

Lab 3 (Hypothesis testing)

This file uses the same data on bid-ask spreads (tradedcost) and candidate determinants used in lab2 HW2.

. Import the data set and set up variable names

```
% import dataset lab2.xls (using Import Dataset)

% rename variables
tradedcost = lab2{ : ,1};
logvolatility= lab2{ : ,2};
logsize = lab2{ : ,3};
logtrades = lab2{ : ,4};
logturn = lab2{ : ,5};
numberanalysts = lab2{ : ,6};
logtradedcost = log(tradedcost);
```

We run a regression of the log of the bid-ask spread on the 5 explanatory variables, reproduced below.

We interpreted the signs of the estimated coefficients of the 5 explanatory variables in light of the theories outlined in the background information of lab2 HW2..

```
x1 = table(logtradedcost,logvolatility,logsize,logtrades,logturn,numberanalysts);
lm1 =fitlm(x1,'logtradedcost ~ logvolatility + logsize + logtrades + logturn + numberanalysts')
```

```
lm1 =
Linear regression model:
logtradedcost ~ 1 + logvolatility + logsize + logtrades + logturn + numberanalysts
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.82932	0.44277	-1.873	0.064173
logvolatility	1.023	0.053245	19.213	2.574e-34
logsize	-0.14029	0.02391	-5.8673	6.5918e-08
logtrades	-0.16936	0.035596	-4.758	7.0564e-06
logturn	-0.098714	0.032502	-3.0371	0.0030911
numberanalysts	5.044e-05	0.0029135	0.017313	0.98622

```
Number of observations: 100, Error degrees of freedom: 94
Root Mean Squared Error: 0.142
R-squared: 0.882, Adjusted R-Squared: 0.876
F-statistic vs. constant model: 141, p-value = 4.16e-42
```

coefTest

Linear hypothesis test on linear regression model coefficients

Description

`p = coefTest(md1)` computes the p -value for an F -test that all coefficient estimates in `md1`, except for the intercept term, are zero.

`p = coefTest(md1,H)` performs an F -test that $H \times B = 0$, where B represents the coefficient vector. Use `H` to specify the coefficients to include in the F -test.

`p = coefTest(md1,H,C)` performs an F -test that $H \times B = C$.

Note: in Lecture 4, $H = R$, and the vector of beta coefficient is $B = \text{Beta}$, and C is the r vector

Examples:

1. Test the null that all coefficients are 0.

```
[p1,F1] = coefTest(lm1)
```

```
p1 = 4.1630e-42  
F1 = 141.1555
```

These match with the output

2. Test the assumption that the coefficients associated with Logsize and Logturn are equal to each other.

```
H = [0 0 1 0 -1 0];  
%p= coefTest(lm1,H)
```

```
[p2,F2] = coefTest(lm1,H)
```

```
p2 = 0.2863  
F2 = 1.1499
```

H is a 1×5 vector, B is the 5×1 vector of coefficients, therefore $H \times B = 0$ results in an equation.

$H \times B = 0$ means in this case: $\text{beta3} - \text{beta5} = 0$

3. Test the assumption that the coefficients associated with all variables except the constant are 0.

```
H = [0 1 1 1 1 1];  
[p3,F3] = coefTest(lm1, H)
```

```
p3 = 1.2639e-17  
F3 = 111.2644
```

4. Test the assumption that the coefficients associated with number analysts is 0.

