

5th set of assignments Financial Econometrics- solutions

5.1 using avewret

System: CAPM
Estimation Method: Generalized Method of Moments
Date: 06/21/05 Time: 17:11
Sample: 1947:2 1993:4
Included observations: 187
Total system (balanced) observations 1870
Kernel: Bartlett, Bandwidth: Fixed (4), No prewhitening
Identity weighting matrix - 2SLS coefficients with GMM standard errors

	Coefficient	Std. Error	t-Statistic	Prob.
C(1)	1.067309	0.056669	18.83400	0.0000
C(2)	-3.211099	1.694037	-1.895531	0.0582
Determinant residual covariance	9.79E-35			
J-statistic	0.011896			

J-Statistics: (for calculation: see assignment sheet)

jval = 2.2245

pval = 0.9733

5.1 using avvwret

System: CAPM2
Estimation Method: Generalized Method of Moments
Date: 06/21/05 Time: 17:29
Sample: 1947:2 1993:4
Included observations: 187
Total system (balanced) observations 1870
Kernel: Bartlett, Bandwidth: Fixed (4), No prewhitening
Identity weighting matrix - 2SLS coefficients with GMM standard errors

	Coefficient	Std. Error	t-Statistic	Prob.
C(1)	1.132061	0.088452	12.79854	0.0000
C(2)	-6.997747	3.561610	-1.964771	0.0496
Determinant residual covariance	1.38E-33			
J-statistic	0.010975			

J-Statistics:

jval2 = 2.0523

pval2 = 0.9794

5.2 using avewret

System: CAPM_EX
Estimation Method: Generalized Method of Moments
Date: 06/21/05 Time: 17:33
Sample: 1947:2 1993:4
Included observations: 187
Total system (balanced) observations 1870
Kernel: Bartlett, Bandwidth: Fixed (4), No prewhitening
Identity weighting matrix - 2SLS coefficients with GMM standard errors

	Coefficient	Std. Error	t-Statistic	Prob.
C(2)	-3.635131	0.800801	-4.539369	0.0000

Determinant residual covariance 2.03E-36
J-statistic 0.013363

J-Statistics:

jval_ex = 2.4988
pval_ex = 0.9809

5.2 using avvwret

System: CAPM_EX2
Estimation Method: Generalized Method of Moments
Date: 06/21/05 Time: 17:34
Sample: 1947:2 1993:4
Included observations: 187
Total system (balanced) observations 1870
Kernel: Bartlett, Bandwidth: Fixed (4), No prewhitening
Identity weighting matrix - 2SLS coefficients with GMM standard errors

	Coefficient	Std. Error	t-Statistic	Prob.
C(2)	-4.924601	1.082375	-4.549809	0.0000

Determinant residual covariance 2.49E-36
J-statistic 0.013145

J-Statistics:

jval_ex2 = 2.4581
pval_ex2 = 0.9820

5.3 using avewret

System: FAMA_FRENCH

Estimation Method: Generalized Method of Moments

Date: 06/21/05 Time: 17:53

Sample: 1947:2 1993:4

Included observations: 187

Total system (balanced) observations 1870

Kernel: Bartlett, Bandwidth: Fixed (4), No prewhitening

Identity weighting matrix - 2SLS coefficients with GMM standard errors

	Coefficient	Std. Error	t-Statistic	Prob.
C(1)	0.631875	0.357282	1.768560	0.0771
C(2)	-0.257423	0.393292	-0.654534	0.5128
C(3)	-0.853027	0.416109	-2.050007	0.0405
Determinant residual covariance	4.57E-64			
J-statistic	0.017830			

J-Statistics:

jval_fafr = 3.3343

pval_fafr = 0.8525

Wald Test:

System: FAMA_FRENCH

Test Statistic	Value	df	Probability
Chi-square	9.944229	2	0.0069

5.3 using avvwret

System: FAMA_FRENCH2

Estimation Method: Generalized Method of Moments

Date: 06/21/05 Time: 17:53

Sample: 1947:2 1993:4

Included observations: 187

Total system (balanced) observations 1870

Kernel: Bartlett, Bandwidth: Fixed (4), No prewhitening

Identity weighting matrix - 2SLS coefficients with GMM standard errors

	Coefficient	Std. Error	t-Statistic	Prob.
C(1)	-0.542606	0.493584	-1.099319	0.2718
C(2)	-0.776924	0.448708	-1.731467	0.0835
C(3)	-0.101936	0.447192	-0.227947	0.8197

