

## F - test motivation

⇒ we want to test the significance of a group of hypotheses (multiple hypotheses)

$$\text{Grade}_{325} = \beta_0 + \beta_1 \# \text{times\_front} + \beta_2 \# \text{times\_back} + \beta_3 \text{hr\_study} + \beta_4 \text{past\_GPA} + \beta_5 \text{gender} + u$$

$H_0$ : seat position doesn't have impact on GPA

$$\beta_1 = 0 \text{ and } \beta_2 = 0 \Rightarrow \beta_1 = \beta_2 = 0$$

$H_a$ : seat position matters

$$\left. \begin{array}{l} \beta_1 \neq 0 \text{ and } \beta_2 \neq 0 \\ \text{or } \beta_1 \neq 0 \text{ and } \beta_2 = 0 \\ \text{or } \beta_1 = 0 \text{ and } \beta_2 \neq 0 \end{array} \right\} \text{at least one of the } \beta_1, \beta_2 \neq 0$$

7 Testing Multiple Linear Restrictions: The F-test

Suppose the model is specified by

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + u$$

$H_0 :$   $\beta_1 = 0$  and  $\beta_2 = 0$   $\rightarrow$  want to test if  $x_1$  &  $x_2$  both have no impact on  $y$ .  
 $H_1 :$   $H_0$  is not true

We can use the F-test to test this type of "multiple hypotheses".

1. Our full model is called the "unrestricted" model (ur). Suppose it can be expressed as:

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 + u$$

is true  
 $\Rightarrow$  Reject  $H_0$

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \dots + \beta_k x_k + u$$

2. The model which takes out  $x$  (which we think its associated  $\beta = 0$ ) is called the restricted model (r).  $\leftarrow$  small model

$\rightarrow Y = \beta_0 + \beta_1 X_1 + u$  is true  $\Rightarrow$  do not reject  $H_0$

o suppose there are "q" number of  $\beta$  that we would like to perform a joint-test of  $= 0$

e.g. in this model  $q=2$

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_{k-q} X_{k-q} + u$$

$$H_0: \beta_{k-q+1} = \beta_{k-q+2} = \dots = \beta_k = 0$$

(the last  $q$   $\beta_s = 0$ )

$H_0$  is not true.

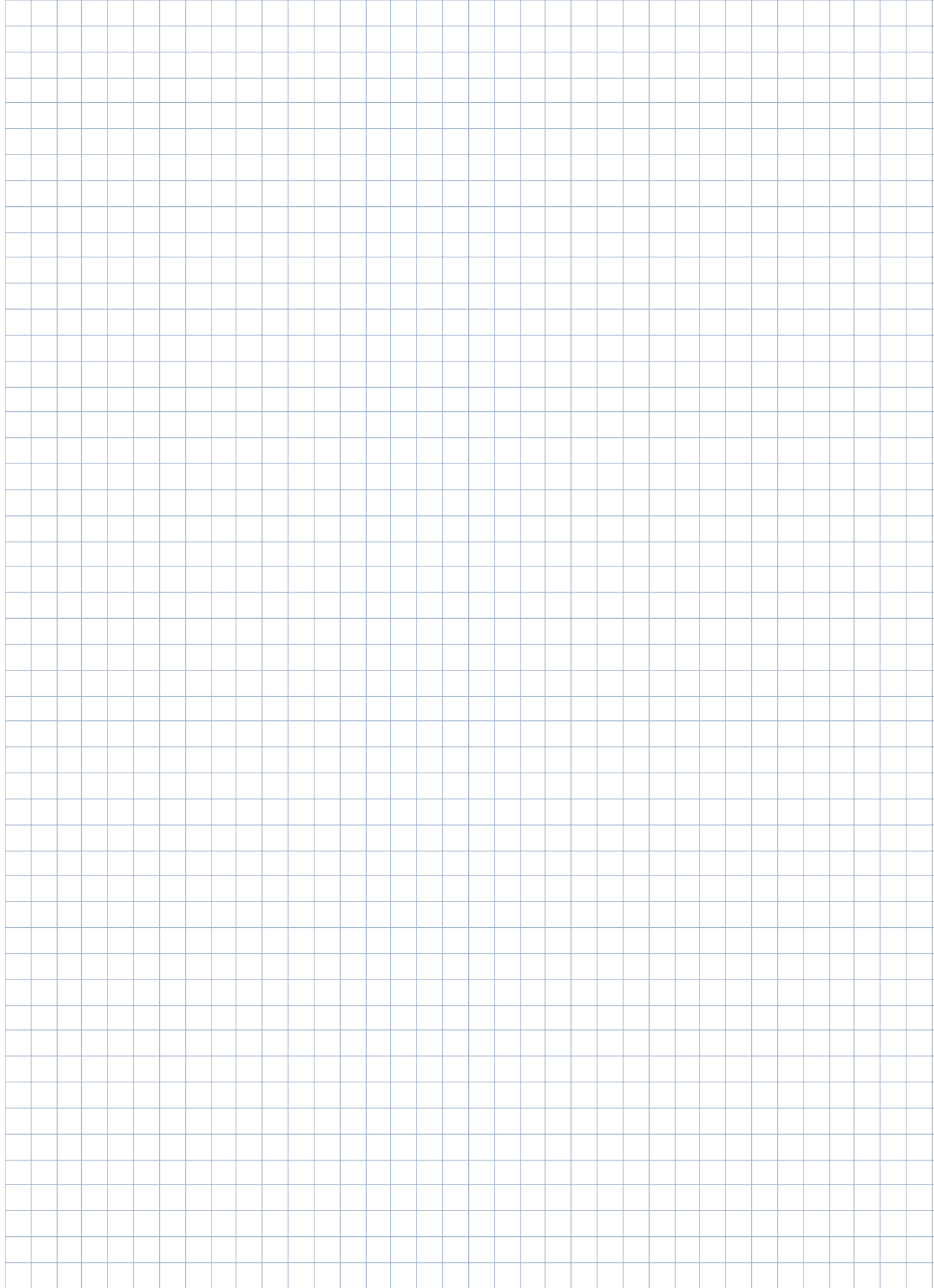
$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_{k-q} X_{k-q} + \beta_{k-q-1} X_{k-q-1} + \beta_{k-q+2} X_{k-q+2} + \dots + \beta_k X_k + u$$

