

Test for σ^2

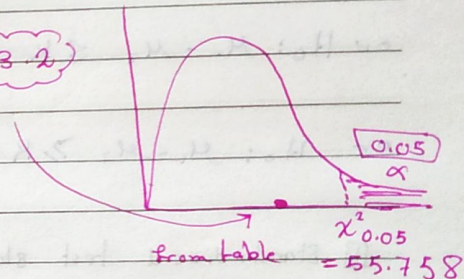
eg) A dairy processing company claims that the variance of the amount of fat in the whole milk processed by the company is no more than 0.25. You suspect this is wrong and find that a random sample of 41 milk containers has a variance of 0.27. At $\alpha = 0.05$, is there enough evidence to reject the company's claim? Assume the population is normally distributed.

$$H_0: \sigma^2 \leq 0.25 \text{ (claim)} \quad \text{and} \quad H_a: \sigma^2 > 0.25 \quad n = 41$$

$$d.f. = 41 - 1 = 40 \quad \text{(right-tailed test)} \quad s^2 = 0.27$$

$$\chi^2 = \frac{(n-1) s^2}{\sigma^2} = \frac{40 (0.27)}{(0.25)} = 43.2$$

Accept $H_0 \rightarrow$ Accept the claim.
Don't reject the claim.



Page 420 Chapter 7. Example (5)

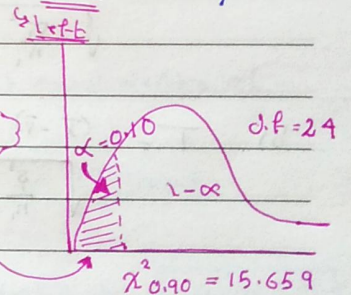
the claim \rightarrow "the standard deviation is less than 1.4 minutes."

$$H_0: \sigma \geq 1.4 \quad \& \quad H_a: \sigma < 1.4 \text{ (claim)}$$

$$[n = 25], \quad [s = 1.1], \quad [\alpha = 0.10]$$

$$\chi^2 = \frac{(n-1) s^2}{\sigma^2} = \frac{(25-1) (1.1)^2}{(1.4)^2} = 14.816$$

[\rightarrow reject $H_0 \rightarrow$ Don't reject the claim] in rejection region



If $X_1, X_2, \dots, X_{n_1} \overset{i.i.d.}{\sim} n(\mu_1, \sigma_1^2)$
 $\& Y_1, Y_2, \dots, Y_{n_2} \overset{i.i.d.}{\sim} n(\mu_2, \sigma_2^2)$ \swarrow Independent sample

mean estimate for μ_1 : \bar{X}_1

" " " " μ_2 : \bar{X}_2

13 independent Samples

* Test for $\mu_1 - \mu_2$ لقد تم تعديلها من قبلنا

$H_0: \mu_1 - \mu_2 = K$ vs $H_a: \mu_1 - \mu_2 \neq K \rightarrow$ two tailed test

or $H_0: \mu_1 - \mu_2 \leq K$ vs $H_a: \mu_1 - \mu_2 > K \rightarrow$ right tailed test

or $H_0: \mu_1 - \mu_2 \geq K$ vs $H_a: \mu_1 - \mu_2 < K \rightarrow$ left tailed test

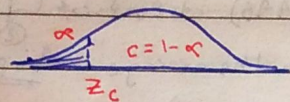
Standardized test statistic is :

a) $Z = \frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)_0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$, if $\sigma_1^2 \& \sigma_2^2$ are known p-value لا يسأل عن الـ μ الا بعد ذلك

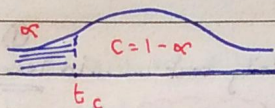
b) $T = \frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)_0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$, if $\sigma_1^2 \neq \sigma_2^2$ (are unknown)
 [d.f = $\min\{n_1 - 1, n_2 - 1\}$] نذكر

c) $T = \frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)_0}{\hat{\sigma} \cdot \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$, if $\sigma_1^2 = \sigma_2^2 = \sigma$ (unknown)
 [d.f = $n_1 + n_2 - 2$]

$\hat{\sigma} \leftarrow$
 $\hat{\sigma} P = \hat{\sigma} = \sqrt{\frac{(n_1 - 1) s_1^2 + (n_2 - 1) s_2^2}{n_1 + n_2 - 2}}$



Left tailed test



Left tailed test

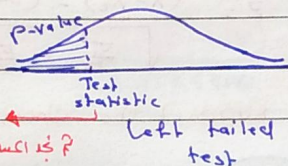
الطرف الأيسر

(reject H_0) إذا كانت القيمة الإحصائية أكبر من القيمة الحرجة

Don't reject H_0 ← (Accept H_0) إذا كانت القيمة الإحصائية أصغر من القيمة الحرجة

↳ failed to reject H_0

P-value الطريقة الأخرى



(table) ما هو الـ P-value
P-value ما هو الـ P-value

if P-value $\leq \alpha$, reject H_0

∴ الـ P-value

and if P-value $> \alpha$, Accept H_0

⇒ σ_1 & σ_2 are known ⇒ Z-test

Example 2 ← Page (444) يجب أن يكون الـ σ_1 و σ_2 معروفين

Claim: "there is a difference in the mean credit card debts of people in California and Florida."

