma^2)$

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

✓ The critical limit theorem provides an important extension to these results.

✓ Regardless of the actual distribution of the individual random variables X_i , the distribution of their average \bar{X} is closely approximated by a $N(\mu, \sigma^2/n)$ for large sample size.

↓

상호독립인 경우

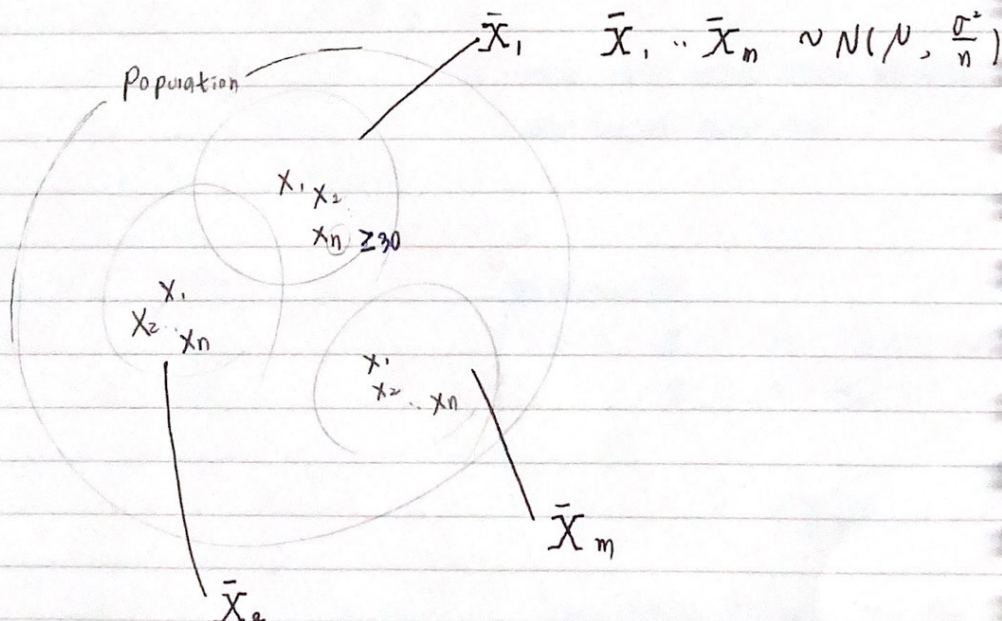
If X_1, X_2, \dots, X_n is a sequence (일련)

of independent (서로소속) identically distributed random variables, with a mean μ
 같은 분포를 가진 \dots μ 의 평균을 가진.

and a variance σ^2 ,

then the distribution of their average \bar{X} can be approximated by $N(\mu, \sigma^2/n)$ for large n (in general $n \geq 30$)

(1) m)



Two Independent populations.

(any distributions) : 모집단이 2개인 경우.

1. $\bar{X}_1 - \bar{X}_2$: $\mu_1 - \mu_2$

- pop. 1 (μ_1, σ_1^2) \rightarrow

; sample 1 with n_1 obs $\rightarrow \bar{X}_1$

- pop 2. (μ_2, σ_2^2) \rightarrow

; sample 2 with n_2 obs $\rightarrow \bar{X}_2$

o Expected Value of $\bar{X}_1 - \bar{X}_2$

$$E[\bar{X}_1 - \bar{X}_2] = E[\bar{X}_1] - E[\bar{X}_2]$$

$$= \mu_1 - \mu_2$$

(독립성이 가능)

o Variance of $\bar{X}_1 - \bar{X}_2$

* Note: \bar{X}_1 and \bar{X}_2 are

"independent random variables."

$$V(\bar{X}_1 - \bar{X}_2) = V(\bar{X}_1) + V(\bar{X}_2)$$

$$= \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$

* IP

pop. 1 is $N(\mu_1, \sigma_1^2)$, and,

pop. 2 is $N(\mu_2, \sigma_2^2)$,

$\bar{X}_1 - \bar{X}_2$ is distributed

$$N(\mu_1 - \mu_2, \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2})$$

예를) $\bar{X}_1 - \bar{X}_2 \sim N(-6, 0.69)$ 라면,

$P[\bar{X}_1 - \bar{X}_2 \leq -6]$?

표준화

$$= P\left[\frac{\bar{X}_1 - \bar{X}_2 - (-6)}{\sqrt{0.69}} \leq \frac{-6 - (-6)}{\sqrt{0.69}}\right]$$

$$= P[Z \leq 2.12]$$

$$= \Phi(2.12) = 0.9830$$

2. S^2 의 분포?

o Sample Variance (표본 분산)

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

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

각각의 제곱 합

* Assume X_1, X_2, \dots, X_n are iid $N(\mu, \sigma^2)$

Then, $(n-1)S^2/\sigma^2 \sim \chi^2(n-1)$

Constant (상수)

σ^2

$S^2 \sim \chi^2(n-1)$,, 분포

chi-square 분포

참 40%