Appendix R Code

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```
library(astsa)
library(MASS)
library(forecast)

## Registered S3 method overwritten by 'quantmod':

## method from

## as.zoo.data.frame zoo

##

## Attaching package: 'forecast'

## The following object is masked from 'package:astsa':

##

## gas

library(ggplot2)
load("finalproject.Rdata")
```

Exploratory Data Analysis

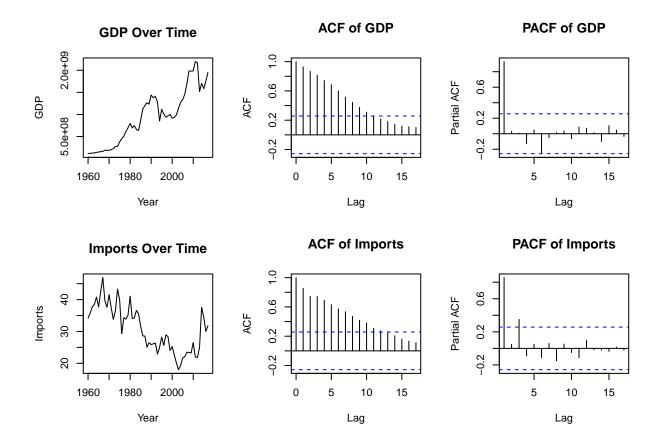
Raw Data

```
# Exploratory Data Analysis
# Raw Data with Year-Based Time Series

# Set up 2x3 plotting area
par(mfrow = c(2,3))

# 1. GDP
gdp_ts <- ts(finalPro_data$GDP, start = 1960, frequency = 1)
plot(gdp_ts, main = "GDP Over Time", ylab = "GDP", xlab = "Year")
acf(gdp_ts, main = "ACF of GDP")
pacf(gdp_ts, main = "PACF of GDP")

# 2. Imports
import_ts <- ts(finalPro_data$Imports, start = 1960, frequency = 1)
plot(import_ts, main = "Imports Over Time", ylab = "Imports", xlab = "Year")
acf(import_ts, main = "ACF of Imports")
pacf(import_ts, main = "PACF of Imports")</pre>
```



Log-Transformed Data

```
# Log-Transformed Data
log.gdp.ts <- ts(log(finalPro_data$GDP), start = 1960, frequency = 1)
log.import.ts <- ts(log(finalPro_data$Imports), start = 1960, frequency = 1)

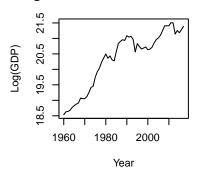
# Set up 2x3 plotting area
par(mfrow = c(2,3))

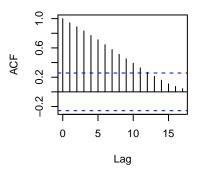
# 1. GDP
plot(log.gdp.ts, main = "Log-Transformed GDP Over Time", ylab = "Log(GDP)", xlab = "Year")
acf(log.gdp.ts, main = "ACF of Log-Transformed GDP")
pacf(log.gdp.ts, main = "PACF of Log-Transformed GDP")

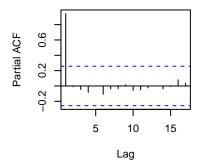
# 2. Imports
plot(log.import.ts, main = "Log-Transformed Imports Over Time", ylab = "Log(Imports)", xlab = "Year")
acf(log.import.ts, main = "ACF of Log-Transformed Imports")
pacf(log.import.ts, main = "PACF of Log-Transformed Imports")</pre>
```

Log-Transformed GDP Over Tir

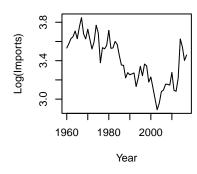
ACF of Log-Transformed GDF PACF of Log-Transformed GD

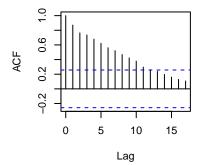


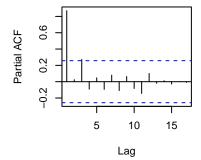




Log-Transformed Imports Over 1 ACF of Log-Transformed Impo PACF of Log-Transformed Impo



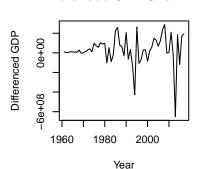




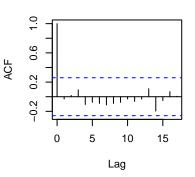
Differencing Data

```
# Differenced Data (first difference removes first observation, so start at 1961)
diff.gdp.ts <- ts(diff(finalPro_data$GDP), start = 1961, frequency = 1)</pre>
diff.import.ts <- ts(diff(finalPro_data$Imports), start = 1961, frequency = 1)</pre>
# Set up 2x3 plotting area
par(mfrow = c(2,3))
# 1. GDP
plot(diff.gdp.ts, main = "Differenced GDP Over Time", ylab = "Differenced GDP", xlab = "Year")
acf(diff.gdp.ts, main = "ACF of Differenced GDP")
pacf(diff.gdp.ts, main = "PACF of Differenced GDP")
# 2. Imports
plot(diff.import.ts, main = "Differenced Imports Over Time", ylab = "Differenced Imports", xlab = "Year
acf(diff.import.ts, main = "ACF of Differenced Imports")
pacf(diff.import.ts, main = "PACF of Differenced Imports")
```

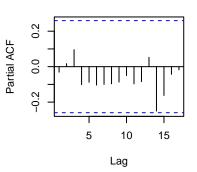
Differenced GDP Over Time



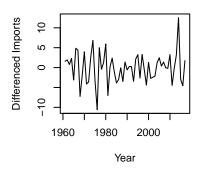
ACF of Differenced GDP



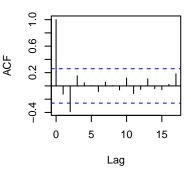
PACF of Differenced GDP



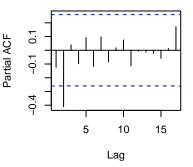
Differenced Imports Over Time



ACF of Differenced Imports



PACF of Differenced Imports



Box-cox Transformed Data

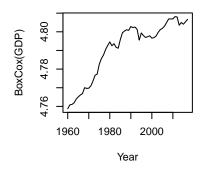
```
# Box-Cox Transformed Data
gdp.lambda <- BoxCox.lambda(finalPro_data$GDP)
boxcox.gdp.ts <- ts(BoxCox(finalPro_data$GDP, gdp.lambda), start = 1960, frequency = 1)
import.lambda <- BoxCox.lambda(finalPro_data$Imports)
boxcox.import.ts <- ts(BoxCox(finalPro_data$Imports, import.lambda), start = 1960, frequency = 1)