(r) ur

$$F = \frac{SSR_r - SSR_{ur}}{q} \div \frac{SSR_{ur}}{(n-k-1)}$$

d.f. of the "ur" model

This is always (+)  
 b/c  $SSR_{ur} < SSR_r$   
 Every time you add 1 more  $x$ , the model will be better explained



3. Some useful facts

①  $R_{ur}^2 > R_r^2$  because any additional  $X$  would increase  $R^2$  (improve fit).  
 $\rightarrow SSR_{ur} < SSR_r$

② By including more  $X$ , the model is certainly better explained. However, we would like to reject  $H_0$  if the conclusion of extra variables does not improve the model enough.

4. Other ways to calculate the F-statistics:

$\Rightarrow$  From  $R^2 = 1 - \frac{SSR}{SST}$  <sup>RSS</sup> <sub>TSS</sub>

We have  $F \equiv \frac{(R_{ur}^2 - R_r^2)}{\frac{(1 - R_{ur}^2)}{n - k - 1}}$

# of  $\beta$  that are set to "0"  $\rightarrow$   $q$    
 intercept  $\leftarrow$   $n - k - 1$    
 $\uparrow$  # of obs.  $\leftarrow$  # of slope  $\beta$ .

$\Rightarrow$  If we want to test the overall significance of the model

$H_0: \beta_1 = \beta_2 = \beta_3 = \dots = \beta_k = 0$

$H_a = \text{otherwise}$

$F \equiv \frac{R^2/k}{(1 - R^2)/(n - k - 1)}$    
 $R^2$  of the model  $\approx$  UR the "r" model has no  $X$  at all

**Example:** Suppose we are interested in understanding the determinant of a baseball player's salary.

r	{	ur	$y$ salary = season salary
			$years$ = years in major leagues
			$gamesyr$ = games per year in the league
			$bavg$ = career batting average
			$hrunsyr$ = homeruns per year
			$rbisyr$ = runs batted in per year

If we want to test whether performance has any impact on salary

$H_0: \beta_{bavg} = \beta_{hrunsyr} = \beta_{rbisyr} = 0$

$H_a: \text{otherwise is true}$

! the unrestricted model (ur) is defined by

ur model



```
. regress log_salary years gamesyr bavg hrunsyr rbisyr
```

Source	SS	df	MS	Number of obs =	353
Model	308.989208	5	61.7978416	F( 5, 347) =	117.06
Residual	183.186327	347	.527914487	Prob > F =	0.0000
				R-squared =	0.6278
				Adj R-squared =	0.6224
Total	492.175535	352	1.39822595	Root MSE =	.72658

