

Regression (x_i, y_i) , $y_i = x_i^T \beta + e_i$

$$\mathbb{E}[e_i | x_i] = 0$$

$$\text{Var}(e_i | x_i) = \sigma^2$$

$$\text{Cov}(e_i, e_j | x_i, x_j) = 0$$

(i) Non parametric bootstrap

$$\text{Bootstrap} \{(x_i, y_i)\} \longrightarrow \{(x_i^*, y_i^*)\}$$

Assume that $(x_i, y_i) \stackrel{\text{iid}}{\sim} F$

(ii) Parametric Bootstrap:

Take estimates $\hat{\beta}, \hat{\sigma}^2$, $x_i^* = x_i$,

$$y_i^* = x_i^T \hat{\beta} + \hat{e}_i^*$$

$$\hat{e}_i^* \stackrel{\text{iid}}{\sim} N(0, \hat{\sigma}^2)$$

Make inferences on (X_i^*, Y_i^*)

Assumptions: $Y_i = x_i^T \beta + e_i$, $e_i | x_i \stackrel{iid}{\sim} N(0, \sigma^2)$

(iii) "Residual" Bootstrap

$$\{\hat{e}_i = Y_i - \hat{Y}_i\} \xrightarrow{\text{Boot}} \{\hat{e}_i^*\}$$

$$\text{then } X_i^* = X_i, Y_i^* = \hat{Y}_i + \hat{e}_i^*$$

assumption: $e_i | x_i \stackrel{iid}{\sim} F_e$

In all cases, obtain bootstrap

Samples $\hat{\beta}^*$ by regressing

$$Y^* \sim X^*$$