

## 4.2 F test.

### 4.2.1 All coefficients = 0.

#### ① Partitioning total variability

$$TSS = RegSS + ResSS$$

$$\sum (y_i - \bar{y})^2 = \sum (\hat{y}_i - \bar{y})^2 + \sum (y_i - \hat{y}_i)^2$$

#### ② Distributions.

$$\frac{ResSS}{\sigma^2} \sim \chi^2_{n-p'} \perp \frac{RegSS}{\sigma^2} \sim \chi^2_{p, \lambda} \Rightarrow \frac{TSS}{\sigma^2} \sim \chi^2_{n-1, \lambda}, \quad \lambda = \frac{1}{\sigma^2} \sum_{i,j=1}^p \beta_i \beta_j S_{x_i x_j}$$

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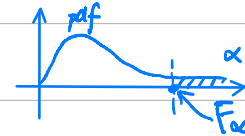
#### ③ Test.

$$H_0: \beta_1 = \dots = \beta_p = 0.$$

$$\text{Test stat: } F = \frac{RegSS/p}{ResSS/(n-p-1)} \stackrel{H_0}{\sim} F(p, n-p-1)$$

Reject  $H_0$  if  $F > F_{\alpha}(p, n-p-1)$ .

$$P(F > F_{\alpha}) = \alpha$$



ANOVA table:

Source	Sum of squares (S.S.)	d.f.	Mean Squares (M.S.)	F
Regression	$Reg.S.S. = \sum_{i=1}^n (\hat{y}_i - \bar{y})^2$	$p$	$SS_{reg}/p$	$F = \frac{SS_{reg}/p}{SS_{res}/(n-p')} = \frac{MS_{reg}}{\hat{\sigma}^2}$
Residual	$Res. S.S. = \sum_{i=1}^n (y_i - \hat{y}_i)^2$	$n - p'$	$SS_{res}/(n - p')$	
Total	$Total S.S. = \sum_{i=1}^n (y_i - \bar{y})^2$	$n - 1$		

### 4.2.2 Subset of regression coefficients

- Whether a reduced model (under  $H_0$ ) is good enough.

- Partitioning reg. SS

$$RegSS|_F = RegSS|_R + \text{Increase in Reg. SS}$$

$$\sim \sigma^2 \chi^2(p, \lambda) \quad \sim \sigma^2 \chi^2(p-r, \lambda) \quad \sim \sigma^2 \chi^2(r, \lambda_2)$$

# free parameters reduced by  $H_0$ .

$$H_0: \beta_1 = \beta_2 = 0$$

# known params specified by  $H_0$ .

$$H_0: \beta_1 = \beta_2$$

- Test

$H_0$ : a reduced model.

Test stat:  $F = \frac{\text{Increase in Reg. SS}}{\hat{\sigma}^2} \stackrel{H_0}{\sim} F(r, n-p-1).$

Reject  $H_0$  if  $F > F_{\alpha}(r, n-p-1).$

#### 4.2.3 Generalized linear hypothesis (including above as special cases)

$H_0: C\beta = d$

Test stat:  $F = \frac{(C\hat{\beta} - d)^T [C(X^T X)^{-1} C^T]^{-1} (C\hat{\beta} - d)}{r \hat{\sigma}^2} \stackrel{H_0}{\sim} F(r, n-p-1), \quad r = \text{rank}(C).$

Reject  $H_0$  if  $F > F_{\alpha}(r, n-p-1)$