

Estimation

- Estimation

- Estimator

- : the function of X to estimate the parameters

- Estimate

- : the computed value from the estimator

< 추정량
추정치

Estimation

- Estimation

- Point estimation **점추정**

- : Calculate the value of “the best guess” for the parameters

- Interval estimation **구간추정**

- : Calculate the interval which probably contain the parameters

Estimation

● Mean

- Point estimator about μ

X_1, X_2, \dots, X_n from the population whose mean is μ and the variance is σ^2

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

	모수	
평균	오평균 ←	표본 평균 (\bar{x})
분산	오분산 ←	표본 분산 (s^2)
비율	오비율 ←	표본 비율 (\hat{p})

Estimation

● Mean

- $(1-\alpha)$ % confidence Intervals (with large sample)

$$\left(\bar{X} - Z_{\frac{\alpha}{2}} \cdot \frac{s}{\sqrt{n}}, \quad \bar{X} + Z_{\frac{\alpha}{2}} \cdot \frac{s}{\sqrt{n}} \right)$$

- $(1-\alpha)$ % confidence Intervals (with small sample)

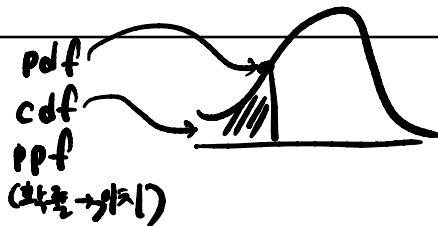
$$\left(\bar{X} - t_{(\frac{\alpha}{2}, n-1)} \cdot \frac{s}{\sqrt{n}}, \quad \bar{X} + t_{(\frac{\alpha}{2}, n-1)} \cdot \frac{s}{\sqrt{n}} \right)$$

Sampling Distributions

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import norm, t
```

- `t` : t-distribution
- `norm`: normal distribution

$$t(\text{len}(x)-1)$$



Sampling Distributions

- `.rvs (size)`
 - Random sample
- `.pdf (x)` or `.pmf (x)`
 - Probability density or probability mass
- `.cdf(x)`
 - Cumulative distribution
- `.ppf(prob)`
 - point percentile

Estimation

● Practice

정답.

- Compute the point estimator of the mean and 95% confidence Intervals of the mean.

```
In [2]: can = np.array([101.8, 101.5, 102.6, 101, 101.8, 96.8, 102.4, 100,
,98.8, 98.1,98.8, 98, 99.4, 95.5, 100.1, 100.5, 97.4
,100.2, 101.4, 98.7,101.4, 99.4, 101.7, 99, 99.7, 98.9
,99.5, 100, 99.7, 100.9,99.7, 99, 98.8, 99.7, 100.9, 99.9
,97.5, 101.5, 98.2, 99.2,98.6, 101.4, 102.1, 102.9, 100.8
,99.4, 103.7, 100.3, 100.2, 101.1,101.8, 100, 101.2, 100.5
,101.2, 101.6, 99.9, 100.5, 100.4, 98.1,100.1, 101.6, 99.3
,96.1, 100, 99.7, 99.7, 99.4, 101.5, 100.9,101.2, 99.9, 99.1
,100.7, 100.8, 100.8, 101.4, 100.3, 98.4,97.2])
```

표본 ← 표본평균 : mean.

신뢰구간 찾기. $[L, U]$

→ 조건 → normal. (가설: 30)
작으면 t

Estimation

● Practice

- Solution:

```
In [7]: n = len(can)
mean_can = np.mean(can)
std_can = np.std(can, ddof=1)
can_norm = norm(loc=0, scale=1)

ll = mean_can - can_norm.ppf(0.975)*std_can/np.sqrt(n)
ul = mean_can + can_norm.ppf(0.975)*std_can/np.sqrt(n)

print("point estimator of mean is %.2f" % mean_can)
print("95% confidence intervals is {} and {}".format(np.round(ll,4), np.round(ul,4)))