$H_0: \mu_1 = \mu_2$ and $H_1: \mu_1 \neq \mu_2$ [Claim]

250 People from each state $\Rightarrow \mu_1 - \mu_2 = 0$ two tailed test

$n_1 = n_2 = 250$, the two samples are independent $\alpha = 0.05$

Assume that $\sigma_1 = 960$, $\sigma_2 = 845$

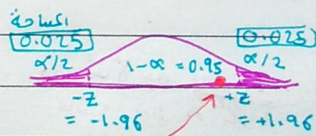
California: $\bar{x}_1 = 3060$, $n_1 = 250$, $\sigma_1 = 960$

Florida: $\bar{x}_2 = 2910$, $n_2 = 250$, $\sigma_2 = 845$

are known

sol)

(Test statistic)
$$Z = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} = \frac{(3060 - 2910) - 0}{\sqrt{\frac{(960)^2}{250} + \frac{(845)^2}{250}}} = 1.85$$



(You fail to reject the H_0) → there isn't enough evidence to support the claim. Z is not in the rejection region.

Claim: "the mean driving cost per mile of a manufacturer's sedans is less than that of its leading competitor."

$H_0: \mu_1 \geq \mu_2 \rightarrow \mu_1 - \mu_2 \geq 0$ $H_a: \mu_1 < \mu_2$ [claim]
 → left tailed test

Manufacturer: $\bar{x}_1 = 0.48$, $s_1 = 0.05$, $n_1 = 30$ $\sigma_1^2 \neq \sigma_2^2$ are unknown

Competitor: $\bar{x}_2 = 0.51$, $s_2 = 0.07$, $n_2 = 32$

Assume the population variances are equal. ($\sigma_1^2 = \sigma_2^2$)

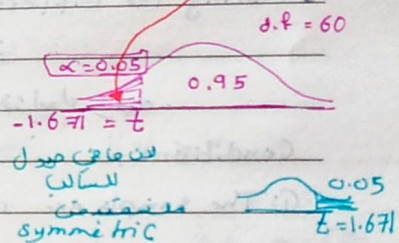
$\alpha = 0.05$

d.f = $n_1 + n_2 - 2$
 = $30 + 32 - 2 = 60$

$t = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)_0}{SP \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$ Standard error
 = $\frac{(0.48 - 0.51) - 0}{0.0155416} \approx -1.930$

standard error $S_{\bar{x}_1 - \bar{x}_2} = \sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2} \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}$
 → = $\sqrt{\frac{(30-1)(0.05)^2 + (32-1)(0.07)^2}{60} \left(\frac{1}{30} + \frac{1}{32} \right)} \approx 0.0155416$

reject $H_0 \rightarrow$ there is enough evidence to support the claim.



8.4 Test for $P_1 - P_2$:

- $H_0: P_1 - P_2 = K$ vs $H_a: P_1 - P_2 \neq K$
- $H_0: P_1 - P_2 \leq K$ vs $H_a: P_1 - P_2 > K$
- $H_0: P_1 - P_2 \geq K$ vs $H_a: P_1 - P_2 < K$

- ① the samples are
 - a) randomly selected
 - b) independent
 - c) large ...

standardized test statistic is :

$Z = \frac{(\hat{P}_1 - \hat{P}_2) - (P_1 - P_2)_0}{\sqrt{\bar{p}\bar{q} \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}}$

use this use ← P-value

large samples: $n_1 p_1 \geq 5, n_2 p_2 \geq 5$
 $n_1 q_1 \geq 5, n_2 q_2 \geq 5$
 $\hat{P}_1 = \frac{x}{n_1}, \hat{P}_2 = \frac{y}{n_2}$
 $\bar{p} = \frac{x+y}{n_1+n_2}, \bar{q} = 1 - \bar{p}$

A Two-Sample z-test For the Difference Between Proportions

Example 1

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$H_0: P_1 = P_2 \rightarrow$ claim and $H_a: P_1 \neq P_2 \rightarrow$ 2-tailed test

Passenger Cars $n_1 = 200$, $\hat{P}_1 = 0.910$, $X_1 = 182 \rightarrow \hat{P} = \frac{x_1}{n_1} \rightarrow x_1 = \hat{P} n_1$

