

# Lecture November 15

Model  $Y_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \varepsilon_i,$   
 $\varepsilon_i \stackrel{iid}{\sim} N(0, \sigma^2)$

$$\mu(x_{1i}, x_{2i}) = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i}$$

$$\mu(a+1, b) = \beta_0 + \beta_1(a+1) + \beta_2(b)$$

$$\mu(a, b) = \beta_0 + \beta_1 a + \beta_2 b$$

$$\Rightarrow \underbrace{\mu(a+1, b) - \mu(a, b)} = \beta_1 \quad \text{①}$$

Change in the mean function  
for every unit increase in  $x_1$   
but keeping  $x_2$  fixed

$$= \beta_2 \quad \text{②}$$

Set of Potential Predictors / Indep Variables  
 $\{U_1, U_2, \dots, U_M\}$

Response variable  $Y$

(1) Forward Selection

(2) Backward Selection

(3) FB Selection

(4) Regularization

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### FORWARD ALGORITHM

Step 0. Model  $Y_i = \beta_0 + \varepsilon_i$

Step 1. Model  $Y_i = \beta_0 + \beta_1 x_{1i} + \varepsilon_i$

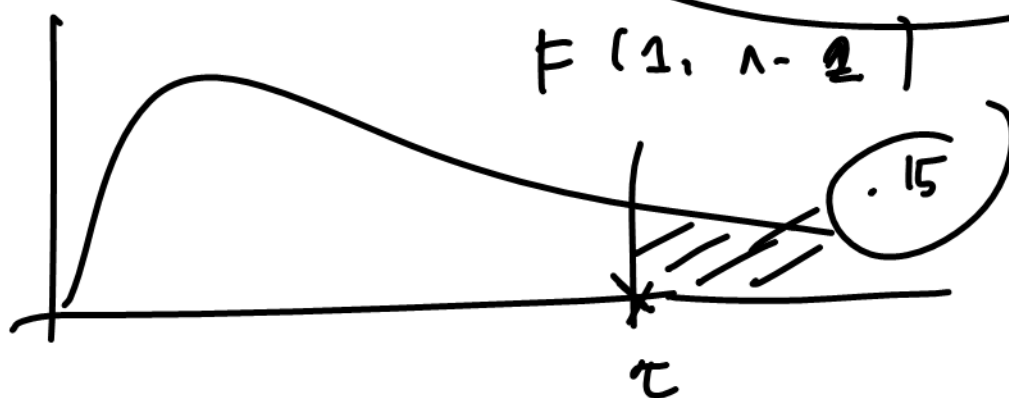
where  $x_{1i}$  is selected from  $\{U_1, \dots, U_M\}$

For each candidate predictor  $U_g, g=1, \dots, M$   
 fit a model  $Y_i = \beta_0 + \beta_1 U_{gi} + \varepsilon_i$

$$SSE(0)$$

$$SSE(U_g)$$

$$F_g = \frac{(SSE(0) - SSE(U_g)) / 1}{SSE(U_g) / (n-2)}$$



Let  $X_1$  be the  $U_g$  s.t.

$$F_g > c \text{ and } F_g = \arg \max_g F_g$$

Step 2. Let  $P_2 = \{U_1, \dots, U_n\} - \{X_1\}$

$$\text{Model: } Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \varepsilon_i$$

For each  $U_f \in \mathcal{P}_2$ , fit the model

$$Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 U_{fi} + \varepsilon_i$$

$$SSE(X_1, U_f)$$

$$F_f = \frac{SSE(X_1) - SSE(X_1, U_f)}{1} \cdot \frac{1}{\frac{SSE(X_1, U_f)}{n-3}}$$

Choose  $U_f$  s.t. that

$$F_f > \tau \equiv F(0.85; 1, n-3) \text{ and}$$

$$f = \arg \max_{U_f \in \mathcal{P}_2} F_f$$

$$X_2 = F_f$$

CONTINUE...

Subsets

$$\{U_1, U_2, U_3\}$$

$$2^3 = 8.$$

$$\bullet Y_i = \beta_0 + \varepsilon_i$$

$$\bullet Y_i = \beta_0 + \beta_1 U_{1i} + \beta_2 U_{2i} + \beta_3 U_{3i} + \varepsilon_i$$

$U_1$

$U_2$

$U_3$

$$\bullet Y_i = \beta_0 + \beta_1 U_{1i} + \beta_2 U_{2i} + \beta_3 U_{3i} + \varepsilon_i$$

$U_1, U_2$

$U_1, U_3$

$U_2, U_3$

$$\bullet Y_i = \beta_0 + \beta_1 U_{1i} + \dots$$

$\begin{matrix} U_1 \\ U_2 \\ \vdots \\ U_{10} \end{matrix}$

Summarize  $\rightarrow$

• Avg  $\bar{U}_{(1..10)}$  WARNING

• PCA

$$S = U_1 - U_2 + \frac{2}{3} U_3 - \dots$$

$U_i$

Summarize  $\rightarrow$

$U_i$

$\rightarrow$   $\Sigma$

## BACKWARD SELECTION

"Complete" model:  $\{U_1, \dots, U_m\}$

$$Y_i = \beta_0 + \beta_1 U_{1i} + \dots + \beta_m U_{mi} + \epsilon_i$$

$$SSE(U_1, \dots, U_m)$$

For each  $g$ :  $U_g$  :  $Y_i = \underline{w/g} U_g + \epsilon_i$