Code **▼**

Unit-Root Nonstatinarity, VAR(p) Model & Cointegration Analysis

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The file m_cofi_4rates.txt contains the monthly rates of the 11th District Cost of Funds Index (COFI), the prime rate of U.S. banks, 1-year and 5-year U.S. Treasury constant maturity rates, and U.S. Treasury 3-month secondary market rates from September 1989 to June 2007. The COFI rates are obtained from the Federal Home Loan Bank of San Francisco, and the other rates are obtained from the Federal Reserve Bank of St. Louis. COFI is a weighted-average interest rate paid by savings institutions headquartered in Arizona, California, and Nevada and is one of the most popular adjustable-rate mortgage (ARM) indices. The prime rate is the interest rate at which banks lend to their most creditworthy customers.

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df <- read.table('m_cofi_4rates.txt', header=TRUE)
this question's code was inspired by https://web.stanford.edu/~xing/statfinbook/_Bo
okFun/ex9.4.5 unitrootCoint.txt</pre>

(a) Perform the augmented Dickey-Fuller test of the unit-root hypothesis for each of these rates.

print('The augmented Dickey-Fuller test of the unit-root hypothesis for the monthly r ates of COFI from September 1989 to June 2007 is:')

[1] "The augmented Dickey-Fuller test of the unit-root hypothesis for the monthly rat es of COFI from September 1989 to June 2007 is:"

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adf.test(df\$cofi)

Augmented Dickey-Fuller Test

data: df\$cofi

Dickey-Fuller = -2.8745, Lag order = 5, p-value = 0.2092

alternative hypothesis: stationary

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print('The augmented Dickey-Fuller test of the unit-root hypothesis for the 1-year U.S. Treasury constant maturity rates from September 1989 to June 2007 is:')

[1] "The augmented Dickey-Fuller test of the unit-root hypothesis for the 1-year U.S. Treasury constant maturity rates from September 1989 to June 2007 is:"

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adf.test(df\$X1ycmt)

Augmented Dickey-Fuller Test

data: df\$X1ycmt

Dickey-Fuller = -2.0479, Lag order = 5, p-value = 0.5559

alternative hypothesis: stationary

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print('The augmented Dickey-Fuller test of the unit-root hypothesis for the 5-year U. S. Treasury constant maturity rates from September 1989 to June 2007 is:')

[1] "The augmented Dickey-Fuller test of the unit-root hypothesis for the 5-year U.S. Treasury constant maturity rates from September 1989 to June 2007 is:"

adf.test(df\$X5ycmt)

Augmented Dickey-Fuller Test

data: df\$X5ycmt

Dickey-Fuller = -2.0488, Lag order = 5, p-value = 0.5555

alternative hypothesis: stationary

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print('The augmented Dickey-Fuller test of the unit-root hypothesis for the prime rat es of U.S. banks from September 1989 to June 2007 is:')

[1] "The augmented Dickey-Fuller test of the unit-root hypothesis for the prime rates of U.S. banks from September 1989 to June 2007 is:"

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adf.test(df\$primeRate)

Augmented Dickey-Fuller Test

data: df\$primeRate

Dickey-Fuller = -2.3052, Lag order = 5, p-value = 0.448

alternative hypothesis: stationary

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print('The augmented Dickey-Fuller test of the unit-root hypothesis for the U.S. Trea sury 3-month secondary market rates from September 1989 to June 2007 is:')

[1] "The augmented Dickey-Fuller test of the unit-root hypothesis for the U.S. Treasury 3-month secondary market rates from September 1989 to June 2007 is:"

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adf.test(df\$X3mTbill.2mkt)

```
Augmented Dickey-Fuller Test

data: df$X3mTbill.2mkt

Dickey-Fuller = -2.5454, Lag order = 5, p-value = 0.3472

alternative hypothesis: stationary
```

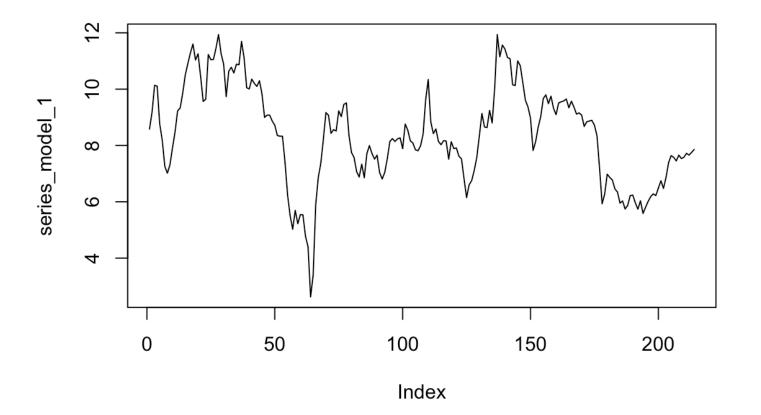
(b) Assuming the VAR(2) model for the multivariate time series of these five rates, perform Johansen's test for the number of cointegration vectors.

```
johansens <- ca.jo(df, type="eigen", K=2, season=NULL)
summary(johansens)</pre>
```