HW:

$$F = \frac{R^2/q}{(1-R^2)/(n-k-1)}$$

= ?

log_salary	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
years	.0688626	.0121145	5.68	0.000	.0450355 .0926898
gamesyr	.0125521	.0026468	4.74	0.000	.0073464 .0177578
bavg	.0009786	.0011035	0.89	0.376	-.0011918 .003149
hrunsyr	.0144295	.016057	0.90	0.369	-.0171518 .0460107
rbisyr	.0107657	.007175	1.50	0.134	-.0033462 .0248776
_cons	11.19242	.2888229	38.75	0.000	10.62435 11.76048

intercept

! the restricted model (r) is defined by

- When considering each of the performance X one-by-one none of them has a significant impact at 5%.

```
. regress log_salary years gamesyr
```

Source	SS	df	MS	Number of obs =	353
Model	293.864058	2	146.932029	F( 2, 350) =	259.32
Residual	198.311477	350	.566604221	Prob > F =	0.0000
				R-squared =	0.5971
				Adj R-squared =	0.5948
Total	492.175535	352	1.39822595	Root MSE =	.75273

log_salary	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
years	.071318	.012505	5.70	0.000	.0467236 .0959124
gamesyr	.0201745	.0013429	15.02	0.000	.0175334 .0228156
_cons	11.2238	.108312	103.62	0.000	11.01078 11.43683

- But when performing an F-test, performances have joint impact.

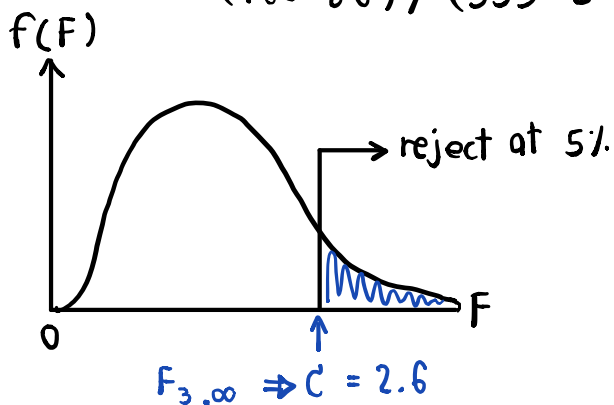
Now, our  $H_0$  and  $H_a$  becomes

$$F \equiv \frac{(SSR_r - SSR_{ur}) / q}{SSR_{ur} / (n - k - 1)}$$

$$\equiv \frac{(198.311 - 183.186) / 3}{(183.86) / (353 - 5 - 1)} \approx 9.55$$

Let's use 5% level of sig.

Since  $F = 9.55 > 2.6$ , we reject  $H_0$  at 5% level and conclude that performances have joint effects on salary.



### 8 How the Hypothesis Testing is done in Practice

1. Check the values of  $t$  " statistic reported by the statistical software (i.e. STATA, SPSS, SAS)

# These  $t$  " statistics are to test  $H_0 : \beta_i = 0$  z-table with 5% Sig. level

# If the d.f. > 30, then when  $t > 1.96$ , we can reject  $H_0$

# **When  $t > 1.96$** , we can say that  $\beta_i$  is **statistically significant** at 5% level. (value of  $\beta_i \neq 0$ )

# **When  $t < 1.96$**  we can say that  $\beta_i$  is **not statistically significant** at 5% level.

# If  $t < 1.96$  we can drop  $x_i$  from the model

# After we drop  $x_i$ , we estimate the new regression function and obtain a new set of  $\hat{\beta}$ .

2. We can also perform other hypothesis testings of interest.

e.g.  $H_0 : \beta_i = \beta_j$

or  $H_0 : \beta_i = 5$  etc.

or perform an F-test for testing multiple linear restrictions

3. Usually, in economics, the estimation results are reported using this form

Dependent Variable: log(salary)			
Independent Variables	(1)	(2)	(3)
log(sales)	.224 (.027)	.158 (.040)	.188 (.040)
log(mktval)	—	.112 (.050)	.100 (.049)
profmarg	—	-.0023 (.0022)	-.0022 (.0021)
ceoten	—	—	.0171 (.0055)
comten	—	—	-.0092 (.0033)
intercept	4.94 (0.20)	4.62 (0.25)	4.57 (0.25)
Observations	177	177	177
R-squared	.281	.304	.353

sales →

other company performance {

CEO characteristics {

↑  
like a simple regression with 1 X.