

①

$$a.) \quad Y_i = \beta_1 + \beta_2 X_i + u_i$$

$$u_i = Y_i - \beta_1 - \beta_2 X_i$$

$$u_i^2 = (Y_i - \beta_1 - \beta_2 X_i)^2$$

$$\min_{\beta_1, \beta_2} \sum \hat{u}_i^2 = \sum (Y_i - \beta_1 - \beta_2 X_i)^2$$

$$\beta_1 : \frac{\partial \sum \hat{u}_i^2}{\partial \hat{\beta}_1} = 0$$

$$\frac{\partial \sum (Y_i - \hat{\beta}_1 - \hat{\beta}_2 X_i)^2}{\partial \hat{\beta}_1} = 0$$

$$2 \sum (Y_i - \hat{\beta}_1 - \hat{\beta}_2 X_i) (-1) = 0$$

$$\sum (Y_i - \hat{\beta}_1 - \hat{\beta}_2 X_i) = 0$$

$$\sum Y_i - \sum \hat{\beta}_1 - \sum \hat{\beta}_2 X_i = 0$$

$$\frac{\sum Y_i}{n} - \frac{\sum \hat{\beta}_1}{n} - \frac{\sum \hat{\beta}_2 X_i}{n} = 0$$

$$\bar{Y} - \hat{\beta}_1 - \hat{\beta}_2 \bar{X} = 0$$

$$\hat{\beta}_1 = \bar{Y} - \hat{\beta}_2 \bar{X} \quad (1)$$

$$\beta_2 : \frac{\partial \sum \hat{u}_i^2}{\partial \hat{\beta}_2} = 0$$

$$\frac{\partial \sum (Y_i - \hat{\beta}_1 - \hat{\beta}_2 X_i)^2}{\partial \hat{\beta}_2} = 0$$

$$2 \sum (Y_i - \hat{\beta}_1 - \hat{\beta}_2 X_i) (-X_i) = 0$$

$$\sum (Y_i - \hat{\beta}_1 - \hat{\beta}_2 X_i) (X_i) = 0 \quad (2)$$

$$\hat{\beta}_1 = \bar{y} - \hat{\beta}_2 \bar{x} \quad (1)$$

$$\sum (y_i - \hat{\beta}_1 - \hat{\beta}_2 x_i)(x_i) = 0 \quad (2)$$

plug (1) into (2), $\sum (y_i - (\bar{y} - \hat{\beta}_2 \bar{x}) - \hat{\beta}_2 x_i)(x_i) = 0$

$$\sum [(y - \bar{y})x_i - \hat{\beta}_2 (\bar{x} - x_i)x_i] = 0$$

$$\sum (y - \bar{y})x_i = \sum \hat{\beta}_2 (\bar{x} - x_i)x_i$$

$$\frac{\sum (y - \bar{y})x_i}{\sum (\bar{x} - x_i)x_i} = \hat{\beta}_2$$

$$\frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (\bar{x} - x_i)^2} = \hat{\beta}_2$$

$$\therefore \hat{\beta}_2 = \frac{20.58}{211} = \frac{1029}{10550} = 0.0975$$

sub $\hat{\beta}_2 = \frac{1029}{10550}$ into $\sum (y_i - \hat{\beta}_1 - \hat{\beta}_2 x_i) = 0$

$$\sum (y_i - \hat{\beta}_1 - \hat{\beta}_2 x_i) = 0$$

$$\frac{\sum (y_i - \hat{\beta}_1 - \hat{\beta}_2 x_i)}{n} = \hat{\beta}_1$$

$$\frac{50.9 - \left(\frac{1029}{10550} \right) (388)}{18} = \hat{\beta}_1$$

$$\hat{\beta}_1 = 0.7253$$

- ① a. The estimator of $\beta_1 = \hat{\beta}_1 = 0.7253$
The estimator of $\beta_2 = \hat{\beta}_2 = 0.0975$

$$Y_i = \beta_1 + \beta_2 X_i + u_i$$

$$Y_i = 0.7253 + 0.0975 X_i + \hat{u}_i$$

The interception is 0.7253 and slope is 0.0975 ~~#~~

①

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$$b.) R^2 = 1 - \frac{\sum \hat{u}_i^2}{\sum (y_i - \bar{y})^2}$$

$$R^2 = 1 - \frac{0.5781}{2.5844}$$

$$R^2 = 0.7763$$

The r^2 is determined by how much the r^2 is described by the SRF, or the measurement of goodness of fit of the fitted regression line comparing to an estimator $\bar{y}$.