point estimator of mean is 100.04
95% confidence intervals is 99.7057 and 100.3733
```

Estimation

- Proportion

- Point estimator about p

X is the number of success from binomial distribution (n, p)

$$\hat{p} = \frac{X}{n}$$

Binomial \rightarrow normal

Estimation

- Proportion

- $(1-\alpha)$ % confidence Intervals (with large sample)

$$\left(\hat{p} - Z_{\frac{\alpha}{2}} \cdot \sqrt{\frac{\hat{p} \cdot (1 - \hat{p})}{n}}, \quad \hat{p} + Z_{\frac{\alpha}{2}} \cdot \sqrt{\frac{\hat{p} \cdot (1 - \hat{p})}{n}} \right)$$

Estimation

● Practice

- A company randomly select 250 customers, and send the promotion code for a new product. Among these, 70 people responded to buy the product using the promotion code. Then what would be the proportion of the customers to buy the product, and what is 90% confidence interval of that.

Estimation

● Practice

- Solution:

```
In [10]: n = 250
x = 70
p_hat = x/n

z_norm = norm(loc=0, scale=1)
ll = p_hat - z_norm.ppf(0.95)*np.sqrt(p_hat*(1-p_hat)/n)
ul = p_hat + z_norm.ppf(0.95)*np.sqrt(p_hat*(1-p_hat)/n)

print("the proportion is : ", p_hat)
print("90% confidence intervals of it is {} and {}".format(np.round(ll,4), np.round(ul,4)))

the proportion is : 0.28
90% confidence intervals of it is 0.2333 and 0.3267
```

Estimation

- Variance

- Point estimator about σ^2

X_1, X_2, \dots, X_n from the population whose mean is μ and the variance is σ^2

$$\widehat{\sigma^2} = s^2 = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n - 1}$$

Estimation

- Variance

- $(1-\alpha)$ % confidence Intervals

$$\left(\frac{(n-1) \cdot s^2}{x_{\left(\frac{\alpha}{2}, n-1\right)}^2}, \quad \frac{(n-1) \cdot s^2}{x_{\left(1-\frac{\alpha}{2}, n-1\right)}^2} \right)$$

Estimation

● Practice

- Find 95% confidence intervals of the variance and variance estimator with 'can' data.

Estimation

● Practice

- Solution:

```
In [11]: from scipy.stats import chi2

n = len(can)
s2 = np.var(can, ddof=1)
chi_dist = chi2(n-1)

ll = (n-1)*s2/chi_dist.ppf(0.975)
ul = (n-1)*s2/chi_dist.ppf(0.025)

print("the variance estimator is : ", s2)
print("95% confidence intervals of it is {} and {}".format(np.round(ll,4), np.round(ul,4)))

the variance estimator is : 2.3130632911392417
95% confidence intervals of it is 1.7325 and 3.2452
```


Estimation

- The difference of two means

- Point estimator of $\mu_1 - \mu_2$

$$\hat{\mu}_1 - \hat{\mu}_2 = \bar{X}_1 - \bar{X}_2$$

Estimation

- The difference of two means

- (1- α)% confidence Intervals (with large sample)

$$\left(\bar{X}_1 - \bar{X}_2 - Z_{\frac{\alpha}{2}} \cdot \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}, \bar{X}_1 - \bar{X}_2 + Z_{\frac{\alpha}{2}} \cdot \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}} \right)$$

Estimation

- The difference of two means
 - (1-α)% confidence Intervals (with small sample and equal variance)

$$s_p^2 = \frac{(n_1 - 1) \cdot s_1^2 + (n_2 - 1) \cdot s_2^2}{(n_1 + n_2 - 2)}$$

$$\left(\bar{X}_1 - \bar{X}_2 - t_{(\frac{\alpha}{2}, n_1+n_2-2)} \cdot s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}, \bar{X}_1 - \bar{X}_2 + t_{(\frac{\alpha}{2}, n_1+n_2-2)} \cdot s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}} \right)$$

Estimation

- Practice
 - Calculate the pooled estimator of variance with A, B groups.
A group is the first 10 observations of 'can' data.
B group is the last 10 observations of 'can' data.