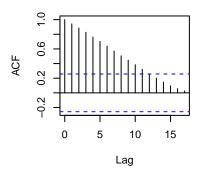
# Set up 2x3 plotting area
par(mfrow = c(2,3))

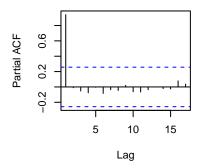
# 1. GDP
plot(boxcox.gdp.ts, main = "Box-Cox Transformed GDP Over Time", ylab = "BoxCox(GDP)", xlab = "Year")
acf(boxcox.gdp.ts, main = "ACF of Box-Cox Transformed GDP")
pacf(boxcox.gdp.ts, main = "PACF of Box-Cox Transformed GDP")

# 2. Imports
plot(boxcox.import.ts, main = "Box-Cox Transformed Imports Over Time", ylab = "BoxCox(Imports)", xlab = acf(boxcox.import.ts, main = "ACF of Box-Cox Transformed Imports")
pacf(boxcox.import.ts, main = "PACF of Box-Cox Transformed Imports")</pre>
```

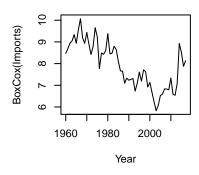
3ox-Cox Transformed GDP Over ACF of Box-Cox Transformed G PACF of Box-Cox Transformed C

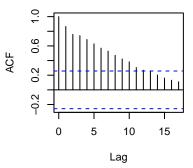


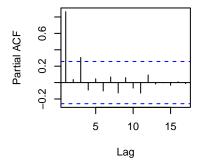




x-Cox Transformed Imports Ove^{ACF} of Box-Cox Transformed ImpACF of Box-Cox Transformed Im







Box-cox Differencing Data

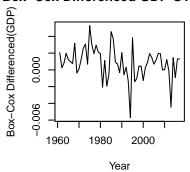
```
# Box-Cox Differenced Data (first difference removes first year, so start = 1961)
boxcox.diff.gdp.ts <- ts(diff(boxcox.gdp.ts), start = 1961, frequency = 1)
boxcox.diff.import.ts <- ts(diff(boxcox.import.ts), start = 1961, frequency = 1)

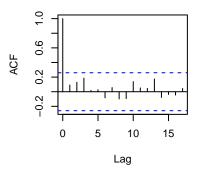
# Set up 2x3 plotting area
par(mfrow = c(2,3))

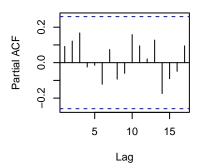
# 1. GDP
plot(boxcox.diff.gdp.ts, main = "Box-Cox Differenced GDP Over Time", ylab = "Box-Cox Differenced(GDP)",
acf(boxcox.diff.gdp.ts, main = "ACF of Box-Cox Differenced GDP")
pacf(boxcox.diff.gdp.ts, main = "PACF of Box-Cox Differenced GDP")

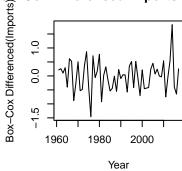
# 2. Imports
plot(boxcox.diff.import.ts, main = "Box-Cox Differenced Imports Over Time", ylab = "Box-Cox Differenced acf(boxcox.diff.import.ts, main = "ACF of Box-Cox Differenced Imports")
pacf(boxcox.diff.import.ts, main = "PACF of Box-Cox Differenced Imports")</pre>
```

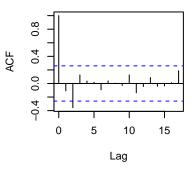
Box-Cox Differenced GDP Over 1 ACF of Box-Cox Differenced GI PACF of Box-Cox Differenced G

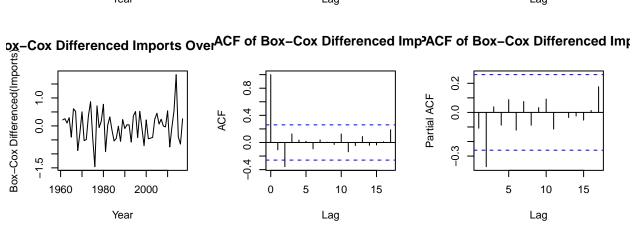












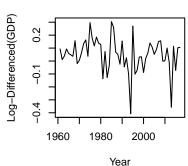
Log-Differencing Data (Final)

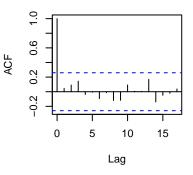
```
# Log-Differenced Data (Final)
log.diff.gdp.ts <- ts(diff(log.gdp.ts), start = 1961, frequency = 1)</pre>
log.diff.import.ts <- ts(diff(log.import.ts), start = 1961, frequency = 1)</pre>
# Set up 2x3 plotting area
par(mfrow = c(2,3))
# 1. GDP
plot(log.diff.gdp.ts, main = "Log-Differenced GDP Over Time", ylab = "Log-Differenced(GDP)", xlab = "Ye
acf(log.diff.gdp.ts, main = "ACF of Log-Differenced GDP")
pacf(log.diff.gdp.ts, main = "PACF of Log-Differenced GDP")
# 2. Imports
plot(log.diff.import.ts, main = "Log-Differenced Imports Over Time", ylab = "Log-Differenced(Imports)",
acf(log.diff.import.ts, main = "ACF of Log-Differenced Imports")
pacf(log.diff.import.ts, main = "PACF of Log-Differenced Imports")
```

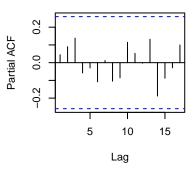
Log-Differenced GDP Over Tin

ACF of Log-Differenced GDP

PACF of Log-Differenced GDF

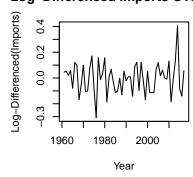


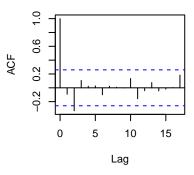


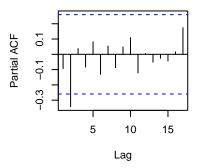


Log-Differenced Imports Over T

ACF of Log-Differenced Impor PACF of Log-Differenced Impor







Model Selection

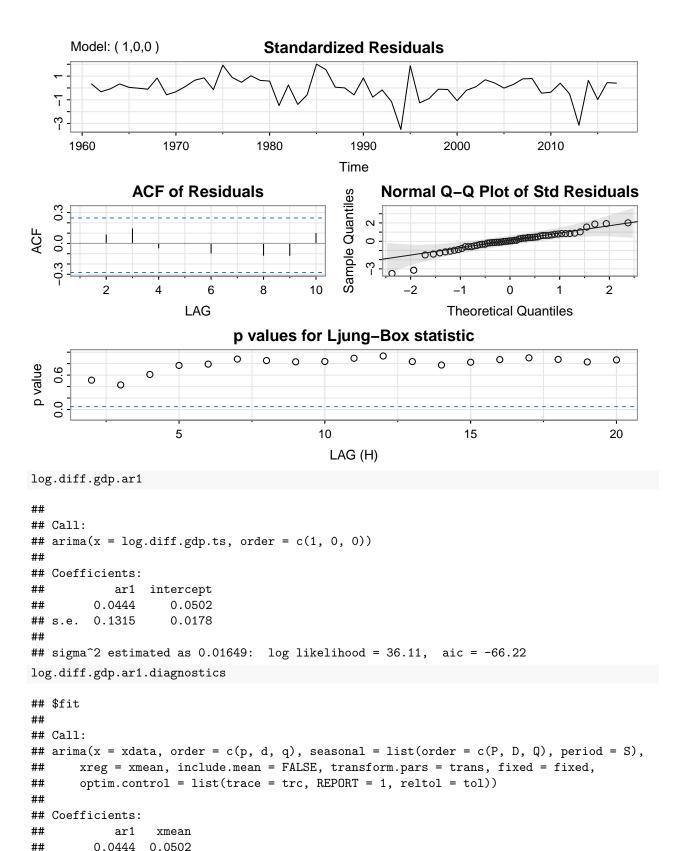
GDP