This mean $r^2=0.7763$ has high confidential because it has small range and the data closes to SRF line.

$$c.) x_i = 30 \quad \hat{y}_i = ?$$

$$\hat{y}_i = \hat{\beta}_1 + \hat{\beta}_2 x_i$$

$$\hat{y}_i = 0.7253 + (0.0975) 30 = 3.6503$$

When $x_i=30$ the expect value is 3.6503

$$d.) \text{Variance } u_i : \hat{\sigma}_{u_i}^2 = \frac{\sum \hat{u}_i^2}{n-k} = \frac{0.5781}{(18-2)} = 0.0361$$

$$\text{Variance } \hat{\beta}_1 : \hat{\sigma}_{\hat{\beta}_1}^2 = \frac{\sum x_i^2}{n \sum (x_i - \bar{x})^2} \hat{\sigma}^2 = \frac{9,620}{18(211)} \times 0.0361 = 0.0914$$

$$\text{Variance } \hat{\beta}_2 : \hat{\sigma}_{\hat{\beta}_2}^2 = \frac{\hat{\sigma}^2}{\sum (x_i - \bar{x})^2} = \frac{0.0361}{211} = 1.7109 \times 10^{-4}$$

① e.) $\alpha = 0.1$

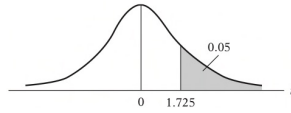
$$P \left[\hat{\beta}_2 - \left(t_{\frac{\alpha}{2}} \cdot se_{\hat{\beta}_2} \right) \leq \beta_2 \leq \hat{\beta}_2 + \left(t_{\frac{\alpha}{2}} \cdot se_{\hat{\beta}_2} \right) \right] = 0.9$$

Example

$$Pr(t > 2.086) = 0.025$$

$$Pr(t > 1.725) = 0.05 \quad \text{for } df = 20$$

$$Pr(t > 1.725) = 0.10$$



Pr df	0.25 0.50	0.10 0.20	0.05 0.10	0.025 0.05	0.01 0.02	0.005 0.010	0.001 0.002
1	1.000	3.078	6.314	12.706	31.821	63.657	318.31
2	0.816	1.886	2.920	4.303	6.965	9.925	22.327
3	0.765	1.638	2.353	3.182	4.541	5.841	10.214
4	0.741	1.533	2.132	2.776	3.747	4.604	7.173
5	0.727	1.476	2.015	2.571	3.365	4.032	5.893
6	0.718	1.440	1.943	2.447	3.143	3.707	5.208
7	0.711	1.415	1.895	2.365	2.998	3.499	4.785
8	0.706	1.397	1.860	2.306	2.896	3.355	4.501
9	0.703	1.383	1.833	2.262	2.821	3.250	4.297
10	0.700	1.372	1.812	2.228	2.764	3.169	4.144
11	0.697	1.363	1.796	2.201	2.718	3.106	4.025
12	0.695	1.356	1.782	2.179	2.681	3.055	3.930
13	0.694	1.350	1.771	2.160	2.650	3.012	3.852
14	0.692	1.345	1.761	2.145	2.624	2.977	3.787
15	0.691	1.341	1.753	2.131	2.602	2.947	3.733
16	0.690	1.337	1.746	2.120	2.583	2.921	3.686
17	0.689	1.333	1.740	2.110	2.567	2.898	3.646
18	0.688	1.330	1.734	2.101	2.552	2.878	3.610
19	0.688	1.328	1.729	2.093	2.539	2.861	3.579
20	0.687	1.325	1.725	2.086	2.528	2.845	3.552
21	0.686	1.323	1.721	2.080	2.518	2.831	3.527
22	0.686	1.321	1.717	2.074	2.508	2.819	3.505
23	0.685	1.319	1.714	2.069	2.500	2.807	3.485
24	0.685	1.318	1.711	2.064	2.492	2.797	3.467
25	0.684	1.316	1.708	2.060	2.485	2.787	3.450
26	0.684	1.315	1.706	2.056	2.479	2.779	3.435
27	0.684	1.314	1.703	2.052	2.473	2.771	3.421
28	0.683	1.313	1.701	2.048	2.467	2.763	3.408
29	0.683	1.311	1.699	2.045	2.462	2.756	3.396
30	0.683	1.310	1.697	2.042	2.457	2.750	3.385
40	0.681	1.303	1.684	2.021	2.423	2.704	3.307
60	0.679	1.296	1.671	2.000	2.390	2.660	3.232
120	0.677	1.289	1.658	1.980	2.358	2.617	3.160
∞	0.674	1.282	1.645	1.960	2.326	2.576	3.090