```
H = [0 0 0 0 0 1];  
[p4,F4] = coefTest(lm1, H)
```

```
p4 = 0.9862  
F4 = 2.9973e-04
```

md1 — Linear regression model object

LinearModel object | CompactLinearModel object

Linear regression model object, specified as a [LinearModel](#) object created by using [fitlm](#) or [stepwiselm](#), or a [CompactLinearModel](#) object created by using [compact](#).

H — Hypothesis matrix

numeric index matrix

Hypothesis matrix, specified as an r-by-s numeric index matrix, where r is the number of coefficients to include in an *F*-test, and s is the total number of coefficients.

- If you specify H, then the output [p](#) is the *p*-value for an *F*-test that $H \times B = 0$, where *B* represents the coefficient vector.
- If you specify H and [C](#), then the output *p* is the *p*-value for an *F*-test that $H \times B = C$.

Example: [1 0 0 0 0] tests the first coefficient among five coefficients

Data Types: single | double

C — Hypothesized value

numeric vector

Hypothesized value for testing the null hypothesis, specified as a numeric vector with the same number of rows as [H](#).

If you specify H and C, then the output [p](#) is the *p*-value for an *F*-test that $H \times B = C$, where *B* represents the coefficient vector.

Data Types: single | double

Output Arguments

[collapse all](#)

p — *p*-value for *F*-test

numeric value in the range [0,1]

p-value for the *F*-test, returned as a numeric value in the range [0,1].

F — Value of test statistic for *F*-test

numeric value

Value of the test statistic for the *F*-test, returned as a numeric value.

r — Numerator degrees of freedom for *F*-test

positive integer

Numerator degrees of freedom for the *F*-test, returned as a positive integer. The *F*-statistic has *r* degrees of freedom in the numerator and mdl.DFE degrees of freedom in the denominator.

Algorithms

The *p*-value, *F*-statistic, and numerator degrees of freedom are valid under these assumptions:

- The data comes from a model represented by the formula in the Formula property of the fitted model.
- The observations are independent, conditional on the predictor values.

Under these assumptions hold, let β represent the (unknown) coefficient vector of the linear regression.

Suppose *H* is a full-rank matrix of size *r*-by-*s*, where *r* is the number of coefficients to include in an *F*-test, and *s* is the total number of coefficients. Let *c* be a vector the same size as β . The following is a test statistic for the hypothesis that $H\beta = c$:

$$F = (H^* \hat{\beta} - v)' (HCH)^{-1} (H^* \hat{\beta} - v).$$

Here $\hat{\beta}$ is the estimate of the coefficient vector β , stored in the Coefficients property, and *V* is the estimated covariance of the coefficient estimates, stored in the CoefficientCovariance property. When the hypothesis is true, the test statistic *F* has an [F Distribution](#) with *r* and *u* degrees of freedom, where *u* is the degrees of freedom for error, stored in the DFE property.

3. Test the assumption that the coefficients associated with all variables except the constant are 0 (general formulation).

```
H = [0 0 0 0 0 0; 0 1 0 0 0 0; 0 0 1 0 0 0; 0 0 0 1 0 0; 0 0 0 0 1 0; 0 0 0 0 0 1 ]
```

H = 6×6

0	0	0	0	0	0
0	1	0	0	0	0
0	0	1	0	0	0
0	0	0	1	0	0
0	0	0	0	1	0
0	0	0	0	0	1

```
C = [0; 0; 0; 0; 0; 0]
```

C = 6×1

```
0
0
0
0
0
0
0
```

```
[p5,F5] = coefTest(lm1, H, C)
```

```
p5 = 4.1630e-42
F5 = 141.1555
```

5. Test the assumption that the coefficient associated with log size equals -0.5.

```
H = [0 0 0 0 0 0; 0 0 0 0 0 0; 0 0 1 0 0 0; 0 0 0 0 0 0; 0 0 0 0 0 0; 0 0 0 0 0 0 ]
```

```
H = 6×6
    0    0    0    0    0    0
    0    0    0    0    0    0
    0    0    1    0    0    0
    0    0    0    0    0    0
    0    0    0    0    0    0
    0    0    0    0    0    0
```

```
C = [0; 0; -0.5; 0; 0; 0]
```

```
C = 6×1
    0
    0
 -0.5000
    0
    0
    0
```

```
[p6,F6] = coefTest(lm1,H,C)
```

```
p6 = 9.1569e-27
F6 = 226.3294
```