Determinant residual covariance

3.05E-60

J-statistic

0.012763

J-Statistics:

jval_fafr2 = 2.3868

pval_fafr2 = 0.9354

Wald Test:

System: FAMA_FRENCH2

Test Statistic	Value	df	Probability
Chi-square	1855.244	3	0.0000

6th set of assignments Financial Econometrics – solutions

pooled time series regression using avvwret

- Sample(adjusted)! 1947:2-1993:4
- *series factor1b = avvwret-avustret
series factor2 = hml_r-1
series factor3 = smb_r-1*
- Objects -> new objects -> pool: pooled time series regression
Dependent variable: decile?-avustret
Cross section specific coefficients: c factor1b factor2 factor3
No intercept, no weighting -> estimate
Make residuals
matrix varcovres6b = @cov(residuals6b)

- group factors: factors6b
- *matrix varcovfactor6b = @cov(factors6b)*
- create mean vector:
*vector(3) meanfac6b
meanfac6b.fill @mean(factor1b), @mean(factor2), @mean(factor3)*

0.018130
0.002613
-0.005883

- GRS statistic:

alpha: see assignment sheet
one: see assignment sheet

*matrix grsb =
t*@inverse(one+(@transpose(meanfac6b)*@inverse(varcovfactor6b)*meanfac6b)) *
@transpose(alpha6b)*@inverse(varcovres6b)*alpha6b*

GRS- Statistic: 10.545

- p-value :
*matrix pavelgrs6b = one- @cchisq(grsb, 10)
p-value = 0.394*

7th set of assignments Financial Econometrics – solutions

1. Alternative1:

Construct a system as in assignment sheet 5:

System: ALTERNATIVE1

Estimation Method: Generalized Method of Moments

Date: 06/23/05 Time: 14:08

Sample: 1947:2 1993:4

Included observations: 187

Total system (balanced) observations 1870

Kernel: Bartlett, Bandwidth: Fixed (4), No prewhitening

2SLS coefficient estimates with GMM standard errors

	Coefficient	Std. Error	t-Statistic	Prob.
C(2)	-0.990917	0.001132	-875.1660	0.0000

Determinant residual 9.26E-79

covariance

J-statistic 0.052227

2. Alternative 2:

- Sample(adjusted): 1947:2 1993:4
- Objects -> new objects -> pool: pooled time series regression: assets
Dependent variable: decile?-avustret
Cross section specific coefficients: c cnsqdifference
No intercept, no weighting -> estimate
Make residuals
matrix varcovres = @cov(residuals)

vector beta = @subextract(assets. @coefs, 11, 1, 20, 1)

4.966555
4.464380
4.248438
3.798502
3.517813
3.172752
3.038945
2.826603
2.220660
2.231426

- *vector(10) avreturn*
avreturn.fill @mean(decile1-avustret), @mean(decile2-avustret), @mean(decile3-avustret), @mean(decile4-avustret), @mean(decile5-avustret), @mean(decile6-avustret), @mean(decile7-avustret), @mean(decile8-avustret), @mean(decile9-avustret), @mean(decile10-avustret)

- compute lambda

$$\text{matrix } \lambda = \text{@inverse}(\text{@transpose}(\beta)^*\beta)^*\text{@transpose}(\beta)^*\alpha$$

$$\lambda = 0.006450$$
- compute alpha

$$\text{matrix } \alpha = \alpha - \beta^*\lambda$$

- Separate the computation of the variance-covariance matrix of alpha:

First part:

```
scalar t = @regobs
matrix firstalpha = 1/t*(ones-
beta^@\text{inverse}(@\text{transpose}(\beta)^*\beta)^*\text{@\text{transpose}}(\beta))^*\text{varcovres}*(ones-
beta^@\text{inverse}(@\text{transpose}(\beta)^*\beta)^*\text{@\text{transpose}}(\beta))
```