```
# Model Selection on Log-Differenced GDP (stationary)
# GDP
# Model 1: MA(1)
log.diff.gdp.ma1 \leftarrow arima(log.diff.gdp.ts, order = c(0,0,1))
log.diff.gdp.ma1.diagnostics <- sarima(log.diff.gdp.ts, 0, 0, 1)</pre>
## initial value -2.051458
          2 value -2.052302
## iter
          3 value -2.052325
## iter
          4 value -2.052325
## iter
   iter
          4 value -2.052325
## iter
          4 value -2.052325
## final value -2.052325
## converged
## initial value -2.052314
          2 value -2.052314
## iter
## iter
          2 value -2.052314
## iter
          2 value -2.052314
## final value -2.052314
## converged
## <><><><>
##
```

```
## Coefficients:
##
         Estimate
                      SE t.value p.value
           0.0382 0.1228 0.3113 0.7568
## ma1
           0.0502 0.0177 2.8416 0.0063
##
   xmean
##
## sigma^2 estimated as 0.01649574 on 55 degrees of freedom
## AIC = -1.161487 AICc = -1.157589 BIC = -1.053958
##
      Model: (0,0,1)
                                  Standardized Residuals
     1960
                   1970
                                 1980
                                               1990
                                                             2000
                                                                           2010
                                              Time
               ACF of Residuals
                                                    Normal Q-Q Plot of Std Residuals
                                              Sample Quantiles
                                                                            000
                                                 α-
                                                 0
                                                        0
                                                       -2
                          6
                                  8
                                         10
                                                                                     2
                                                              Theoretical Quantiles
                       LAG
                              p values for Ljung-Box statistic
p value
   9.0
   0.0
                     5
                                           10
                                                                15
                                                                                      20
                                            LAG (H)
log.diff.gdp.ma1
##
## arima(x = log.diff.gdp.ts, order = c(0, 0, 1))
##
## Coefficients:
##
            ma1
                 intercept
##
         0.0382
                    0.0502
## s.e. 0.1228
                    0.0177
## sigma^2 estimated as 0.0165: log likelihood = 36.1, aic = -66.2
log.diff.gdp.ma1.diagnostics
## $fit
##
```

Call:

```
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
##
      xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
      optim.control = list(trace = trc, REPORT = 1, reltol = tol))
##
## Coefficients:
##
           ma1
                 xmean
        0.0382 0.0502
## s.e. 0.1228 0.0177
##
## sigma^2 estimated as 0.0165: log likelihood = 36.1, aic = -66.2
## $degrees_of_freedom
## [1] 55
##
## $ttable
##
        Estimate
                     SE t.value p.value
## ma1
          0.0382 0.1228 0.3113 0.7568
## xmean
          0.0502 0.0177 2.8416 0.0063
##
## $ICs
##
        AIC
                 AICc
                            RTC
## -1.161487 -1.157589 -1.053958
# Model 2: AR(1)
log.diff.gdp.ar1 \leftarrow arima(log.diff.gdp.ts, order = c(1,0,0))
log.diff.gdp.ar1.diagnostics <- sarima(log.diff.gdp.ts, 1, 0, 0)</pre>
## initial value -2.043604
## iter 2 value -2.044628
## iter
        3 value -2.044635
## iter 4 value -2.044638
## iter 4 value -2.044638
## iter 4 value -2.044638
## final value -2.044638
## converged
## initial value -2.052438
        2 value -2.052450
## iter
         3 value -2.052458
## iter
## iter
       3 value -2.052458
       3 value -2.052458
## iter
## final value -2.052458
## converged
## <><><><>
##
## Coefficients:
##
        Estimate
                     SE t.value p.value
          0.0444 0.1315 0.3380 0.7367
## xmean 0.0502 0.0178 2.8207 0.0067
## sigma^2 estimated as 0.01649082 on 55 degrees of freedom
## AIC = -1.161776 AICc = -1.157878 BIC = -1.054247
##
```

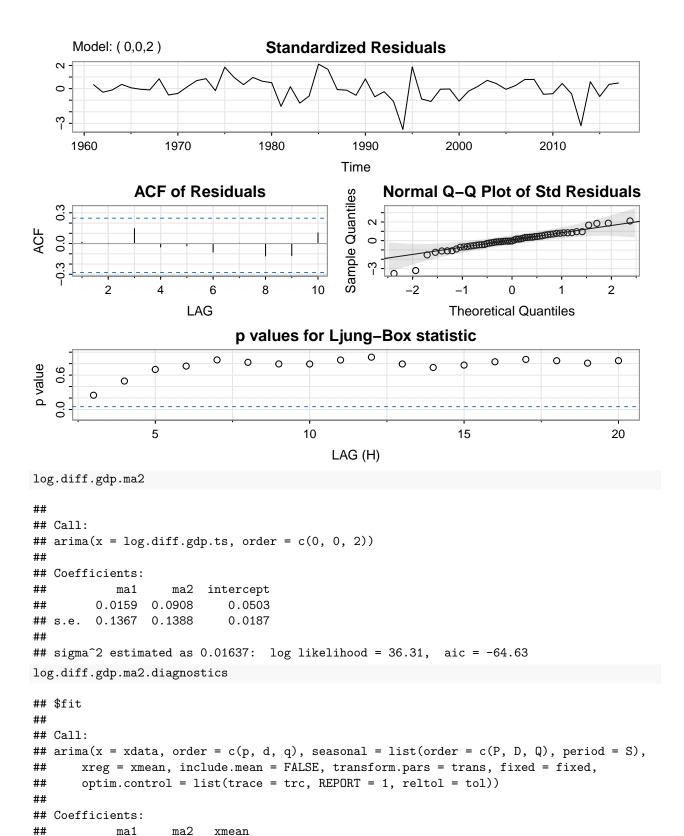


$sigma^2$ estimated as 0.01649: log likelihood = 36.11, aic = -66.22

##

s.e. 0.1315 0.0178