Note: The smaller probability shown at the head of each column is the area in one tail; the larger probability is the area in both tails.

$$t_{\frac{\alpha}{2}} = t_{0.05} = 1.746 \quad \text{when } df = 16$$

$$\text{degree of freedom} = n - k = 18 - 2 = 16$$

$$se^2 = \text{variance}$$

$$se = \sqrt{\text{variance}}$$

$$se_{\hat{\beta}_2} = \sqrt{1.7109 \times 10^{-4}} = 0.0131$$

$$P[0.0975 - (1.746 \times 0.0131) \leq \beta_2 \leq 0.0975 + (1.746 \times 0.0131)] = 0.9$$

$$P[0.0975 - 0.0228726 \leq \beta_2 \leq 0.0975 + 0.0228726]$$

$$P[0.0746 \leq \beta_2 \leq 0.1204] = 0.9$$

From 90 out of 100 times β_2 is between 0.0746 and 0.1204. ✗

① f.) step 1: $H_0: \beta_2 = 0$ - Null hypothesis

$H_a: \beta_2 \neq 0$ - Alternative hypothesis

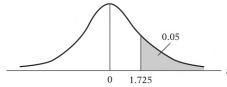
step 2: $\alpha = 0.05$

$$\text{step 3: } t_{\text{cal}} = \frac{\hat{\beta}_2 - \beta_2}{\text{se } \hat{\beta}_2} = \frac{0.0995 - 0}{0.0131} = 7.4427$$

step 4: The upper boundary = $t_{\frac{\alpha}{2}} = t_{\frac{0.05}{2}} = t_{0.0025}$ df = 16 = 2.120

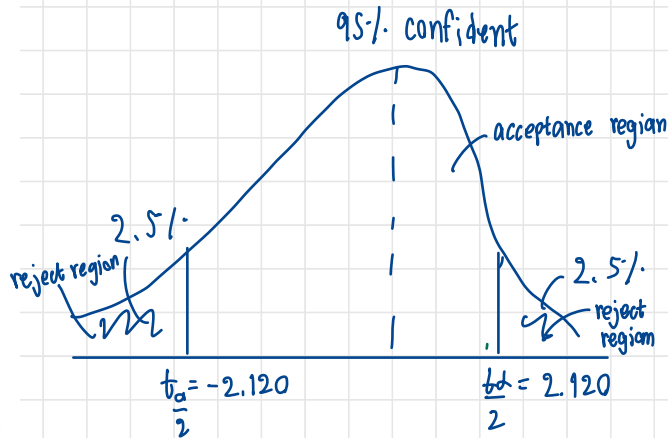
The lower boundary = $t_{\frac{\alpha}{2}} = -2.120$

Example
Pr(t > 2.086) = 0.025
Pr(t > 1.725) = 0.05
Pr(|t| > 1.725) = 0.10



Pr	0.25	0.10	0.05	0.025	0.01	0.005	0.001
df	0.20	0.10	0.05	0.025	0.01	0.005	0.001
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step 5: t_{cal} lies beyond any boundary of test value critical region, we can reject null hypothesis at the significance level of 95%.
We are sure that β_2 is not zero from sample 95 out of 100 times.

② a.) step 1: $H_0: \beta_1 = 0$ - null hypothesis

$H_a: \beta_1 \neq 0$ - alternative hypothesis

step 2: $\alpha = 0.05$

$$\text{step 3: } t_{\text{cal } \beta_1} = \frac{\hat{\beta}_1 - \beta_1}{\text{se}_{\hat{\beta}_1}} = \frac{0.9299898 - 0}{0.0140339} = 30.4969$$

$$\text{step 4: upper boundary } t_{\frac{\alpha}{2}} = t_{0.025} \text{ df } 29,885 = 1.960$$

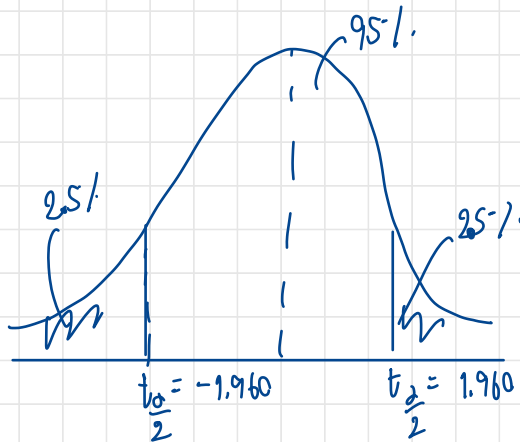
$$\text{lower boundary} = -1.960$$

Example
Pr($t > 2.086$) = 0.025
Pr($t > 1.725$) = 0.05 for df = 20
Pr($t > 1.725$) = 0.10



