

## Estimating CAPM using GMM

## Testing the CAPM in a GMM setting

Consider the CAPM relationship, specified by the “moment condition”

$$E[er_{it}] = \beta_i E[er_{mt}]$$

where  $er_{it}$  is the excess return on an asset or a portfolio, and  $er_{mt}$  the excess return on the market portfolio.

One can alternatively specify the CAPM more generally as

$$E[r_{it}] = E[r_{zt}] + \beta_i (E[r_{mt}] - E[r_{zt}])$$

where we let  $r_{zt}$  be the return on a “zero covariance” portfolio.

The typical way of testing this relationship, the Black et al. [1972] method, estimates the corresponding regression

$$er_{it} = \alpha_i + \beta_i er_{mt} + \varepsilon_{it}$$

and tests whether  $\alpha_i = 0$  on an equation by equation basis. Econometrically wasteful: the restriction on the constant will hold for all assets, one want to test this jointly, in a multivariate setting. Doing this was proposed by Gibbons [1982], who showed how one could construct a multivariate statistic for testing this. His method was later expanded upon by Gibbons et al. [1989]. These statistic were developed under distributional assumptions that allowed us to use Maximum Likelihood, namely multivariate normality.

What if these are not fulfilled, can we still construct a similar test statistic?

This is done in MacKinlay and Richardson [1991] (MR). They construct a test statistic that essentially tests the same restriction, that  $\alpha_j = 0$  or that  $\alpha_j = E[r_{zt}](1 - \beta_j)$ , but in a GMM framework, not a ML.

The setup:

The usual regression

$$r_{it} = \alpha_i + \beta_i r_{mt} + \varepsilon_{it}$$

Assume that

$$E[\varepsilon_{it}|r_{mt}] = 0$$

Two moment restrictions for each asset  $i$ :

$$E[\varepsilon_{it}] = E[(r_{it} - \alpha_i - \beta_i r_{mt})] = 0$$

$$E[\varepsilon_{it} r_{mt}] = E[(r_{it} - \alpha_i - \beta_i r_{mt}) r_{mt}] = 0$$

The model is exactly identified. We can “stack” these moment conditions and estimate the parameters  $\{\alpha_i, \beta_i\}$  of the model, by the usual formulation using sample moments.

The tests discussed in the paper are different ways of testing the parametric restriction  $\alpha_j = 0$ .

We will show how the GMM framework can be used to test the CAPM in a couple of examples involving US and Norwegian Data.

## Exercise

Consider the 5 (ew) size based portfolios at the US. Estimate the CAPM using GMM with the Mackinlay method. Use data 1926-2012.

1. Does the CAPM seem like a sufficient model for pricing this crosssection of stock returns?

## Exercise Solution

R code to do analysis

Read data

```
source("read_pricing_factors.R")
source("read_size_portfolios.R")
eRi <- FFSize5EW - RF
data <- merge(eRi,RMRF,all=FALSE)
eRi <- as.matrix(data[,1:5])
eRm <- as.matrix(data[,6])
```



Index	Lo20	Qnt2	
Min. :1926	Min. :-32.010	Min. :-31.9600	Min.
1st Qu.:1948	1st Qu.: -3.110	1st Qu.: -2.8650	1st Qu.
Median :1970	Median : 0.960	Median : 1.1700	Median
Mean :1970	Mean : 1.373	Mean : 0.9704	Mean
3rd Qu.:1991	3rd Qu.: 4.695	3rd Qu.: 4.5350	3rd Qu.
Max. :2013	Max. :110.670	Max. : 81.1900	Max.
Qnt4	Hi20	RMRF	
Min. :-29.760	Min. :-30.100	Min. :-28.980	
1st Qu.: -2.470	1st Qu.: -2.195	1st Qu.: -2.105	
Median : 1.160	Median : 0.930	Median : 1.010	
Mean : 0.787	Mean : 0.655	Mean : 0.628	
3rd Qu.: 4.125	3rd Qu.: 3.640	3rd Qu.: 3.655	
Max. : 50.010	Max. : 41.790	Max. : 37.770	

## Exercise Solution

First look at the OLS estimates,  
low20: (smallest size)

Call:

```
lm(formula = Lo20 ~ eRm)
```

Residuals:

Min	1Q	Median	3Q	Max
-18.245	-2.920	-0.618	1.890	76.656

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.47537	0.19852	2.395	0.0168 *
eRm	1.42903	0.03633	39.334	<2e-16 ***

