

STAT 151A Lecture 21

Henry Liev

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Remark 0.1

For (j in 1:B)

Resample (z_1, \dots, z_m)

(z_1^*, \dots, z_m^*)

$\hat{\theta}_j^*$

Organize $\hat{\theta}_j^*$ smallest to biggest, then form confidence interval

Method 0.2 (Studentized Interval)

For (j in 1:B)

Resample (z_1, \dots, z_m)

(z_1^*, \dots, z_m^*)

$\hat{\theta}_j^*$

$\frac{\hat{\theta}_j^* - \hat{\theta}}{\widehat{SE}(\hat{\theta}_j^*)}$

Use a formula to estimate $\widehat{SE}(\hat{\theta}_j^*)$ or

Use a second bootstrap

For (k in 1:B₂)

Resample (z_1^*, \dots, z_n^*)

$(z_1^{**}, \dots, z_n^{**})$

$\hat{\theta}_k^{**}$

$\widehat{SE}(\hat{\theta}_j^*) = \sqrt{\frac{1}{B_2-1} \sum (\hat{\theta}_k^{**} - \bar{\hat{\theta}}^{**})^2}$

Plug this into formula

$q_{(1)}, \dots, q_{(B)}$, $q_{(i)}$ a quantile from smallest to largest

Extract percentiles

Reverse engineer

$$\mathbb{P}(q_{(\alpha/2)B} \leq \frac{\hat{\theta}_j^* - \hat{\theta}}{\widehat{SE}(\hat{\theta}_j^*)} \leq q_{(1-\alpha/2)B}) = 1 - \alpha \approx \mathbb{P}(q_{(\alpha/2)B} \leq \frac{\hat{\theta} - \theta}{\widehat{SE}(\hat{\theta})} \leq q_{(1-\alpha/2)B})$$

$$= \mathbb{P}(-q_{(1-\alpha/2)B} \leq \frac{\theta - \hat{\theta}}{\widehat{SE}(\hat{\theta})} \leq -q_{(\alpha/2)B})$$

$$\mathbb{P}(\hat{\theta} - q_{(1-\alpha/2)B} \widehat{SE}(\hat{\theta}) \leq \theta \leq \hat{\theta} + q_{(\alpha/2)B} \widehat{SE}(\hat{\theta}))$$

Remark 0.3 (Bootstrap for Regression Models)

$(y_1, \vec{x}_1), \dots, (y_n, \vec{x}_n)$

How does the bootstrap help us with inference tasks here? $\hat{y} = \mathbf{X}\hat{\beta}$

Confidence intervals for $\hat{\beta}_j$

$H_0 : \beta_j = 0$ (t-stat)

$H_0 : \beta_1 = \dots = \beta_k = 0, L\beta = c$ (F-test)

2 general bootstrapping approaches:

Bootstrapping cases: $(\vec{x}_1, y_1), \dots, (\vec{x}_n, y_n) \rightarrow (\vec{x}_1^*, y_1), \dots, (\vec{x}_n^*, y_n) \rightarrow \hat{y}^* = \mathbf{X}\hat{\beta}^*$

Repeat B times to get $(\hat{y}_{(k)}^*, \mathbf{X}_{(k)}^*, \hat{\beta}_{(j)}^*)$

Focus on CI's for now:

Extract $(\hat{y}_{(k)}^*, \mathbf{X}_{(k)}^*, \hat{\beta}_{(k)}^*)$