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∞	0.674	1.282	1.645	1.960	2.326	2.576	3.090

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steps: t_{cal} lies beyond any boundary of test value critical region, we can reject null hypothesis at the significance level of 95%.
we are sure that β_1 is not zero from sampling 95 out of 100 times,

2.) a.) step 1. $H_0: \beta_2 = 0$ - null hypothesis
 $H_a: \beta_2 \neq 0$ - alternative hypothesis

step 2 $\alpha = 0.05$

step 3 $t_{cal} = \frac{\hat{\beta}_2 - \beta_2}{se_{\hat{\beta}_2}} = \frac{0.0031338 - 0}{0.0002292} = 13.6728$

step 4 upper boundary $t_{\alpha/2} = t_{0.025}$ $df = 27,885 = 1.960$

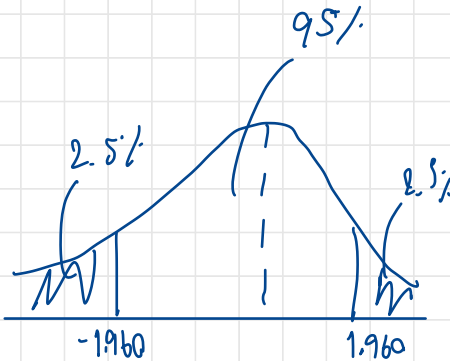
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lower boundary = -1.960

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40	0.681	1.303	1.684	2.021	2.423	2.704	3.307	
60	0.679	1.296	1.671	2.000	2.390	2.660	3.232	
120	0.677	1.289	1.658	1.980	2.358	2.617	3.166	
∞	0.674	1.282	1.645	1.960	2.326	2.576	3.090	

Note: The smaller probability shown at the head of each column is the area in one tail; the larger probability is the area in both tails.



step 5: t_{cal} lies beyond any boundary of test value critical region, we can reject null hypothesis at the significance level of 95%.
 We are sure that β_2 is not zero from sampling 95 out of 100 times.

2. b) $\hat{\beta}_2$ mean when age increase 1-year-old the average to visit to hospital will increase by 0.0031338 time. $\hat{\beta}_2$ make economic sense because when age increase the opportunity to visit hospital also higher.

$$c.) \hat{out}_i = \hat{\beta}_1 + \hat{\beta}_2 \text{ age}_i$$

$$\hat{out}_i = 0.4279898 + 0.031338 \text{ age}_i$$

$$\ln \hat{out}_i = 0.4279898 + 0.031338 \text{ age}_i$$

$$\frac{\ln \hat{out}_i}{\partial \text{age}_i} = 0.031338$$

An increase of age by 1 year is associated with opportunity to visit hospital in 2015 increase of $100 \times 0.031338 = 3.1338$ percent

- d.) If age_i variable is divided by 10, the coefficients, standard errors, and confidence intervals of age are change. the constant is 0.4279898 time and slope (coefficient of age) is 0.00031338 year.

2. c) $\hat{y}_i = \beta_1 + \beta_2 x_i$ $x_0 = 50$ $\alpha = 0.01$ $\text{Var}(\hat{y}_0) = 0.00002$

$\hat{y}_i = 0.4279898 + 0.031338 x_i$

$\hat{y}_0 = 0.4279898 + 0.0031338 \times 50$

$\hat{y}_0 = 0.5847$

$\hat{se}_{\hat{y}_0} = \sqrt{\text{Var}(\hat{y}_0)} = \sqrt{0.00002} = 4.4721 \times 10^{-3}$

$t_{\frac{\alpha}{2}} = t_{0.005}$ $df = 27,885 = 2.576$

Example
 $\Pr(z > 2.086) = 0.025$
 $\Pr(z > 1.725) = 0.05$
 $\Pr(t > 1.725) = 0.10$