```
##
## $degrees_of_freedom
## [1] 55
##
## $ttable
##
                     SE t.value p.value
        Estimate
        0.0444 0.1315 0.3380 0.7367
## xmean 0.0502 0.0178 2.8207 0.0067
##
## $ICs
##
        AIC
                 AICc
                            BIC
## -1.161776 -1.157878 -1.054247
# Model 3: MA(2)
log.diff.gdp.ma2 \leftarrow arima(log.diff.gdp.ts, order = c(0,0,2))
log.diff.gdp.ma2.diagnostics <- sarima(log.diff.gdp.ts, 0, 0, 2)</pre>
## initial value -2.051458
## iter 2 value -2.055718
## iter 3 value -2.056120
## iter 4 value -2.056142
## iter 5 value -2.056143
## iter 5 value -2.056143
## iter 5 value -2.056143
## final value -2.056143
## converged
## initial value -2.056002
## iter 2 value -2.056006
## iter 3 value -2.056006
## iter 3 value -2.056006
## iter 3 value -2.056006
## final value -2.056006
## converged
## <><><><>
##
## Coefficients:
##
        Estimate
                     SE t.value p.value
          0.0159 0.1367 0.1162 0.9079
## ma1
## ma2
          0.0908 0.1388 0.6540 0.5159
## xmean 0.0503 0.0187 2.6901 0.0095
## sigma^2 estimated as 0.01636998 on 54 degrees of freedom
## AIC = -1.133784 AICc = -1.12584 BIC = -0.9904121
##
```



0.0159

s.e. 0.1367 0.1388

##

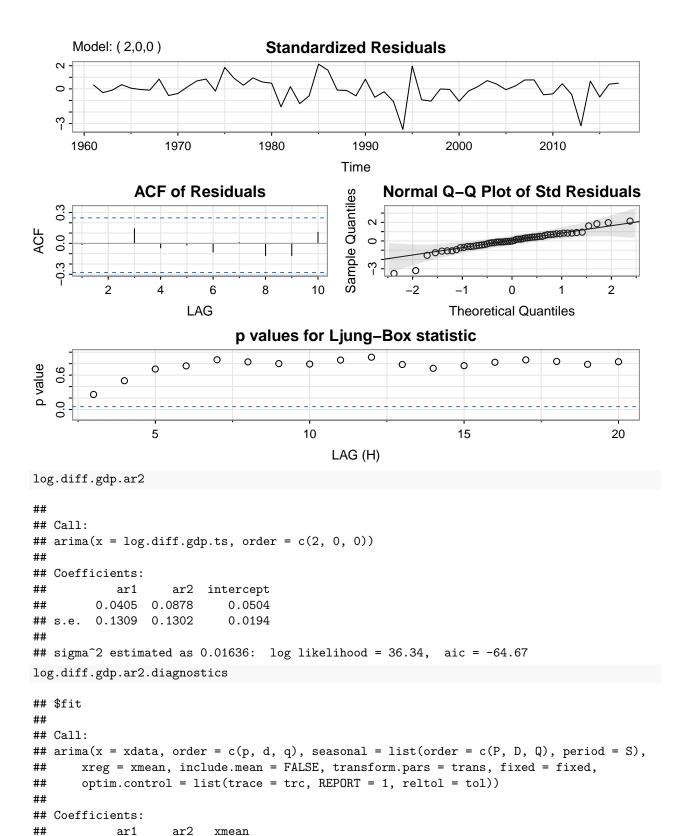
0.0908

0.0503

0.0187

$sigma^2$ estimated as 0.01637: log likelihood = 36.31, aic = -64.63

```
##
## $degrees_of_freedom
## [1] 54
##
## $ttable
##
                     SE t.value p.value
        Estimate
## ma1
        0.0159 0.1367 0.1162 0.9079
          0.0908 0.1388 0.6540 0.5159
## ma2
## xmean 0.0503 0.0187 2.6901 0.0095
##
## $ICs
##
         AIC
                   AICc
## -1.1337841 -1.1258397 -0.9904121
# Model 4: AR(2)
log.diff.gdp.ar2 \leftarrow arima(log.diff.gdp.ts, order = c(2,0,0))
log.diff.gdp.ar2.diagnostics <- sarima(log.diff.gdp.ts, 2, 0, 0)</pre>
## initial value -2.035413
## iter 2 value -2.040611
## iter 3 value -2.040622
## iter 4 value -2.040622
## iter 4 value -2.040622
## iter 4 value -2.040622
## final value -2.040622
## converged
## initial value -2.056414
## iter 2 value -2.056422
## iter 3 value -2.056422
## iter 3 value -2.056422
## iter 3 value -2.056422
## final value -2.056422
## converged
## <><><><><>
##
## Coefficients:
##
        Estimate
                     SE t.value p.value
## ar1
          0.0405 0.1309 0.3097 0.7580
## ar2
          0.0878 0.1302 0.6738 0.5033
## xmean 0.0504 0.0194 2.6026 0.0119
\#\# sigma^2 estimated as 0.01635617 on 54 degrees of freedom
## AIC = -1.134616 AICc = -1.126672 BIC = -0.991244
##
```



0.0405

s.e. 0.1309 0.1302 0.0194

##

0.0878

0.0504

$sigma^2$ estimated as 0.01636: log likelihood = 36.34, aic = -64.67