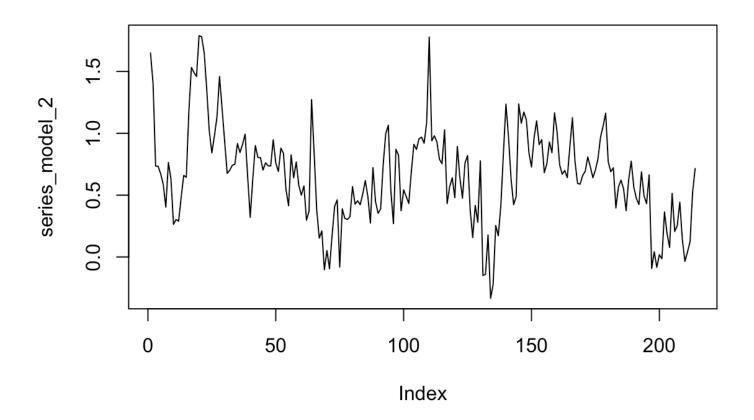
```
########################
# Johansen-Procedure #
###########################
Test type: maximal eigenvalue statistic (lambda max) , with linear trend
Eigenvalues (lambda):
[1] 0.28115617 0.18230111 0.12979406 0.07225424 0.03464075
Values of teststatistic and critical values of test:
          test 10pct 5pct 1pct
         7.47 6.50 8.18 11.65
r <= 3 | 15.90 12.91 14.90 19.19
r <= 2 | 29.47 18.90 21.07 25.75
r <= 1 | 42.67 24.78 27.14 32.14
r = 0 | 69.98 30.84 33.32 38.78
Eigenvectors, normalised to first column:
(These are the cointegration relations)
                  cofi.12 X1ycmt.12 X5ycmt.12 primeRate.12 X3mTbill.2mkt.12
cofi.12
                  1.000000 1.0000000 1.0000000
                                                    1.000000
                                                                    1.0000000
X1ycmt.12
                -5.351875 2.7399460 2.4996108
                                                                    2.9771100
                                                    1.384167
                 1.534273 -0.9231277 -1.5278353
X5ycmt.12
                                                   -1.699396
                                                                   -4.6375957
primeRate.12
                1.908798 0.3260260 0.9516442
                                                                   -1.7744334
                                                  -3.724769
X3mTbill.2mkt.l2 1.443581 -3.2981390 -3.0118510
                                                   2.822877
                                                                   -0.6152458
Weights W:
(This is the loading matrix)
                    cofi.12
                              X1ycmt.12 X5ycmt.12 primeRate.12
cofi.d
                -0.02811300 -0.028599110 0.00204150 -0.005949579
X1ycmt.d
               -0.02835492 0.032656460 -0.02063724 -0.003506453
                -0.02717289 0.007413592 0.03393117 0.017832126
X5ycmt.d
                -0.04463161 0.011492886 -0.04974080 0.009034110
primeRate.d
X3mTbill.2mkt.d -0.04605080 0.116701546 -0.02176690 -0.009198674
               X3mTbill.2mkt.12
cofi.d
                    2.823119e-05
X1ycmt.d
                   8.259722e-03
X5ycmt.d
                    9.537910e-03
                    3.817692e-04
primeRate.d
X3mTbill.2mkt.d
                    4.400697e-03
```

(c) Estimate the cointegration vectors and use them to describe the equilibrium relationship between the five rates.

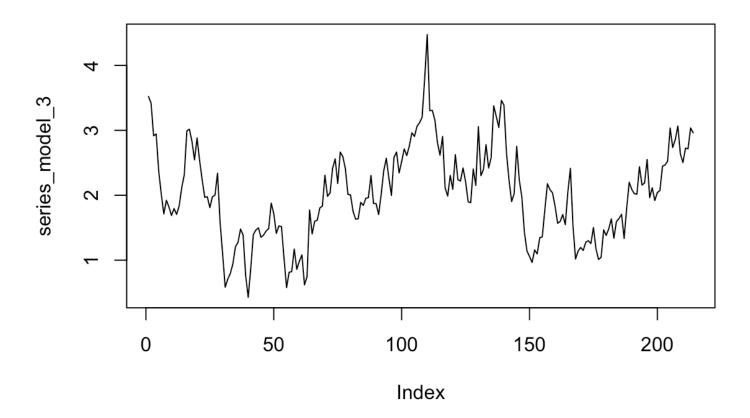
```
coin_vec_1 \leftarrow c(1, -5.351875, 1.534274, 1.908798, 1.443581)
coin_vec_2 <- c(1, 2.739946, -0.9231277, 0.3260260, -3.2981390)
coin vec 3 < -c(1, 2.4996108, -1.5278353, 0.9516442, -3.0118510)
coin vec 4 < -c(1, 1.384167, -1.699396, -3.724769, 2.822877)
coin 1 \leftarrow df$cofi * 1 - 5.351875 * df$X1ycmt + 1.534274 * df$X5ycmt + 1.908798 * df$p
rimeRate + 1.443581 * df$X3mTbill.2mkt
coin_2 \leftarrow df$cofi * 1 + 2.739946 * df$X1ycmt - 0.9231277 * df$X5ycmt + 0.3260260 * df
$primeRate - 3.2981390 * df$X3mTbill.2mkt
coin_3 <- df$cofi * 1 + 2.4996108 * df$X1ycmt - 1.5278353 * df$X5ycmt + 0.9516442 * d
f$primeRate - 3.0118510 * df$X3mTbill.2mkt
coin 4 <- df$cofi * 1 + 1.384167 * df$X1ycmt - 1.699396 * df$X5ycmt - 3.724769 * df$p
rimeRate + 2.822877 * df$X3mTbill.2mkt
series_model_1 <- as.matrix(df) %*% coin_vec 1</pre>
series model 2 <- as.matrix(df) %*% coin vec 2</pre>
series model 3 <- as.matrix(df) %*% coin vec 3
series model 4 <- as.matrix(df) %*% coin vec 4
plot(series_model_1, type = 'l')
```