Second part:

```
matrix varfactor = @var(cnsqdiff)
matrix lastalpha = (one+ @\text{transpose}(\lambda)^*\text{@\text{inverse}}(varfactor)^*\lambda)
create a scalar lastalpha3 from matrix lastalpha:
scalar lastalpha3 = 2.34757952912
```

compute the symmetric variance-covariance matrix of alpha by multiplying both parts:

$$\text{sym varcovalpha} = \text{lastalpha3} * \text{firstalpha}$$

since the variance-covariance matrix of alpha is not invertable: compute the pseudo-inverse for the test statistic

```
scalar rang = @rank(varcovalpha)
matrix v
vector w
matrix u=@svd(varcovalpha,w,v)
matrix w1=@makediagonal(w)
matrix w2=@subextract(w1,1,1,rang,rang)
matrix v2=@subextract(@\text{transpose}(v),1,1,rang,10)
matrix u2=@subextract(@\text{transpose}(u),1,1,rang,10)
```

```
matrix pseudoinvcoalph=@\text{transpose}(v2)^*\text{@\text{inverse}}(@\text{transpose}(w2)^*w2)^*
@\text{transpose}(w2)^*u2
```

- compute the test statistic

```
matrix statistic7=@\text{transpose}(\alpha)^*\text{pseudoinvcoalph}*\alpha
test statistic = 6.964663
p-value = 0.640799
```

3. Alternative3: (Does not work with Eviews)

Construct a system as follows:

```
(decile1-avustret)-c(1)-c(2)*cnsqdiff=0  
((decile1-avustret)-c(1)-c(2)*cnsqdiff)*cnsqdiff=0  
(decile2-avustret)-c(3)-c(4)*cnsqdiff=0  
((decile2-avustret)-c(3)-c(4)*cnsqdiff)*cnsqdiff=0  
(decile3-avustret)-c(5)-c(6)*cnsqdiff=0  
((decile3-avustret)-c(5)-c(6)*cnsqdiff)*cnsqdiff=0  
(decile4-avustret)-c(7)-c(8)*cnsqdiff=0  
((decile4-avustret)-c(7)-c(8)*cnsqdiff)*cnsqdiff=0  
(decile5-avustret)-c(9)-c(10)*cnsqdiff=0  
((decile5-avustret)-c(9)-c(10)*cnsqdiff)*cnsqdiff=0  
(decile6-avustret)-c(11)-c(12)*cnsqdiff=0  
((decile6-avustret)-c(11)-c(12)*cnsqdiff)*cnsqdiff=0  
(decile7-avustret)-c(13)-c(14)*cnsqdiff=0  
((decile7-avustret)-c(13)-c(14)*cnsqdiff)*cnsqdiff=0  
(decile8-avustret)-c(15)-c(16)*cnsqdiff=0  
((decile8-avustret)-c(15)-c(16)*cnsqdiff)*cnsqdiff=0  
(decile9-avustret)-c(17)-c(18)*cnsqdiff=0  
((decile9-avustret)-c(17)-c(18)*cnsqdiff)*cnsqdiff=0  
(decile10-avustret)-c(19)-c(20)*cnsqdiff=0  
((decile10-avustret)-c(19)-c(20)*cnsqdiff)*cnsqdiff=0  
(decile1-avustret)-c(2)*c(21)=0  
(decile2-avustret)-c(4)*c(21)=0  
(decile3-avustret)-c(6)*c(21)=0  
(decile4-avustret)-c(8)*c(21)=0  
(decile5-avustret)-c(10)*c(21)=0  
(decile6-avustret)-c(12)*c(21)=0  
(decile7-avustret)-c(14)*c(21)=0  
(decile8-avustret)-c(16)*c(21)=0  
(decile9-avustret)-c(18)*c(21)=0  
(decile10-avustret)-c(20)*c(21)=0  
  
param C(1) -4.960049 C(2) 4.966816 C(3) -4.459504 C(4) 4.464615 C(5) -4.243049  
C(6) 4.248657 C(7) -3.791239 C(8) 3.798705 C(9) -3.511227 C(10) 3.518007  
C(11) -3.163616 C(12) 3.172926 C(13) -3.030564 C(14) 3.039106 C(15)  
-2.817843 C(16) 2.826760 C(17) -2.210248 C(18) 2.220781 C(19) -2.225134  
C(20) 2.231552 c(21) 1  
  
@inst c
```