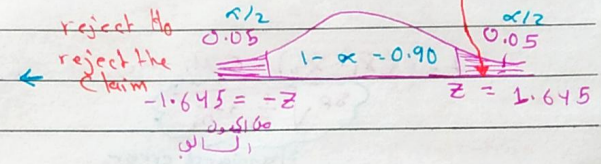
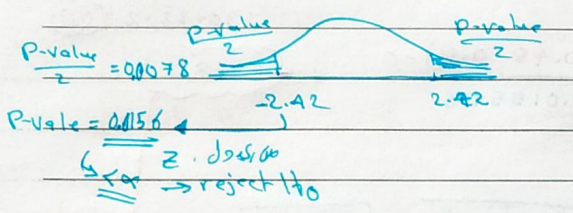
Pickup Trucks $n_2 = 250$, $\hat{P}_2 = 0.832$, $X_2 = 208 \rightarrow x_2 = \hat{P} n_2$

$\alpha = 0.10$

$$z = \frac{(\hat{P}_1 - \hat{P}_2) - (P_1 - P_2)}{\sqrt{\bar{p}\bar{q}\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}} = \frac{(0.910 - 0.832) - (0)}{\sqrt{(0.8667)(0.1333)\left(\frac{1}{200} + \frac{1}{250}\right)}} = 2.42$$

$$\bar{p} = \frac{x_1 + x_2}{n_1 + n_2} = \frac{182 + 208}{200 + 250} = 0.8667, \quad \bar{q} = 1 - 0.8667 = 0.1333$$

in rejection region



8.3 Dependent Samples

• Testing the difference between means [Dependent samples]

إذا كان الاختبار على نفس الأشخاص قبل وبعد إجراء سنجين ...
مثلاً يتم تجريبه سواء قبل أو بعد العلاج على نفس الأشخاص من قبل إجراء سنجين.

Conditions:

- ① The sample are randomly selected.
- ② the samples are dependent (paired).
- ③ The populations are normally distributed or the number n of pairs of data is at least 30.

x_i	y_i	$d_i = x_i - y_i$	d_i^2
$\sum d_i$			$\sum d_i^2$

$$\bar{d} = \frac{\sum d_i}{n}$$

$$sd = \sqrt{\frac{\sum d_i^2}{n-1} - \frac{(\sum d_i)^2}{n(n-1)}}$$

C.I for μ_d : $\bar{d} \pm t_c \cdot \frac{sd}{\sqrt{n}}$

$\bar{d} - t_c \frac{sd}{\sqrt{n}} < \mu_d < \bar{d} + t_c \frac{sd}{\sqrt{n}}$

standardized test statistic is : $t = \frac{\bar{x} - \mu_d}{sd/\sqrt{n}}$ d.f = n-1

Example 1

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Athlete	1	2	3	4	5	6	7	8		
vertical jump..(x _i) before..	24	22	25	28	35	32	30	27	$\alpha = 0.10$ claim :- (After) > (Before) y_i x_i	
vertical jump..(y _i) After using..	26	25	25	29	33	34	35	30		
Difference (d _i)	-2	-3	0	-1	+2	-3	-5	-3		$\sum d_i = -14$
	(d _i) ²	4	9	0	1	4	9	25	9	$\sum d_i^2 = 56$

$sd = \sqrt{\frac{\sum d_i^2}{n-1} - \frac{(\sum d_i)^2}{n(n-1)}} = \sqrt{\frac{56}{8} - \frac{(-14)^2}{8(7)}} = 2.1213$

claim:- $(x_i - y_i) < 0 \rightarrow H_a: \mu_d < 0$ [claim] and $H_0: \mu_d \geq 0$
 ↳ Lefty tailed test

$\bar{d} = \frac{\sum d}{n} = \frac{-14}{8} = -1.75$

$\alpha = 0.10$ $d.f = 8-1 = 7$

مقطع الوجب من الجدود $t = -1.415$

The standardized test statistic is

$t = \frac{\bar{d} - \mu_d}{sd/\sqrt{n}} = \frac{-1.75 - 0}{2.1213/\sqrt{8}} = -2.333$

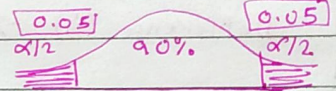
in rejection region

Patient	1	2	3	4	5	6	7	8	$(\sum d)$
without drug x_i	1.8	2.0	3.4	3.5	3.7	3.8	3.9	3.9	}
using the drug y_i	3.0	3.6	4.0	4.4	4.5	5.2	5.5	5.7	
$d_i (x_i - y_i)$	-1.2	-1.6	-0.6	-0.9	-0.8	-1.4	-1.6	-1.8	$\sum d_i = -9.9$
$(d_i)^2$	1.44	2.56	0.36	0.81	0.64	1.96	2.56	3.24	$\sum d_i^2 = 13.57$

