

Solutions to Selected Computer Lab Problems and Exercises in Chapter 15 of *Statistics and Data Analysis for Financial Engineering, 2nd ed.* by David Ruppert and David S. Matteson

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Problem 3. The output is below. If we use the 1% cutoff, then we accept that $r \leq 4$ but reject that $r \leq 3$ and so conclude that the number of cointegration vectors is $r = 4$.

```
> library(urca)
> mk.maturity = read.csv("mk.zero2.csv", header=T)
> options(digits=3,width=60)
> yields.cajo = ca.jo(mk.maturity[,2:11])
> summary(yields.cajo)

#####
# Johansen-Procedure #
#####

Test type: maximal eigenvalue statistic (lambda max) , with linear trend

Eigenvalues (lambda):
[1] 0.7518 0.6630 0.5981 0.5498 0.4675 0.4263 0.3486 0.2474
[9] 0.1162 0.0218

Values of teststatistic and critical values of test:

      test 10pct  5pct 1pct
r <= 9 |  1.43   6.5  8.18 11.6
r <= 8 |  8.03  12.9 14.90 19.2
r <= 7 | 18.48  18.9 21.07 25.8
r <= 6 | 27.87  24.8 27.14 32.1
r <= 5 | 36.12  30.8 33.32 38.8
r <= 4 | 40.96  36.2 39.43 44.6
r <= 3 | 51.87  42.1 44.91 51.3
r <= 2 | 59.25  48.4 51.07 57.1
r <= 1 | 70.71  54.0 57.00 63.4
r = 0  | 90.57  59.0 62.42 68.6

Eigenvectors, normalised to first column:
(These are the cointegration relations)

      M.1.12  M.2.12  M.3.12  M.4.12  M.5.12  M.6.12
M.1.12  1.0000  1.0000  1.000  1.000  1.0  1.0000
M.2.12 -4.7344  0.0688 -5.226 -2.685 -1.1 -3.7018
M.3.12 10.4480 -16.1326 14.436 -1.216 -14.9  4.7046
```

| | | | | | | |
|---------|----------|----------|---------|---------|--------|-----------|
| M.4.12 | -11.4070 | 55.1551 | -27.040 | 11.825 | 59.1 | -2.0077 |
| M.5.12 | 0.0861 | -95.6143 | 37.192 | -14.876 | -92.7 | 22.2520 |
| M.6.12 | 16.0781 | 92.4013 | -38.015 | -1.399 | 18.5 | -93.9857 |
| M.7.12 | -18.4791 | -46.0499 | 23.988 | 15.680 | 130.5 | 156.3148 |
| M.8.12 | 6.4942 | 11.3213 | -4.385 | -0.994 | -160.4 | -122.2777 |
| M.9.12 | 1.9199 | -3.1913 | -2.660 | -14.820 | 70.3 | 37.7554 |
| M.10.12 | -1.4045 | 1.0592 | 0.708 | 7.487 | -10.2 | -0.0391 |
| | M.7.12 | M.8.12 | M.9.12 | M.10.12 | | |
| M.1.12 | 1.0000 | 1.00 | 1.0 | 1.000 | | |
| M.2.12 | -0.0754 | -2.81 | -13.8 | -0.215 | | |
| M.3.12 | -16.5754 | -2.03 | 72.0 | -16.884 | | |
| M.4.12 | 36.8685 | 17.14 | -99.1 | 39.882 | | |
| M.5.12 | -19.4532 | -22.91 | -29.7 | -20.449 | | |
| M.6.12 | -14.0171 | 2.50 | 46.0 | -32.564 | | |
| M.7.12 | 8.9956 | 8.83 | 254.0 | 42.682 | | |
| M.8.12 | 13.2719 | 20.93 | -427.4 | -1.593 | | |
| M.9.12 | -12.1269 | -39.08 | 258.6 | -23.037 | | |
| M.10.12 | 2.0773 | 16.42 | -61.9 | 11.081 | | |

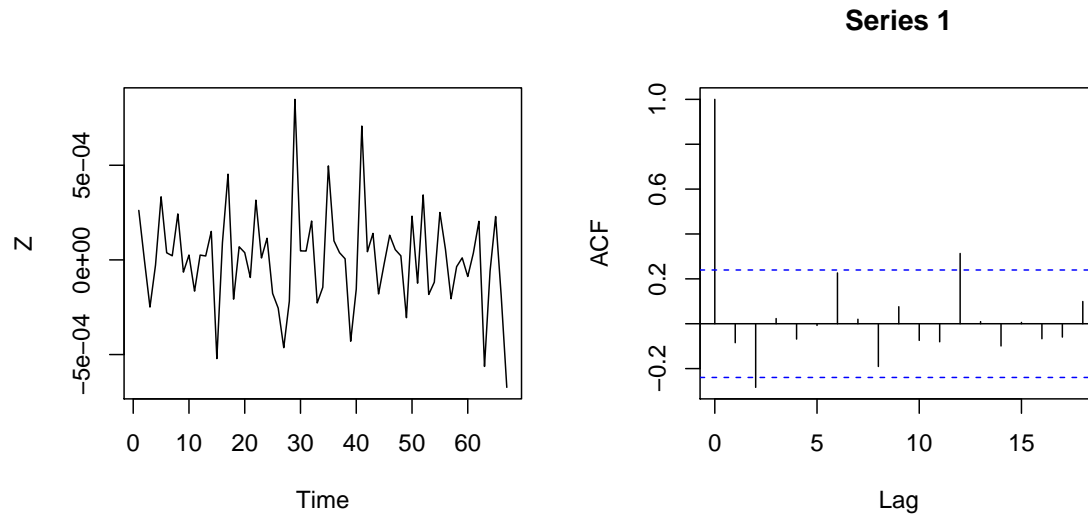
Weights W:

(This is the loading matrix)

| | M.1.12 | M.2.12 | M.3.12 | M.4.12 | M.5.12 | M.6.12 | M.7.12 |
|--------|--------|-----------|---------|--------|---------|--------|--------|
| M.1.d | 11.41 | 0.0231 | 2.93 | -6.142 | 1.1392 | -2.056 | -1.108 |
| M.2.d | 6.09 | 0.3243 | 4.68 | -1.920 | 0.3127 | -1.732 | -1.555 |
| M.3.d | 3.61 | 0.6442 | 5.03 | -0.994 | 0.0122 | -1.526 | -1.644 |
| M.4.d | 3.51 | 0.9286 | 4.65 | -1.621 | -0.0172 | -1.325 | -1.482 |
| M.5.d | 4.02 | 1.0595 | 4.08 | -2.191 | -0.0374 | -1.143 | -1.286 |
| M.6.d | 4.49 | 0.9969 | 3.57 | -2.271 | -0.1221 | -0.997 | -1.151 |
| M.7.d | 4.85 | 0.8585 | 3.15 | -2.108 | -0.2187 | -0.913 | -1.060 |
| M.8.d | 5.01 | 0.7304 | 2.90 | -1.942 | -0.2757 | -0.890 | -0.978 |
| M.9.d | 4.97 | 0.6658 | 2.85 | -1.912 | -0.2788 | -0.935 | -0.883 |
| M.10.d | 4.85 | 0.6689 | 2.89 | -2.035 | -0.2280 | -1.015 | -0.765 |
| | M.8.12 | M.9.12 | M.10.12 | | | | |
| M.1.d | -0.196 | -0.069935 | 0.104 | | | | |
| M.2.d | -0.718 | 0.000369 | 0.152 | | | | |
| M.3.d | -0.819 | 0.039527 | 0.168 | | | | |
| M.4.d | -0.707 | 0.057404 | 0.170 | | | | |
| M.5.d | -0.586 | 0.067680 | 0.178 | | | | |
| M.6.d | -0.517 | 0.077499 | 0.190 | | | | |
| M.7.d | -0.466 | 0.087693 | 0.202 | | | | |
| M.8.d | -0.416 | 0.097029 | 0.212 | | | | |
| M.9.d | -0.363 | 0.104393 | 0.217 | | | | |
| M.10.d | -0.335 | 0.109826 | 0.218 | | | | |

```
> Z = as.matrix(mk.maturity[,2:11]) %*% yields.cajo@V[,1]
> par(mfrow=c(1,2))
> ts.plot(Z)
> acf(Z)
```

The time series and ACF plots of Z are below. Neither plot shows any sign of nonstationarity.



Problem 5. The output is below. If we test that $r = 0$, then the test statistic (8.48) is below the 10% cutoff (12.91), so we accept the null hypothesis that $r = 0$, that is, that there are no cointegration vectors.

If there was a cointegration vector then there would be a portfolio of Coke and Pepsi stock whose price was mean-reverting and this portfolio would be an arbitrage opportunity. Arbitrage opportunities are rare, so it is not surprising that we did not find a cointegrating vector.

```
> summary(ca.jo(CokePepsi))
```

```
#####
# Johansen-Procedure #
#####
```

```
Test type: maximal eigenvalue statistic (lambda max) , with linear trend
```

```
Eigenvalues (lambda):
[1] 0.0057313636 0.0004665101
```

```
Values of teststatistic and critical values of test:
```

```

      test 10pct  5pct  1pct
r <= 1 | 0.69   6.50   8.18 11.65
r = 0  | 8.48  12.91 14.90 19.19
```

```
Eigenvectors, normalised to first column:
(These are the cointegration relations)
```

| | KO.Adjusted.12 | PEP.Adjusted.12 |
|-----------------|----------------|-----------------|
| KO.Adjusted.12 | 1.000000 | 1.00000000 |
| PEP.Adjusted.12 | -1.070219 | -0.04649554 |

Weights W:

(This is the loading matrix)

| | KO.Adjusted.12 | PEP.Adjusted.12 |
|----------------|----------------|-----------------|
| KO.Adjusted.d | 0.003396076 | -0.0010994476 |
| PEP.Adjusted.d | 0.013419081 | -0.0004738957 |