```
##
## $degrees_of_freedom
## [1] 54
##
## $ttable
##
                     SE t.value p.value
        Estimate
## ar1
          0.0405 0.1309 0.3097 0.7580
          0.0878 0.1302 0.6738 0.5033
## ar2
## xmean
          0.0504 0.0194 2.6026 0.0119
##
## $ICs
##
        AIC
                 AICc
                            BIC
## -1.134616 -1.126672 -0.991244
# Model 5: ARMA(1,1)
log.diff.gdp.arma11 \leftarrow arima(log.diff.gdp.ts, order = c(1,0,1))
log.diff.gdp.arma11.diagnostics <- sarima(log.diff.gdp.ts, 1, 0, 1)</pre>
## initial value -2.043604
        2 value -2.044276
## iter
## iter
       3 value -2.044596
## iter
       4 value -2.044598
## iter
        5 value -2.044625
## iter
       6 value -2.044786
       7 value -2.044823
## iter
## iter
       8 value -2.044926
## iter
        9 value -2.044934
## iter 10 value -2.044947
## iter 11 value -2.045576
## iter 12 value -2.046165
## iter 13 value -2.046254
## iter 14 value -2.047052
## iter 15 value -2.047177
## iter 16 value -2.047178
## iter 16 value -2.047178
## iter 16 value -2.047178
## final value -2.047178
## converged
## initial value -2.056080
## iter
        2 value -2.056116
       3 value -2.056124
## iter
## iter
       4 value -2.056126
## iter 5 value -2.056128
## iter 6 value -2.056139
## iter
        7 value -2.056145
## iter
         8 value -2.056149
## iter
        9 value -2.056149
## iter 10 value -2.056149
## iter 10 value -2.056149
## iter 10 value -2.056149
## final value -2.056149
## converged
## <><><><>
##
## Coefficients:
```

```
##
         Estimate
                       SE t.value p.value
## ar1
           0.6353 0.4481 1.4177 0.1620
          -0.5615 0.4727 -1.1879 0.2401
           0.0502 0.0202 2.4854 0.0161
##
   xmean
## sigma^2 estimated as 0.0163663 on 54 degrees of freedom
## AIC = -1.134071 AICc = -1.126126 BIC = -0.9906985
                                   Standardized Residuals
      Model: (1,0,1)
   ^{\circ}
     1960
                    1970
                                  1980
                                                1990
                                                              2000
                                                                            2010
                                              Time
               ACF of Residuals
                                                    Normal Q-Q Plot of Std Residuals
                                              Sample Quantiles
                                                                             000
                                                  α-
                                                  0
                                                        0
                                                        -2
                          6
                                  8
                                          10
                                                                       0
                                                                                      2
                                                              Theoretical Quantiles
                       LAG
                              p values for Ljung-Box statistic
p value
                                                                                       0
   9.0
                  0
                 5
                                        10
                                                               15
                                                                                      20
                                             LAG (H)
log.diff.gdp.arma11
##
## arima(x = log.diff.gdp.ts, order = c(1, 0, 1))
##
## Coefficients:
##
            ar1
                           intercept
                      ma1
##
         0.6353
                 -0.5615
                              0.0502
                   0.4727
                              0.0202
## s.e. 0.4481
## sigma^2 estimated as 0.01637: log likelihood = 36.32, aic = -64.64
log.diff.gdp.arma11.diagnostics
## $fit
##
```

Call:

```
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
##
       xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
       optim.control = list(trace = trc, REPORT = 1, reltol = tol))
##
## Coefficients:
##
                           xmean
            ar1
                     ma1
         0.6353 -0.5615 0.0502
## s.e. 0.4481
                  0.4727 0.0202
##
## sigma^2 estimated as 0.01637: log likelihood = 36.32, aic = -64.64
## $degrees_of_freedom
## [1] 54
##
## $ttable
##
         Estimate
                      SE t.value p.value
## ar1
           0.6353 0.4481 1.4177 0.1620
## ma1
          -0.5615 0.4727 -1.1879 0.2401
## xmean 0.0502 0.0202 2.4854 0.0161
##
## $ICs
##
          AIC
                    AICc
                                BIC
## -1.1340705 -1.1261261 -0.9906985
# Model 6: ARMA(1,2)
log.diff.gdp.arma12 \leftarrow arima(log.diff.gdp.ts, order = c(1,0,2))
log.diff.gdp.arma12.diagnostics <- sarima(log.diff.gdp.ts, 1, 0, 2)
## initial value -2.043604
## iter
        2 value -2.046177
## iter
        3 value -2.048533
## iter
        4 value -2.048624
        5 value -2.048815
## iter
## iter
         6 value -2.049700
## iter
         7 value -2.051473
## iter
         8 value -2.051869
         9 value -2.052374
## iter
## iter 10 value -2.052542
## iter 11 value -2.052542
## iter 11 value -2.052543
## iter 11 value -2.052543
## final value -2.052543
## converged
## initial value -2.061088
## iter
         2 value -2.061098
## iter
         3 value -2.061118
## iter
        4 value -2.061120
## iter
        5 value -2.061122
## iter
         6 value -2.061133
         7 value -2.061138
## iter
## iter
          8 value -2.061139
         9 value -2.061139
## iter
## iter
          9 value -2.061139
## iter
          9 value -2.061139
## final value -2.061139
```