```
In [13]: A_group = can[0:10]
         B_group = can[-10:]

         print("A group:", A_group)
         print("B group:", B_group)
```

```
A group: [101.8 101.5 102.6 101.  101.8  96.8 102.4 100.  98.8  98.1]
B group: [101.2  99.9  99.1 100.7 100.8 100.8 101.4 100.3  98.4  97.2]
```

Estimation

- Practice

- Solution:

```
In [15]: n1 = len(A_group)
          n2 = len(B_group)

          s2_pool = ((n1-1)*np.var(A_group, ddof=1) + (n2-1)*np.var(B_group, ddof=1))/(n1+n2-2)
          print("pooled estimator of the variance is ", np.round(s2_pool,4))

          pooled estimator of the variance is  2.8729
```

Estimation

- Practice

- Calculate 95% of the confidence intervals of the difference of the means.

Estimation

- Practice

- Solution:

```
In [18]: a_mean = np.mean(A_group)
         b_mean = np.mean(B_group)

         t_can = t(n1+n2-2)
         ll = (a_mean-b_mean) - t_can.ppf(0.975)*np.sqrt(s2_pool)*np.sqrt(1/n1 + 1/n2)
         ul = (a_mean-b_mean) + t_can.ppf(0.975)*np.sqrt(s2_pool)*np.sqrt(1/n1 + 1/n2)

         print("The difference of two means :", np.round(a_mean-b_mean, 4))
         print("95% of CI of the difference of two means is {} and {}".format(np.round(ll,4), np.round(ul,4)))

The difference of two means : 0.5
95% of CI of the difference of two means is -1.0925 and 2.0925
```

Estimation

- The difference of two means

- Point estimator of $p_1 - p_2$

$$\hat{p}_1 - \hat{p}_2 = \frac{X_1}{n_1} - \frac{X_2}{n_2}$$

Estimation

- The difference of two proportions
 - $(1-\alpha)\%$ confidence Intervals (with large sample)

$$\left(\hat{p}_1 - \hat{p}_2 - Z_{\frac{\alpha}{2}} \cdot \sqrt{\frac{\hat{p}_1 \cdot (1 - \hat{p}_1)}{n_1} + \frac{\hat{p}_2 \cdot (1 - \hat{p}_2)}{n_2}}, \hat{p}_1 - \hat{p}_2 + Z_{\frac{\alpha}{2}} \cdot \sqrt{\frac{\hat{p}_1 \cdot (1 - \hat{p}_1)}{n_1} + \frac{\hat{p}_2 \cdot (1 - \hat{p}_2)}{n_2}} \right)$$

Estimation

- Practice
 - A survey company asked 100 male and 100 female if they married or not. 62 male and 29 female responded that they are married.

Compute the 90% confidence Intervals of the difference of two proportions.

Estimation

● Practice

▪ Solution:

```
In [19]: n1 = n2 = 100
p1_hat = 62/100
p2_hat = 29/100
z_norm = norm(loc=0, scale=1)

ll = (p1_hat - p2_hat) - z_norm.ppf(0.975)*np.sqrt(p1_hat*(1-p1_hat)/n1 + p2_hat*(1-p2_hat)/n2)
ul = (p1_hat - p2_hat) + z_norm.ppf(0.975)*np.sqrt(p1_hat*(1-p1_hat)/n1 + p2_hat*(1-p2_hat)/n2)

print("The difference of two proportions :", np.round(p1_hat-p2_hat, 4))
print("95% of CI of the difference of two proportions is {} and {}".format(np.round(ll,4), np.round(ul,4)))

The difference of two proportions : 0.33
95% of CI of the difference of two proportions is 0.1998 and 0.4602
```

Estimation

● Practice

▪ Solution:

```
In [19]: n1 = n2 = 100
p1_hat = 62/100
p2_hat = 29/100
z_norm = norm(loc=0, scale=1)

ll = (p1_hat - p2_hat) - z_norm.ppf(0.975)*np.sqrt(p1_hat*(1-p1_hat)/n1 + p2_hat*(1-p2_hat)/n2)
ul = (p1_hat - p2_hat) + z_norm.ppf(0.975)*np.sqrt(p1_hat*(1-p1_hat)/n1 + p2_hat*(1-p2_hat)/n2)

print("The difference of two proportions :", np.round(p1_hat-p2_hat, 4))
print("95% of CI of the difference of two proportions is {} and {}".format(np.round(ll,4), np.round(ul,4)))

The difference of two proportions : 0.33
95% of CI of the difference of two proportions is 0.1998 and 0.4602
```