Appendix D: Statistical Tables 879



Pr	0.25	0.10	0.05	0.025	0.01	0.005	0.001
1	1.000	0.978	0.954	0.930	0.905	0.881	0.854
2	0.816	0.886	0.920	0.930	0.965	0.975	0.983
3	0.755	0.838	0.894	0.920	0.954	0.969	0.977
4	0.741	0.833	0.894	0.920	0.954	0.969	0.977
5	0.727	0.826	0.894	0.920	0.954	0.969	0.977
6	0.718	0.820	0.894	0.920	0.954	0.969	0.977
7	0.711	0.815	0.894	0.920	0.954	0.969	0.977
8	0.706	0.811	0.894	0.920	0.954	0.969	0.977
9	0.703	0.808	0.894	0.920	0.954	0.969	0.977
10	0.700	0.805	0.894	0.920	0.954	0.969	0.977
11	0.697	0.803	0.894	0.920	0.954	0.969	0.977
12	0.695	0.801	0.894	0.920	0.954	0.969	0.977
13	0.694	0.799	0.894	0.920	0.954	0.969	0.977
14	0.692	0.797	0.894	0.920	0.954	0.969	0.977
15	0.691	0.795	0.894	0.920	0.954	0.969	0.977
16	0.690	0.793	0.894	0.920	0.954	0.969	0.977
17	0.689	0.791	0.894	0.920	0.954	0.969	0.977
18	0.688	0.789	0.894	0.920	0.954	0.969	0.977
19	0.688	0.787	0.894	0.920	0.954	0.969	0.977
20	0.687	0.785	0.894	0.920	0.954	0.969	0.977
21	0.686	0.783	0.894	0.920	0.954	0.969	0.977
22	0.686	0.781	0.894	0.920	0.954	0.969	0.977
23	0.685	0.779	0.894	0.920	0.954	0.969	0.977
24	0.685	0.777	0.894	0.920	0.954	0.969	0.977
25	0.684	0.775	0.894	0.920	0.954	0.969	0.977
26	0.684	0.773	0.894	0.920	0.954	0.969	0.977
27	0.684	0.771	0.894	0.920	0.954	0.969	0.977
28	0.683	0.769	0.894	0.920	0.954	0.969	0.977
29	0.683	0.767	0.894	0.920	0.954	0.969	0.977
30	0.683	0.765	0.894	0.920	0.954	0.969	0.977
40	0.681	0.763	0.894	0.920	0.954	0.969	0.977
60	0.679	0.761	0.894	0.920	0.954	0.969	0.977
100	0.677	0.759	0.894	0.920	0.954	0.969	0.977
z	0.674	0.752	0.894	0.920	0.954	0.969	0.977

Note: The smaller probability shown at the head of each column is the area in the tail; the larger probability in the body is the area to the left.

$\Pr(\hat{y}_0 - (t_{\frac{\alpha}{2}} \cdot se_{\hat{y}_0}) \leq y_0 \leq \hat{y}_0 + (t_{\frac{\alpha}{2}} \cdot se_{\hat{y}_0})) = 1 - \alpha$

$\Pr(0.5847 - (2.576 \times 4.4721 \times 10^{-3}) \leq y_0 \leq 0.5847 + (2.576 \times 4.4721 \times 10^{-3})) = 0.99$

$\Pr(0.5732 \leq y_0 \leq 0.5962) = 0.99$

at 99% confidence y_0 is between 0.5732 and 0.5962

3. Mean prediction is providing that mean estimation follows this equation $\hat{Y}_0 = \hat{\beta}_1 + \hat{\beta}_2 x_0$ when $\hat{Y}_0$ is an estimator of $E(Y|x_0)$ while x_0 represents a value of interest.

Individual prediction is contrast to the mean prediction, which estimates the variance around Y_0 , individual prediction focuses on forecasting error (f_e). $f_e = \hat{Y}_0 - Y_0$. Therefore, we define the variance of this f_e as $\text{var}(f_e) = \text{var}(\hat{Y}_0 - Y) = \sigma^2 \left[1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{\sum (x_i - \bar{x})^2} \right]$

Similarly, replacing the unknown σ^2 with the unbiased estimator $\hat{\sigma}^2$, we can derive the CI for Y_0 corresponding to x_0 . $\Pr \left[\hat{Y}_0 - \left(t_{\frac{\alpha}{2}} \cdot \text{se}_{f_e} \right) \leq Y_0 \leq \hat{Y}_0 + \left(t_{\frac{\alpha}{2}} \cdot \text{se}_{f_e} \right) \right] = 1 - \alpha$

if $x_0 - \bar{x}$ is larger so the variance is larger and $\text{se}(\hat{Y}_0)$ also increase ($\text{se}(\hat{Y}_0) = \sqrt{\text{var}(\hat{Y}_0)}$).

The big variance make the data lower confidence.