```
## converged
## <><><><>
##
##
  Coefficients:
##
         Estimate
                      SE t.value p.value
## ar1
           0.4806 0.4350 1.1048 0.2742
          -0.4623 0.4343 -1.0644 0.2920
  ma1
           0.1225 0.1510 0.8116 0.4206
## ma2
##
  xmean
           0.0505 0.0212 2.3799 0.0209
##
  sigma^2 estimated as 0.01619399 on 53 degrees of freedom
## AIC = -1.108962 AICc = -1.095467 BIC = -0.9297474
##
      Model: (1,0,2)
                                  Standardized Residuals
   ^{\circ}
     1960
                                 1980
                                              1990
                   1970
                                                            2000
                                                                          2010
                                             Time
              ACF of Residuals
                                                   Normal Q-Q Plot of Std Residuals
                                             Sample Quantiles
                                                 ი -
                                                 0 -
                  4
                         6
                                 8
                                         10
                                                      -2
                                                                     0
                                                                                    2
                      LAG
                                                             Theoretical Quantiles
                             p values for Ljung-Box statistic
                                                                                     0
   9.0
             5
                                    10
                                                                                    20
                                                             15
                                           LAG (H)
log.diff.gdp.arma12
##
## Call:
## arima(x = log.diff.gdp.ts, order = c(1, 0, 2))
## Coefficients:
##
            ar1
                     ma1
                             ma2
                                  intercept
##
         0.4806
                -0.4623
                          0.1225
                                     0.0505
                                     0.0212
## s.e. 0.4350
                  0.4343
                          0.1510
## sigma^2 estimated as 0.01619: log likelihood = 36.61, aic = -63.21
```

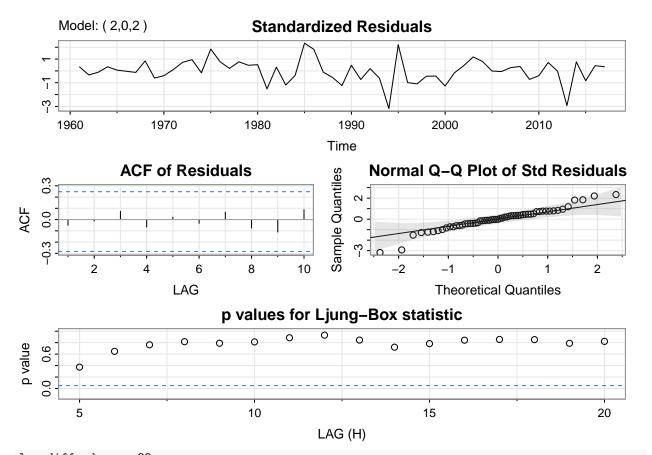
```
log.diff.gdp.arma12.diagnostics
## $fit
##
## Call:
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
       xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
       optim.control = list(trace = trc, REPORT = 1, reltol = tol))
##
## Coefficients:
##
                            ma2
            ar1
                     ma1
                                  xmean
##
        0.4806 -0.4623 0.1225 0.0505
## s.e. 0.4350
                 0.4343 0.1510 0.0212
##
## sigma^2 estimated as 0.01619: log likelihood = 36.61, aic = -63.21
## $degrees_of_freedom
## [1] 53
##
## $ttable
##
                     SE t.value p.value
        Estimate
## ar1
          0.4806 0.4350 1.1048 0.2742
## ma1
         -0.4623 0.4343 -1.0644 0.2920
## ma2
          0.1225 0.1510 0.8116 0.4206
          0.0505 0.0212 2.3799 0.0209
## xmean
##
## $ICs
##
         AIC
                    AICc
                                BIC
## -1.1089624 -1.0954671 -0.9297474
# Model 7: ARMA(2,1)
log.diff.gdp.arma21 \leftarrow arima(log.diff.gdp.ts, order = c(2,0,1))
log.diff.gdp.arma21.diagnostics <- sarima(log.diff.gdp.ts, 2, 0, 1)</pre>
## initial value -2.035413
## iter 2 value -2.038759
## iter
        3 value -2.040329
## iter
        4 value -2.040504
## iter
       5 value -2.040649
## iter
        6 value -2.041805
## iter
       7 value -2.042863
## iter
        8 value -2.043277
        9 value -2.043499
## iter
## iter 10 value -2.043624
## iter 11 value -2.043626
## iter 12 value -2.043627
## iter 13 value -2.043627
## iter 14 value -2.043629
## iter 15 value -2.043629
## iter 16 value -2.043630
## iter 16 value -2.043630
## iter 16 value -2.043630
## final value -2.043630
## converged
```

```
## initial value -2.059446
## iter
          2 value -2.059459
          3 value -2.059466
          4 value -2.059467
## iter
  iter
          5 value -2.059470
          6 value -2.059475
##
  iter
## iter
          7 value -2.059479
          8 value -2.059480
## iter
## iter
          9 value -2.059480
          9 value -2.059480
## iter
## iter
          9 value -2.059480
## final value -2.059480
   converged
   <><><><><>
##
##
##
   Coefficients:
##
         Estimate
                      SE t.value p.value
           0.4057 0.5161 0.7862 0.4353
##
   ar1
##
           0.0918 0.1378 0.6661
                                  0.5082
  ar2
          -0.3718 0.5064 -0.7343
##
                                  0.4660
##
   xmean
           0.0504 0.0209 2.4092 0.0195
##
## sigma^2 estimated as 0.01625255 on 53 degrees of freedom
## AIC = -1.105644 AICc = -1.092148 BIC = -0.9264285
##
      Model: (2,0,1)
                                  Standardized Residuals
   \alpha
   0
     1960
                   1970
                                 1980
                                               1990
                                                             2000
                                                                           2010
                                             Time
               ACF of Residuals
                                                    Normal Q-Q Plot of Std Residuals
                                              Sample Quantiles
   0.3
                                                                           000
                                                 ი -
                                                 0 -
          2
                  4
                          6
                                  8
                                         10
                                                       -2
                                                              -1
                                                                      0
                                                                                     2
                       LAG
                                                             Theoretical Quantiles
                              p values for Ljung-Box statistic
p value
                  0
             5
                                     10
                                                             15
                                                                                     20
                                            LAG (H)
```

```
\log.diff.gdp.arma21
##
## Call:
## arima(x = log.diff.gdp.ts, order = c(2, 0, 1))
## Coefficients:
##
                   ar2
                                 intercept
            ar1
                            ma1
                                     0.0504
##
         0.4057 0.0918 -0.3718
                                     0.0209
## s.e. 0.5161 0.1378
                        0.5064
## sigma^2 estimated as 0.01625: log likelihood = 36.51, aic = -63.02
log.diff.gdp.arma21.diagnostics
## $fit
##
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
      xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
       optim.control = list(trace = trc, REPORT = 1, reltol = tol))
##
## Coefficients:
##
           ar1
                   ar2
                            ma1
                                  xmean
        0.4057 0.0918 -0.3718 0.0504
##
## s.e. 0.5161 0.1378
                         0.5064 0.0209
## sigma^2 estimated as 0.01625: log likelihood = 36.51, aic = -63.02
## $degrees_of_freedom
## [1] 53
##
## $ttable
                     SE t.value p.value
##
        Estimate
         0.4057 0.5161 0.7862 0.4353
          0.0918 0.1378 0.6661 0.5082
## ar2
## ma1
         -0.3718 0.5064 -0.7343 0.4660
## xmean 0.0504 0.0209 2.4092 0.0195
## $ICs
         AIC
                   AICc
## -1.1056435 -1.0921482 -0.9264285
# Model 8: ARMA(2,2) (final)
log.diff.gdp.arma22 \leftarrow arima(log.diff.gdp.ts, order = c(2,0,2))
log.diff.gdp.arma22.diagnostics <- sarima(log.diff.gdp.ts, 2, 0, 2)</pre>
## initial value -2.035413
## iter
        2 value -2.038802
## iter
       3 value -2.040478
## iter
        4 value -2.040641
## iter
       5 value -2.040791
       6 value -2.042896
## iter
## iter
        7 value -2.044919
## iter 8 value -2.044997
```

