

OLS when dimension is $\leq n$:

$$EY = X \beta^0 + \varepsilon, \quad \dim(X) = n \times p \text{ with } p \leq n$$

$$E[\|X(\hat{\beta}_{OLS} - \beta^0)\|_2^2 / n] = \underbrace{\text{"Bias"}^2}_{=0} + \underbrace{\text{"Variance"}}_{=0}$$

$$= \sigma^2 \frac{r}{n}$$

$$\leadsto \|X(\hat{\beta}_{OLS} - \beta^0)\|_2^2 / n = \sigma^2 \left(\frac{r}{n}\right) \quad (*)$$

if oracle tells me S_0

$$\text{where } Y = X \begin{pmatrix} \beta^0 + \varepsilon \\ \beta^1 \end{pmatrix}_{n \times 1} = X_{S_0} \beta_{S_0}^0 + \varepsilon$$



~~dim of~~

run OLS on this model

since $\dim(S_0) = |I| = s_0$

$$(*) \quad \left\| X \begin{pmatrix} \beta_{OLS}^1 \\ \beta_{Oracle}^1 \end{pmatrix} - \beta^0 \right\|_2^2 / n = O_p \left(\frac{s_0}{n} \right)$$

$$\|\hat{\beta} - \beta^0\|_2 \xrightarrow{p} 0 \quad \text{if} \quad s_0 = o\left(\frac{n}{\log(n)}\right)$$

$$\|\hat{\beta} - \beta^0\|_1 \xrightarrow{p} 0 \quad \text{if} \quad s_0 = o\left(\sqrt{\frac{n}{\log(n)}}\right)$$

Lasso - OLS hybrid :

(1) Run Lasso $\rightarrow \hat{\beta}$; $|\hat{\beta}| \leq \min(k, p)$

(2) Run OLS on $\hat{\beta} \rightarrow \hat{\beta}_{OLS}$

this is good if $\hat{\beta} \neq \beta_0$

Why? Because OLS has no bias

goal: "something" which is computable
for general X and with lower bias than
Lasso

(1) Lasso-OLS hybrid

(2) Adaptive Lasso

→ I just look at (2) it is orthonormal design
to illustrate what's going on

Adaptive Lasso:

$$\text{if } \hat{\beta}_{\text{init},j} = 0$$

$\Rightarrow \hat{\beta}_{\text{adaptive},j}(\lambda) = 0$ because there is infinite penalty on j