

Quantitative Methods in Finance

Tutorial, Part 5:

Testing hypotheses about regression coefficients and their linear combinations.

Example 1: We gathered a sample of data for 32 European countries for the year 2003. We have the following variables available (the data are provided in Stata Data file `health.dta`, while the programming code is given in Stata Do file `health-commands-t05.do`):

- ♦ life expectancy at birth (*LIFE*; in years);
- ♦ health expenditure per capita (*EXP*; in U.S. dollars);
- ♦ percentage of smokers among adults (*TOBACCO*);
- ♦ consumption of alcohol per capita (*ALCO*; in litres of distilled spirits).

Estimate the linear regression model: $LIFE_i = \beta_1 + \beta_2 EXP_i + \beta_3 TOBACCO_i + u_i$ and answer the questions below by applying the appropriate procedures of hypotheses testing.

- a) Does the percentage of smokers among adults affect the life expectancy at birth?
- b) Can we claim that an increase of health expenditure per capita by 100 U.S. dollars extends the life expectancy at birth by more than two months (given that the percentage of smokers among adults remains unchanged)?
- c) Test the null hypothesis that a decrease of life expectancy at birth due to an increased percentage of smokers among adults by one percentage point can be »compensated« by an increase of health expenditure per capita of 100 U.S. dollars.

Computer printout of the results in Stata:

Estimation of the model for Exercises a, b and c

```
. regress life exp tobacco
```

Source	SS	df	MS	Number of obs =	32
Model	385.751827	2	192.875914	F(2, 29) =	31.97
Residual	174.97295	29	6.03354999	Prob > F =	0.0000
Total	560.724777	31	18.087896	R-squared =	0.6880
				Adj R-squared =	0.6664
				Root MSE =	2.4563

life	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
exp	.0023323	.0003709	6.29	0.000	.0015736 .0030909
tobacco	-.2503555	.0889983	-2.81	0.009	-.4323774 -.0683335
_cons	79.62409	2.796632	28.47	0.000	73.90433 85.34384

Testing hypotheses by applying the t-test in scalar notation (Exercises a, b and c)

```
. estat vce
```

Covariance matrix of coefficients of regress model

e(V)	exp	tobacco	_cons
exp	1.376e-07		
tobacco	9.960e-06	.0079207	
_cons	-.00048708	-.24150696	7.8211512

```
. display 2*ttail(29, abs(-2.81))
.00878239
```

```
. display ttail(29, 1.78)
.04277883
```

```
. display 2*ttail(29, abs(-0.16))
.87399033
```

Testing hypotheses by applying the t-test in matrix notation (Exercises b and c)

```
. matrix beta=(e(b))'
. matrix list beta
```

```
beta[3,1]
      y1
exp    .00233227
tobacco -.25035545
_cons   79.624085
```

```
. matrix vcm=e(V)
. matrix list vcm
```

```
symmetric vcm[3,3]
      exp    tobacco    _cons
exp    1.376e-07
tobacco 9.960e-06    .0079207
_cons  -.00048708   -.24150696   7.8211512
```

```
. matrix c=(100\0\0)
. matrix list c
```

```
c[3,1]
      c1
r1    100
r2     0
r3     0
```

```
. matrix varcb=c'*vcm*c
. matrix list varcb
```

```
symmetric varcb[1,1]
      c1
c1    .00137602
```

```
. scalar secb=sqrt(varcb[1,1])
. display secb
.03709481
```

```
. matrix tb=(c'*beta-0.167)/secb
```

```

. matrix list tb

symmetric tb[1,1]
      c1
r1  1.7853439

. display ttail(29, 1.7853439)
.04233458

. matrix c=(100\1\0)
. matrix list c

c[3,1]
      c1
r1  100
r2   1
r3   0

. matrix varcb=c'*vcm*c
. matrix list varcb

symmetric varcb[1,1]
      c1
c1  .01128871

. scalar secb=sqrt(varcb[1,1])
. display secb
.10624835

. matrix tc=(c'*beta-0)/secb
. matrix list tc

symmetric tc[1,1]
      c1
r1  -.16121158

. display 2*ttail(29, abs(-.16121158))
.87304462