```
## iter
        9 value -2.045682
## iter 10 value -2.045871
## iter 11 value -2.046747
## iter 12 value -2.047273
## iter
        13 value -2.047581
## iter 14 value -2.047650
## iter 15 value -2.047702
        16 value -2.047784
## iter
## iter 17 value -2.047792
## iter
        18 value -2.047798
## iter
        19 value -2.047815
## iter
        20 value -2.047853
## iter
       21 value -2.047961
        22 value -2.048189
## iter
## iter
        23 value -2.048265
## iter
        24 value -2.048319
       25 value -2.048351
## iter
## iter
        26 value -2.048358
       27 value -2.048491
## iter
## iter 28 value -2.048729
## iter 29 value -2.049412
## iter 30 value -2.050278
## iter 31 value -2.050624
## iter 32 value -2.050668
## iter 33 value -2.051024
## iter
       34 value -2.051982
## iter
        35 value -2.056774
        36 value -2.059606
## iter
## iter
        37 value -2.062725
## iter
       38 value -2.067098
## iter
        39 value -2.072222
## iter
        40 value -2.088772
## iter
        41 value -2.094336
       42 value -2.097030
## iter
## iter
        43 value -2.099760
## iter 44 value -2.099886
## iter 45 value -2.100899
## iter 46 value -2.101130
## iter 47 value -2.101203
## iter 48 value -2.101207
        49 value -2.101215
## iter
## iter 50 value -2.101240
## iter 51 value -2.101241
## iter 51 value -2.101241
## iter 51 value -2.101241
## final value -2.101241
## converged
## initial value -2.050397
## iter
        2 value -2.056026
## iter
        3 value -2.081594
## iter
         4 value -2.084062
## iter
        5 value -2.085851
## iter
        6 value -2.086765
## iter
        7 value -2.087484
```

```
## iter 8 value -2.089790
## iter 9 value -2.091474
## iter 10 value -2.092000
## iter 11 value -2.093082
## iter 12 value -2.093234
## iter 13 value -2.093239
## iter 14 value -2.093261
## iter 15 value -2.093270
## iter 16 value -2.093279
## iter 17 value -2.093280
## iter 18 value -2.093280
## iter 19 value -2.093280
## iter 20 value -2.093280
## iter 21 value -2.093280
## iter 21 value -2.093280
## iter 21 value -2.093280
## final value -2.093280
## converged
## <><><><>
##
## Coefficients:
##
       Estimate
                    SE t.value p.value
## ar1
        1.5232 0.1236 12.3269 0.0000
## ar2
         -0.9044 0.0924 -9.7861 0.0000
## ma1
       -1.5297 0.1006 -15.2047 0.0000
## ma2
         1.0000 0.1000 10.0004 0.0000
## xmean 0.0508 0.0196
                        2.5979 0.0122
## sigma^2 estimated as 0.01443647 on 52 degrees of freedom
## AIC = -1.138156 AICc = -1.117517 BIC = -0.9230984
##
```



log.diff.gdp.arma22

##

```
## Call:
## arima(x = log.diff.gdp.ts, order = c(2, 0, 2))
## Coefficients:
##
            ar1
                     ar2
                              ma1
                                   ma2
                                         intercept
##
         1.5232
                -0.9044
                          -1.5297
                                   1.0
                                           0.0508
                  0.0924
                           0.1006 0.1
                                           0.0196
## s.e. 0.1236
## sigma^2 estimated as 0.01444: log likelihood = 38.44, aic = -64.87
log.diff.gdp.arma22.diagnostics
```

```
## $fit
##
## Call:
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
       xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
##
       optim.control = list(trace = trc, REPORT = 1, reltol = tol))
##
## Coefficients:
##
            ar1
                     ar2
                              ma1
                                   ma2
                                          xmean
                 -0.9044
                                        0.0508
##
         1.5232
                          -1.5297
                                   1.0
                  0.0924
                                        0.0196
## s.e.
        0.1236
                           0.1006
                                  0.1
## sigma^2 estimated as 0.01444: log likelihood = 38.44, aic = -64.87
```

```
##
## $degrees_of_freedom
## [1] 52
##
## $ttable
##
                      SE t.value p.value
        Estimate
## ar1
          1.5232 0.1236 12.3269 0.0000
          -0.9044 0.0924 -9.7861 0.0000
## ar2
## ma1
         -1.5297 0.1006 -15.2047
                                   0.0000
## ma2
           1.0000 0.1000 10.0004 0.0000
## xmean
           0.0508 0.0196
                           2.5979 0.0122
##
## $ICs
##
          AIC
                    AICc
                                BIC
## -1.1381564 -1.1175166 -0.9230984
```

GDP Final Model: ARMA(2,2)

Imports

```
# Model Selection on Log-Differenced Imports (stationary)
# Imports
# Model 1: MA(1)
log.diff.import.ma1 \leftarrow arima(log.diff.import.ts, order = c(0,0,1))
log.diff.import.ma1.diagnostics <- sarima(log.diff.import.ts, 0, 0, 1)</pre>
## initial value -2.132354
## iter 2 value -2.139659
## iter
       3 value -2.143882
## iter 4 value -2.144278
## iter 5 value -2.144320
## iter
         6 value -2.144360
## iter
       7 value -2.144361
## iter
       8 value -2.144361
        9 value -2.144361
## iter
         9 value -2.144361
## iter
## iter
         9 value -2.144361
## final value -2.144361
## converged
## initial value -2.143999
## iter
        2 value -2.144005
## iter
       3 value -2.144005
        4 value -2.144007
## iter
## iter
       4 value -2.144007
## iter
         4 value -2.144007
## final value -2.144007
## converged
## <><><><>
##
## Coefficients:
                     SE t.value p.value
        Estimate
         -0.2314 0.1836 -1.2604 0.2128
## ma1
## xmean -0.0016 0.0120 -0.1367 0.8917
```

