Econometria de Séries Temporais*

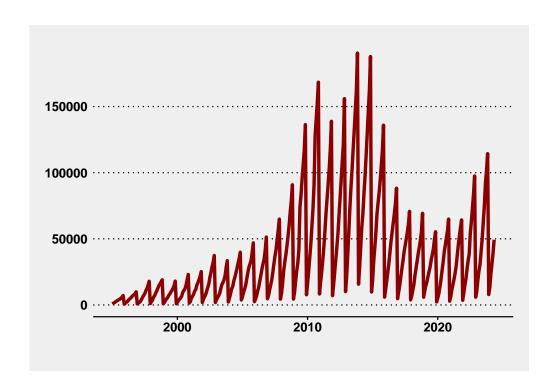
Comentários sobre as resoluções propostas para os exercícios sobre seleção de modelos e diagnóstico de resíduos

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Questão 1

No que tange a definição (ainda que não formal) das hipóteses dos testes, é importante lembrar que elas dizem respeito ao comportamento dos erros (e não dos resíduos).

a)



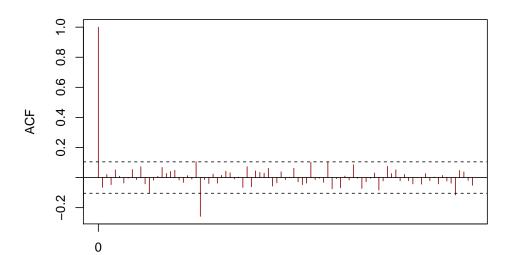
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b)

```
c)
##
## Call:
## arima(x = dbndes, order = c(1, 0, 1))
##
## Coefficients:
                    ma1 intercept
##
           ar1
        0.5054 -0.9538
                            0.0073
##
                            0.0041
## s.e. 0.0499
                0.0136
##
## sigma^2 estimated as 0.5944: log likelihood = -408.61, aic = 825.23
##
## Call:
## arima(x = dbndes, order = c(2, 0, 2))
##
## Coefficients:
##
           ar1
                    ar2
                             ma1
                                     ma2 intercept
        1.4174 -0.6738 -1.8947 0.9438
                                             0.0083
##
## s.e. 0.0414 0.0413 0.0214 0.0207
                                             0.0072
##
## sigma^2 estimated as 0.4931: log likelihood = -377.06, aic = 766.12
##
## Call:
```

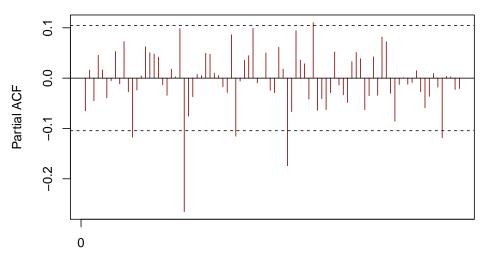
```
## arima(x = dbndes, order = c(2, 0, 2), seasonal = c(1, 1, 0))
##
## Coefficients:
##
             ar1
                     ar2
                             ma1
                                       ma2
                                               sar1
         -0.0332
                  0.8835
                          0.0098
                                  -0.9901
##
                                            -0.3827
          0.0349
                  0.0355
                          0.0197
                                    0.0196
## s.e.
                                             0.0506
## sigma^2 estimated as 0.01325: log likelihood = 249.82, aic = -487.64
##
## Call:
## arima(x = dbndes, order = c(1, 0, 0), seasonal = c(1, 0, 0))
##
## Coefficients:
##
                          intercept
             ar1
                    sar1
         -0.0518
                  0.9874
                             0.0245
##
## s.e.
          0.0532
                 0.0044
                             0.2064
##
## sigma^2 estimated as 0.01664: log likelihood = 199.32, aic = -390.64
  d)
## [1] -0.006771663
  e)
```

Resíduos



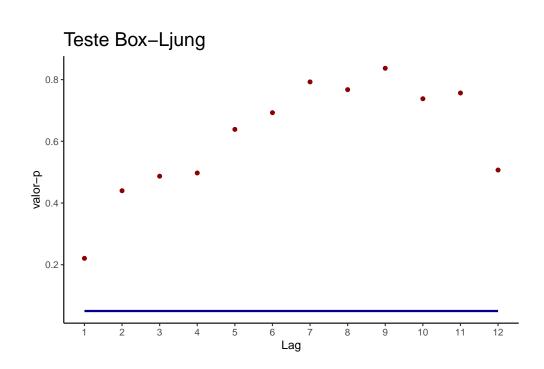
Lag

Resíduos



Lag

f)

