

Lecture Notes 4

1. Confidence Intervals

Point estimate is a single number, like $\hat{\beta}$

To calculate a confidence interval, we start with the following equation

:

$$\Pr\left(-t_c \leq \frac{\hat{\beta}_j - \beta_j}{std.error(\hat{\beta}_j)} \leq t_c\right) = 1 - \alpha$$

where Pr denotes probability

t_c the critical value from the t-distribution

associated with a

α is the level of significance, usually $\alpha = 5\%$

$1 - \alpha$ is the chance that the true parameter lies within that interval

$\hat{\beta}$ and $std.error(\hat{\beta})$ are estimated values from your OLS regressions.

β is the true value.

$$\Pr(-t_c \cdot std.error(\hat{\beta}) \leq \hat{\beta}_j - \beta \leq t_c \cdot std.error(\hat{\beta})) = 1 - \alpha.$$

Subtracting $\hat{\beta}_j$ from both sides gives:

$$\Pr(-\hat{\beta}_j - t_c \cdot std.error(\hat{\beta}) \leq -\beta \leq -\hat{\beta}_j + t_c \cdot std.error(\hat{\beta})) = 1 - \alpha.$$

Multiplying by -1 gives:

$$\Pr(\hat{\beta}_j + t_c \cdot std.error(\hat{\beta}) \geq \beta \geq \hat{\beta}_j - t_c \cdot std.error(\hat{\beta})) = 1 - \alpha.$$

Rearranging terms gives:

$$(4) \quad \Pr(\hat{\beta}_j - t_c \cdot std.error(\hat{\beta}) \leq \beta \leq \hat{\beta}_j + t_c \cdot std.error(\hat{\beta})) = 1 - \alpha.$$

This gives the $1 - \alpha$ confidence interval. The confidence interval is known as an interval estimate in contrast to the point estimates.

Example – Homework #4

ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	93.05434	93.05434	1032.473	3.27436E-38
Residual	57	5.137275	0.090128		
Total	58	98.19162			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>	<i>Lower 95%</i>	<i>Upper 95%</i>	<i>Lower 95.0%</i>	<i>Upper 95.0%</i>
Intercept	4.820175	0.079173	60.8815	1.46E-53	4.661633978	4.978717	4.661634	4.978717
X Variable 1	0.073747	0.002295	32.13212	3.27E-38	0.06915094	0.078343	0.069151	0.078343

$df = 57$, using $\alpha = 0.05$, then $t_c = 2.00$

Confidence interval for intercept

$$\Pr(4.820 - 2.00 \cdot 0.079 \leq \beta_1 \leq 4.820 + 2.00 \cdot 0.079) = 1 - .05$$

$$\Pr(4.662 \leq \beta_1 \leq 4.978) = 95\%$$

There is a 95% chance that the true parameter, β_1 , lies between 4.662 and 4.978

$$\Pr(0.0737 - 2.00 \cdot 0.0023 \leq \beta_2 \leq 0.0737 + 2.00 \cdot 0.0023) = 1 - .05$$

$$\Pr(0.0691 \leq \beta_2 \leq 0.0783) = 95\%$$

There is a 95% chance that the true parameter, β_2 , lies between 0.0691 and 0.0783.

If a confidence interval includes zero, then you cannot rule out that the parameter equals zero. The estimate for the parameter would have an insignificant t-statistic.

Prediction Confidence Intervals

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