```

Testing a hypothesis by applying the F-test (Exercise c)

```

. gen x_transf=exp-100*tobacco

. regress life x_transf

```

Source	SS	df	MS	Number of obs = 32		
Model	385.59502	1	385.59502	F(1, 30)	=	66.05
Residual	175.129757	30	5.83765858	Prob > F	=	0.0000
Total	560.724777	31	18.087896	R-squared	=	0.6877
				Adj R-squared	=	0.6773
				Root MSE	=	2.4161

life	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x_transf	.0023683	.0002914	8.13	0.000	.0017732	.0029634
_cons	79.18374	.5903236	134.14	0.000	77.97814	80.38934

```

. display Ftail(1,29,0.026)
.87301842

```

```
. regress life exp tobacco
```

Source	SS	df	MS	Number of obs = 32		
Model	385.751827	2	192.875914	F(2, 29) = 31.97		
Residual	174.97295	29	6.03354999	Prob > F = 0.0000		
Total	560.724777	31	18.087896	R-squared = 0.6880		
				Adj R-squared = 0.6664		
				Root MSE = 2.4563		

life	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
exp	.0023323	.0003709	6.29	0.000	.0015736	.0030909
tobacco	-.2503555	.0889983	-2.81	0.009	-.4323774	-.0683335
_cons	79.62409	2.796632	28.47	0.000	73.90433	85.34384

```
. test 100*exp+tobacco=0
```

```
( 1) 100*exp + tobacco = 0

F( 1, 29) = 0.03
Prob > F = 0.8730
```

■

Example 2: We analyse an aggregate consumption function for the period 1965–1989, which has the following specification (the data are provided in Stata Data file `consumption.dta`, while the programming code is given in Stata Do file `consumption-commands.do`):

$$C_t = \beta_1 + \beta_2 Y_t + \beta_3 ST_t + \beta_4 TR_t + \beta_5 FT_t + u_t,$$

where C represents the consumption of households, Y denotes personal incomes, ST represents social transfers, TR denotes other transfers, and FT represents foreign transfers (all in mil. monetary units and in 1972 prices).

Verify with the trinity of econometric tests whether we can exclude other transfers and foreign transfers from the above model specification.

Computer printout of the results in Stata:

Estimation of the model

```
. regress c y st tr ft
```

Source	SS	df	MS	Number of obs = 25		
Model	3.69668581	4	.924171451	F(4, 20) = 130.82		
Residual	.141293465	20	.007064673	Prob > F = 0.0000		
Total	3.83797927	24	.159915803	R-squared = 0.9632		
				Adj R-squared = 0.9558		
				Root MSE = .08405		

c	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
y	.3981591	.1938749	2.05	0.053	-.0062569	.802575

st		1.628632	.5127575	3.18	0.005	.559039	2.698226
tr		7.506952	6.290407	1.19	0.247	-5.614608	20.62851
ft		.4450372	.5246166	0.85	0.406	-.6492939	1.539368
_cons		.3635363	.1152333	3.15	0.005	.1231638	.6039088

Testing a reduction of the model with the F-test

. regress c y st

Source		SS	df	MS		Number of obs =	25
Model		3.67855587	2	1.83927794		F(2, 22) =	253.82
Residual		.1594234	22	.007246518		Prob > F =	0.0000
Total		3.83797927	24	.159915803		R-squared =	0.9585
						Adj R-squared =	0.9547
						Root MSE =	.08513

c		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
y		.6648524	.0872809	7.62	0.000	.4838429	.8458619
st		.9483866	.283467	3.35	0.003	.360512	1.536261
_cons		.2653599	.0937114	2.83	0.010	.0710144	.4597054