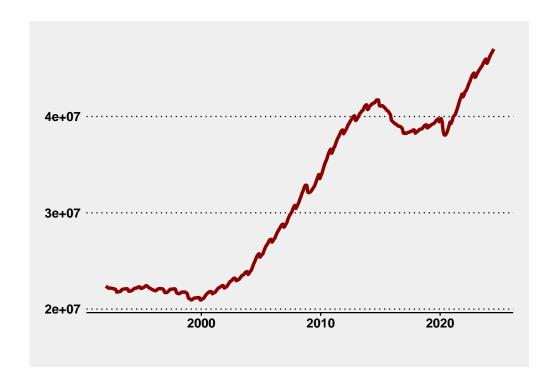


```
g)
##
##
   Jarque Bera Test
##
## data: resids
## X-squared = 330.96, df = 2, p-value < 2.2e-16
 h)
##
## Shapiro-Wilk normality test
##
## data: resids
## W = 0.89589, p-value = 8.826e-15
  i)
##
## ARCH LM-test; Null hypothesis: no ARCH effects
## data: resids
## Chi-squared = 82.666, df = 12, p-value = 1.277e-12
  j)
##
## RESET test
##
## data: resids ~ 0
## RESET = 0, df1 = 2, df2 = 350, p-value = 1
```

Questão 2

No que tange a definição (ainda que não formal) das hipóteses dos testes, é importante lembrar que elas dizem respeito ao comportamento dos erros (e não dos resíduos).

a)



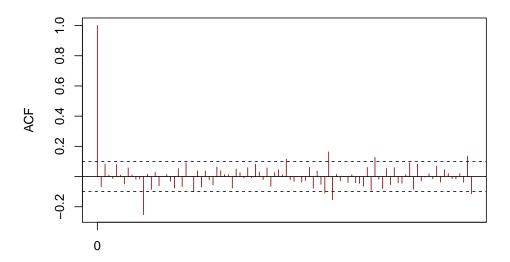
b)

```
0.00 -0.01 -0.02 -0.02 -0.02
```

```
c)
##
## Call:
## arima(x = dcaged, order = c(1, 0, 1))
##
## Coefficients:
                         intercept
##
            ar1
                     ma1
                             0.0019
##
         0.5442 - 0.1509
                             0.0005
## s.e. 0.1067
                0.1278
##
## sigma^2 estimated as 2.585e-05: log likelihood = 1506.36, aic = -3004.73
##
## Call:
## arima(x = dcaged, order = c(2, 0, 2))
##
## Coefficients:
##
            ar1
                    ar2
                            ma1
                                    ma2 intercept
         0.0613 0.3358 0.3384 -0.1771
                                             0.0019
##
## s.e. 0.3252 0.1547 0.3285
                                             0.0005
                                 0.1089
##
## sigma^2 estimated as 2.577e-05: log likelihood = 1506.97, aic = -3001.94
##
## Call:
```

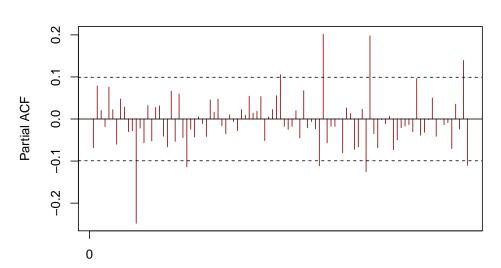
```
## arima(x = dcaged, order = c(2, 0, 2), seasonal = c(1, 1, 0))
##
## Coefficients:
##
           ar1
                   ar2
                           ma1
                                   ma2
                                           sar1
        0.5566 0.1225 0.0304 0.0200 -0.3630
##
## s.e. 0.6161 0.4652 0.6150 0.1292
                                         0.0472
## sigma^2 estimated as 7.22e-06: log likelihood = 1700.01, aic = -3388.03
##
## Call:
## arima(x = dcaged, order = c(1, 0, 0), seasonal = c(1, 0, 0))
##
## Coefficients:
                  sar1 intercept
##
           ar1
##
        0.6363 0.8407
                           0.0015
## s.e. 0.0393 0.0257
                           0.0021
##
## sigma^2 estimated as 7.788e-06: log likelihood = 1732.76, aic = -3457.52
##
## Call:
## arima(x = dcaged, order = c(1, 0, 1), seasonal = c(0, 1, 2))
##
## Coefficients:
##
           ar1
                    ma1
                             sma1
                                     sma2
        0.7902 -0.1742 -0.5936 -0.1360
## s.e. 0.0449
                0.0714
                          0.0514
                                   0.0517
##
## sigma^2 estimated as 6.102e-06: log likelihood = 1728.55, aic = -3447.11
 d)
## [1] 6.448369e-05
  e)
```

Resíduos



Lag

Resíduos



Lag

f)

```
Teste Box–Ljung

0.4

0.2

1 2 3 4 5 6 7 8 9 10 11 12
```

```
g)
##
    Jarque Bera Test
##
## data: resids
## X-squared = 4927.2, df = 2, p-value < 2.2e-16
  h)
##
   Shapiro-Wilk normality test
##
##
## data: resids
## W = 0.86065, p-value < 2.2e-16
  i)
##
   ARCH LM-test; Null hypothesis: no ARCH effects
##
## data: resids
## Chi-squared = 60.476, df = 12, p-value = 1.848e-08
  j)
##
##
   RESET test
##
```

```
## data: resids ~ 0
## RESET = 0, df1 = 2, df2 = 388, p-value = 1
```