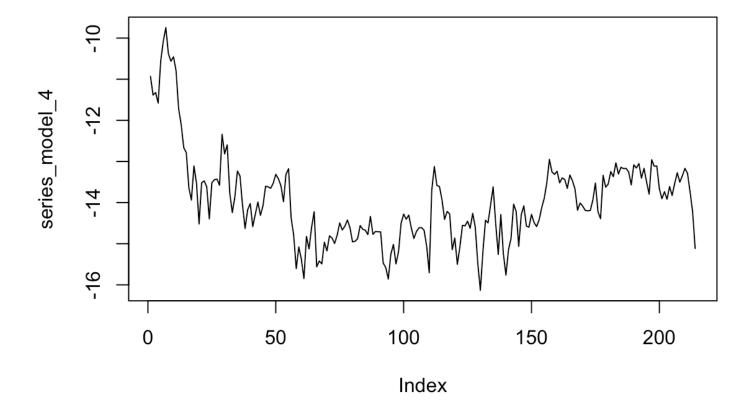
```
plot(series model 2, type = 'l')
```



```
plot(series_model_3, type = '1')
```



```
plot(series_model_4, type = '1')
```



print('Given our estimates of the cointegration vectors and their respective plots, we see that even though the indvidual components are not stationary and have variances that diverge to infinity, they share common trends that result in beta_i^top y_t having long-run equilibrium for i=1,...,4.')

[1] "Given our estimates of the cointegration vectors and their respective plots, we see that even though the indvidual components are not stationary and have variances t hat diverge to infinity, they share common trends that result in beta_i^top y_t havin g long-run equilibrium for $i=1,\ldots,4$."

(d) Regress COFI on the four other rates. Discuss the economic meaning of this regression relationship and whether the regression is spurious.

```
fit <- lm(df$cofi ~ df$X1ycmt + df$X5ycmt + df$primeRate + df$X3mTbill.2mkt)
summary(fit)</pre>
```

```
Call:
lm(formula = df$cofi ~ df$X1ycmt + df$X5ycmt + df$primeRate +
    df$X3mTbill.2mkt)
Residuals:
              10
                   Median
                                3Q
-0.71351 -0.21455 -0.03518 0.16148 1.26509
Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
(Intercept)
                -0.874633
                            0.338637 -2.583
                                               0.0105 *
                            0.116331 -24.905 <2e-16 ***
df$X1ycmt
                -2.897190
df$X5ycmt
                 1.236494
                            0.046559 26.558 <2e-16 ***
                            0.087235 -0.039
df$primeRate
                -0.003372
                                               0.9692
df$X3mTbill.2mkt 2.859960
                            0.140077 20.417 <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.3328 on 209 degrees of freedom
Multiple R-squared: 0.9607,
                               Adjusted R-squared:
F-statistic: 1279 on 4 and 209 DF, p-value: < 2.2e-16
```

print('Given the high R^2 we can conclude that the model explains the variability of the response data around its mean fairly well. The COFI is a regional average of inte rest expenses incurred by financial institutions, which in turn is used as a base for calculating variable rate loans. It makes sense that these rates would vary according to one another so the regression is unlikely to be spurious. Futhermore, from part (c) since \beta_{ji}\neq 0, the linear regression of y_{tj} on the other components of y_t is not spurious even though y_t is unit root stationary.')

[1] "Given the high R^2 we can conclude that the model explains the variability of the response data around its mean fairly well. The COFI is a regional average of interest expenses incurred by financial institutions, which in turn is used as a base for calculating variable rate loans. It makes sense that these rates would vary according to one another so the regression is unlikely to be spurious. Futhermore, from part (c) since β_i on the other components of y_t is not spurious even though y_t is unit root stationary."