. display ((.1594234-.141293465)/2)/(.141293465/(25-5))
1.2831404

. display Ftail(2,20,1.2831404)
.29901989

. qui regress c y st tr ft
. test tr=ft=0

(1) tr - ft = 0
(2) tr = 0

F(2, 20) = 1.28
Prob > F = 0.2990

Testing a reduction of the model with the LM-test

. regress c y st

Source		SS	df	MS		Number of obs =	25
Model		3.67855587	2	1.83927794		F(2, 22) =	253.82
Residual		.1594234	22	.007246518		Prob > F =	0.0000
Total		3.83797927	24	.159915803		R-squared =	0.9585
						Adj R-squared =	0.9547
						Root MSE =	.08513

c		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
y		.6648524	.0872809	7.62	0.000	.4838429	.8458619
st		.9483866	.283467	3.35	0.003	.360512	1.536261
_cons		.2653599	.0937114	2.83	0.010	.0710144	.4597054

. predict test, resid

```
. regress test y st tr ft
```

Source	SS	df	MS	Number of obs = 25		
Model	.018129935	4	.004532484	F(4, 20) = 0.64		
Residual	.141293467	20	.007064673	Prob > F = 0.6391		
Total	.159423401	24	.006642642	R-squared = 0.1137		
				Adj R-squared = -0.0635		
				Root MSE = .08405		

test	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
y	-.2666934	.1938749	-1.38	0.184	-.6711094	.1377226
st	.6802458	.5127575	1.33	0.200	-.3893476	1.749839
tr	7.506952	6.290407	1.19	0.247	-5.614608	20.62851
ft	.4450372	.5246166	0.85	0.406	-.6492939	1.539368
_cons	.0981764	.1152333	0.85	0.404	-.1421961	.338549

```
. scalar lm=25*0.1137
```

```
. display lm
```

```
2.8425
```

```
. display chi2tail(2,lm)
```

```
.24141206
```

Testing a reduction of the model with the LR-test

```
. regress c y st tr ft
```

Source	SS	df	MS	Number of obs = 25		
Model	3.69668581	4	.924171451	F(4, 20) = 130.82		
Residual	.141293465	20	.007064673	Prob > F = 0.0000		
Total	3.83797927	24	.159915803	R-squared = 0.9632		
				Adj R-squared = 0.9558		
				Root MSE = .08405		

c	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
y	.3981591	.1938749	2.05	0.053	-.0062569	.802575
st	1.628632	.5127575	3.18	0.005	.559039	2.698226
tr	7.506952	6.290407	1.19	0.247	-5.614608	20.62851
ft	.4450372	.5246166	0.85	0.406	-.6492939	1.539368
_cons	.3635363	.1152333	3.15	0.005	.1231638	.6039088

```
. scalar llb=e(ll)
```

```
. display llb
```

```
29.223937
```

```
. display -25/2*(ln(2*_pi)+ln(0.141293465/25)+1)
```

```
29.223937
```

```
. regress c y st
```

Source	SS	df	MS	Number of obs = 25		
Model	3.67855587	2	1.83927794	F(2, 22) = 253.82		
Residual	.1594234	22	.007246518	Prob > F = 0.0000		
Total	3.83797927	24	.159915803	R-squared = 0.9585		
				Adj R-squared = 0.9547		
				Root MSE = .08513		

c	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
y	.6648524	.0872809	7.62	0.000	.4838429	.8458619
st	.9483866	.283467	3.35	0.003	.360512	1.536261
_cons	.2653599	.0937114	2.83	0.010	.0710144	.4597054

```

. scalar llr=e(ll)
. display llr
27.714881

. display -25/2*(ln(2*_pi)+ln(0.1594234/25)+1)
27.714881

. display 2*(llb-llr)
3.0181129

. display chi2tail(2, 2*(llb-llr))
.22111852

. qui regress c y st tr ft
. estimates store mb

. qui regress c y st
. estimates store mr

. lrtest mr mb

```

Likelihood-ratio test
(Assumption: mr nested in mb)

LR chi2(2) = 3.02
Prob > chi2 = 0.2211

■