$$sd = \sqrt{\frac{\sum d_i^2}{n-1} - \frac{(\sum d_i)^2}{n(n-1)}} = \sqrt{\frac{13.57}{7} - \frac{(-9.9)^2}{8(7)}} = \sqrt{0.18839}$$

$$sd = 0.434, \quad \bar{d} = \frac{\sum d_i}{n} = \frac{-9.9}{8} = -1.238$$

$$\bar{d} - t_c \frac{sd}{\sqrt{n}} < \mu_d < \bar{d} + t_c \frac{sd}{\sqrt{n}}$$



$$\left(-1.238 - (-1.895 \cdot \frac{0.434}{\sqrt{8}}) < \mu_d < -1.238 + (-1.895 \cdot \frac{0.434}{\sqrt{8}}) \right) \quad -t_c = -1.895$$

$$-1.5288 < \mu_d < -0.947$$

For $\mu \rightarrow$ point estimate (\bar{x})

$$C.I. \Rightarrow \bar{x} \pm z_c \cdot \frac{\sigma}{\sqrt{n}} \quad (\sigma \text{ known})$$

$$\bar{x} \pm t_c \cdot \frac{s}{\sqrt{n}} \quad (\sigma \text{ unknown})$$

$$\bar{x} \pm z_c \cdot \frac{s}{\sqrt{n}} \quad (n \geq 30)$$

test statistic \Rightarrow

$$Z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} \quad (\sigma \text{ known})$$

$$T = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} \quad (\sigma \text{ unknown})$$

For $p \rightarrow$ Point estimate $\hat{p} = \frac{x}{n}$

$$C.I \rightarrow \hat{p} \pm z_c \sqrt{\frac{\hat{p}\hat{q}}{n}}$$

$$\text{best statistic } t \Rightarrow \frac{\hat{p} - p_0}{\sqrt{\frac{p_0 q_0}{n}}}$$

For $\sigma^2 \rightarrow$ Point estimate is S^2

$$C.I \Rightarrow \left(\frac{(n-1)S^2}{\chi^2_R}, \frac{(n-1)S^2}{\chi^2_L} \right)$$

$$\text{best statistic is: } \frac{(n-1)S^2}{\sigma^2} \rightarrow d.f = n-1$$

For $\mu_1 - \mu_2 \rightarrow$ Point estimate is $\bar{X} - \bar{Y}$

$$C.I \rightarrow \bar{X} - \bar{Y} \pm z_c \cdot \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}} \quad (\sigma_1, \sigma_2 \text{ known})$$

$$\bullet (\bar{X} - \bar{Y}) \pm t_c \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}} \quad (\sigma_1 \neq \sigma_2 \text{ unknown})$$

$$\bullet (\bar{X} - \bar{Y}) \pm t_c \hat{\sigma} \cdot \sqrt{\frac{1}{n_1} + \frac{1}{n_2}} \quad (\sigma_1 = \sigma_2 \text{ unknown})$$

$$\bullet * \hat{\sigma} = \sqrt{\frac{(n_1-1)s_1^2 + (n_2-1)s_2^2}{n_1 + n_2 - 2}}$$

$$\text{Test statistic is :- } Z = \frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)_0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \quad (\sigma_1, \sigma_2 \text{ known})$$

$$T = \frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)_0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}} \quad (\sigma_1 \neq \sigma_2 \text{ unknown}) \quad \left. \begin{array}{l} d.f = \min\{n_1-1, n_2-1\} \end{array} \right\}$$

$$T = \frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)_0}{\hat{\sigma} \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \quad \left. \begin{array}{l} \sigma_1 = \sigma_2 = \sigma \text{ unknown} \\ d.f = n_1 + n_2 - 2 \end{array} \right\}$$

For $P_1 - P_2 \rightarrow$ Point estimate is $\hat{P}_1 - \hat{P}_2$

$$C.I \rightarrow (\hat{P}_1 - \hat{P}_2) \pm z_c \cdot \sqrt{\frac{P_1 \hat{q}_1}{n_1} + \frac{P_2 \hat{q}_2}{n_2}}$$

test statistic is :-

$$Z = \frac{(\hat{P}_1 - \hat{P}_2) - (P_1 - P_2)_0}{\sqrt{\bar{P} \bar{q} \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

$$\hat{q} \rightarrow \bar{P} = \frac{X_1 + X_2}{n_1 + n_2}, \bar{q} = 1 - \bar{P}$$

* Paired data:-

$$C.I \quad \bar{d} \pm t_c \frac{S_d}{\sqrt{n}}, \quad d.f = n - 1$$

$$\text{Test statistic :- } \frac{\bar{d} - (\mu_d)_0}{S_d / \sqrt{n}}$$

• $\mu_d < 0 \rightarrow$ Before $<$ After

$\mu_d > 0 \rightarrow$ Before $>$ After

$\mu_d = 0 \rightarrow$ Before = After.