

$$2. \quad \overline{err} = \frac{1}{N} \sum_{i=1}^N (y_i - x_i^T \hat{\beta})^2 \quad \overline{Err} = \frac{1}{M} \sum_{i=1}^M (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2$$

$$\mathbb{E}[\overline{Err}] = \mathbb{E}\left[\frac{1}{M} \sum_{i=1}^M (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2\right]$$

$$= \frac{1}{M} \sum_{i=1}^M \mathbb{E}[(\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2]$$

$$= \mathbb{E}[(\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2] \text{ since each } \tilde{y}_i - \tilde{x}_i^T \hat{\beta} \text{ is i.i.d}$$

This means that $\mathbb{E}[\overline{Err}]$ is independent of M so we can set $M=N$, $\mathbb{E}[\overline{Err}] = \mathbb{E}\left[\frac{1}{N} \sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2\right]$

Then given $((\tilde{x}_1, \tilde{y}_1), \dots, (\tilde{x}_N, \tilde{y}_N))$ we will define new OLS estimate for this set $\hat{\beta} = \arg \min_{\beta} \sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \beta)^2$

Thus by definition we know that $\hat{\beta}$ minimizes $\sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \beta)^2$ so

$$\sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2 \leq \sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \beta)^2 \text{ for any } \beta \text{ so we}$$

$$\sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2 \leq \sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2 \text{ choose } \hat{\beta} \text{ for this}$$

$$\frac{1}{N} \sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2 \leq \frac{1}{N} \sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2$$

$$\mathbb{E}\left[\frac{1}{N} \sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2\right] \leq \mathbb{E}\left[\frac{1}{N} \sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2\right]$$

$$\mathbb{E}\left[\frac{1}{N} \sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2\right] \leq \mathbb{E}[\overline{Err}]$$

$\frac{1}{N} \sum_{i=1}^N (\tilde{y}_i - \tilde{x}_i^T \hat{\beta})^2$ and $\frac{1}{N} \sum_{i=1}^N (y_i - x_i^T \hat{\beta})^2$ are identically distributed so their expectations are equal

$$\mathbb{E}[\overline{err}] \leq \mathbb{E}[\overline{Err}]$$