Residual standard error: 6.344 on 1033 degrees of freedom

Multiple R-squared: 0.5996, Adjusted R-squared: 0.5992

F-statistic: 1547 on 1 and 1033 DF, p-value: < 2.2e-16

## Exercise Solution

	1	2	3	4	5
(Intercept)	0.475** (0.199)	0.096 (0.077)	0.096 (0.077)	0.056 (0.051)	-0.004 (0.029)
eRm	1.429*** (0.036)	1.260*** (0.014)	1.260*** (0.014)	1.164*** (0.009)	1.049*** (0.005)
Adj. R <sup>2</sup>	0.599	0.886	0.886	0.939	0.974
Num. obs.	1035	1035	1035	1035	1035

\*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.1$

## Exercise Solution

```
gmm(g = eRi ~ eRm, x = eRm)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(>
Lo20_(Intercept)	4.7537e-01	1.8122e-01	2.6232e+00	8.
Qnt2_(Intercept)	1.3007e-01	1.0484e-01	1.2406e+00	2.
Qnt3_(Intercept)	9.5565e-02	7.1679e-02	1.3332e+00	1.
Qnt4_(Intercept)	5.6185e-02	5.1781e-02	1.0850e+00	2.
Hi20_(Intercept)	-3.7811e-03	2.7477e-02	-1.3761e-01	8.
Lo20_eRm	1.4290e+00	1.0552e-01	1.3543e+01	8.
Qnt2_eRm	1.3382e+00	5.9089e-02	2.2646e+01	1.5
Qnt3_eRm	1.2601e+00	3.6174e-02	3.4833e+01	7.7
Qnt4_eRm	1.1637e+00	1.9760e-02	5.8889e+01	0.
Hi20_eRm	1.0490e+00	1.1175e-02	9.3867e+01	0.

J-Test: degrees of freedom is 0

	J-test	P-value
Test E(g)=0:	1.7729137049673e-23	*****

## Exercise Solution

	Model 1
Lo20 (Intercept)	0.475 (0.181)***
Qnt2 (Intercept)	0.130 (0.105)
Qnt3 (Intercept)	0.096 (0.072)
Qnt4 (Intercept)	0.056 (0.052)
Hi20 (Intercept)	-0.004 (0.027)
Lo20 eRm	1.429 (0.106)***
Qnt2 eRm	1.338 (0.059)***
Qnt3 eRm	1.260 (0.036)***
Qnt4 eRm	1.164 (0.020)***
Hi20 eRm	1.049 (0.011)***
Criterion function	0.000
Num. obs.	1035

\*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.1$

## Exercise Solution

Now, let us get to the test of whether the intercept is zero:  
Need to formulate the linear restrictions in matrix form.

```
> R <- cbind(diag(5),matrix(0,5,5))
> c <- rep(0,5)
> print(R)
      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10]
[1,]    1    0    0    0    0    0    0    0    0    0
[2,]    0    1    0    0    0    0    0    0    0    0
[3,]    0    0    1    0    0    0    0    0    0    0
[4,]    0    0    0    1    0    0    0    0    0    0
[5,]    0    0    0    0    1    0    0    0    0    0
> print(c)
[1] 0 0 0 0 0
```

## Exercise Solution

and then perform the test:

```
> linearHypothesis(res,R,c,test="F")
```

Linear hypothesis test

Hypothesis:

```
Lo20_((Intercept) = 0
```

```
Qnt2_((Intercept) = 0
```

```
Qnt3_((Intercept) = 0
```

```
Qnt4_((Intercept) = 0
```

```
Hi20_((Intercept) = 0
```

Model 1: restricted model

Model 2:  $eR_i \sim eR_m$

	Df	Chisq	Pr(>Chisq)
1			
2	5	13.588	0.01845 *

Fisher Black, Michael Jensen, and Myron Scholes. The capital asset pricing model, some empirical tests. In Michael C Jensen, editor, *Studies in the theory of capital markets*. Preager, 1972.

Michael R Gibbons. Multivariate tests of financial models, a new approach. *Journal of Financial Economics*, 10:3–27, March 1982.

Michael R Gibbons, Stephen A Ross, and Jay Shanken. A test of the efficiency of a given portfolio. *Econometrica*, 57:1121–1152, 1989.

A Craig MacKinlay and Matthew P Richardson. Using generalized method of moments to test mean-variance efficiency. *Journal of Finance*, 46:511–27, 1991.