```
##
## sigma^2 estimated as 0.01371892 on 55 degrees of freedom
##
## AIC = -1.344874 AICc = -1.340975 BIC = -1.237345
##
      Model: (0,0,1)
                                    Standardized Residuals
   4
   \alpha
   0
   7
                                  1980
                                                 1990
      1960
                    1970
                                                                2000
                                                                              2010
                                               Time
               ACF of Residuals
                                                      Normal Q-Q Plot of Std Residuals
                                                Sample Quantiles
   0.0
                                                   0
           2
                   4
                           6
                                   8
                                           10
                                                         -2
                                                                 -1
                                                                         0
                                                                                 1
                                                                                        2
                       LAG
                                                                Theoretical Quantiles
                               p values for Ljung-Box statistic
p value
                                                                                 0
                                                                            0
                               0
   0
            - O-
                      5
                                            10
                                                                   15
                                                                                         20
                                              LAG (H)
log.diff.import.ma1
##
## Call:
## arima(x = log.diff.import.ts, order = c(0, 0, 1))
##
## Coefficients:
##
             ma1
                   intercept
##
         -0.2314
                     -0.0016
## s.e.
          0.1836
                      0.0120
## sigma^2 estimated as 0.01372: log likelihood = 41.33, aic = -76.66
log.diff.import.ma1.diagnostics
## $fit
```

arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),

xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,

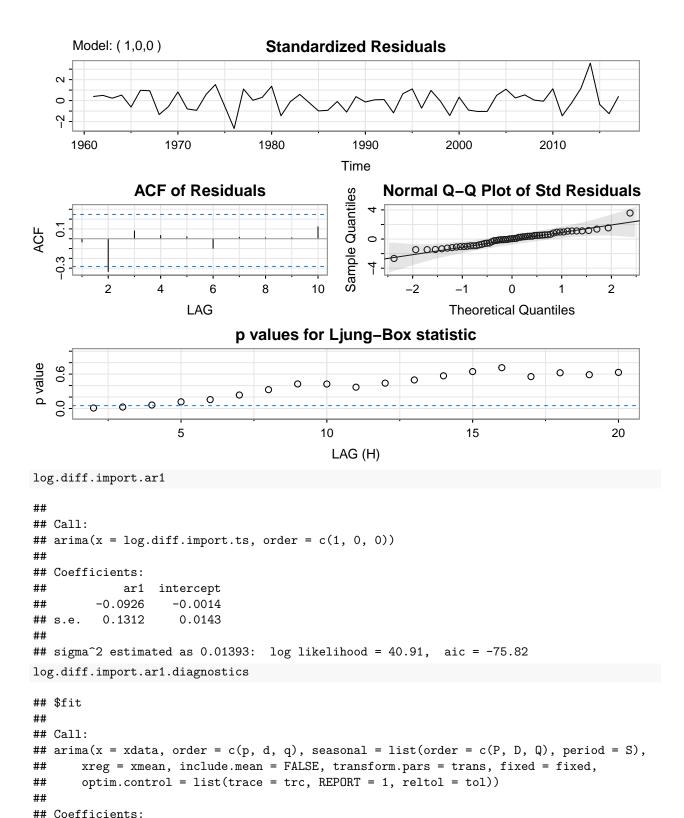
optim.control = list(trace = trc, REPORT = 1, reltol = tol))

##

##

##

```
## Coefficients:
##
            ma1
                   xmean
##
        -0.2314 -0.0016
## s.e. 0.1836
                 0.0120
## sigma^2 estimated as 0.01372: log likelihood = 41.33, aic = -76.66
## $degrees_of_freedom
## [1] 55
##
## $ttable
##
                     SE t.value p.value
        Estimate
         -0.2314 0.1836 -1.2604 0.2128
## xmean -0.0016 0.0120 -0.1367 0.8917
##
## $ICs
##
         AIC
                 AICc
                            BIC
## -1.344874 -1.340975 -1.237345
# Model 2: AR(1)
log.diff.import.ar1 \leftarrow arima(log.diff.import.ts, order = c(1,0,0))
log.diff.import.ar1.diagnostics <- sarima(log.diff.import.ts, 1, 0, 0)</pre>
## initial value -2.124853
## iter 2 value -2.129291
## iter 3 value -2.129301
## iter 4 value -2.129306
## iter 4 value -2.129306
## iter 4 value -2.129306
## final value -2.129306
## converged
## initial value -2.136680
## iter 2 value -2.136694
## iter 3 value -2.136701
## iter 3 value -2.136701
## iter 3 value -2.136701
## final value -2.136701
## converged
## <><><><>
##
## Coefficients:
        Estimate
                     SE t.value p.value
         -0.0926 0.1312 -0.7056 0.4834
## ar1
## xmean -0.0014 0.0143 -0.1002 0.9205
##
## sigma^2 estimated as 0.01393219 on 55 degrees of freedom
## AIC = -1.330262 AICc = -1.326364 BIC = -1.222733
##
```



$sigma^2$ estimated as 0.01393: log likelihood = 40.91, aic = -75.82

##

##

##

s.e.

ar1

-0.0926

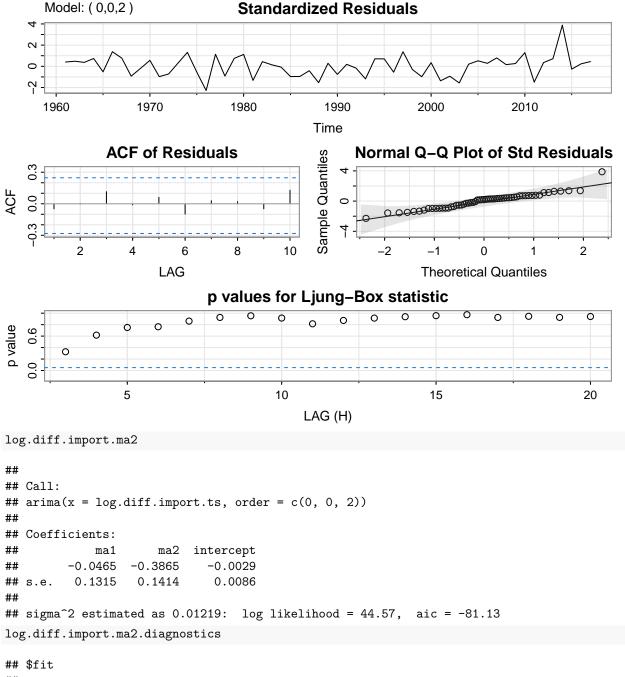
0.1312

xmean

-0.0014

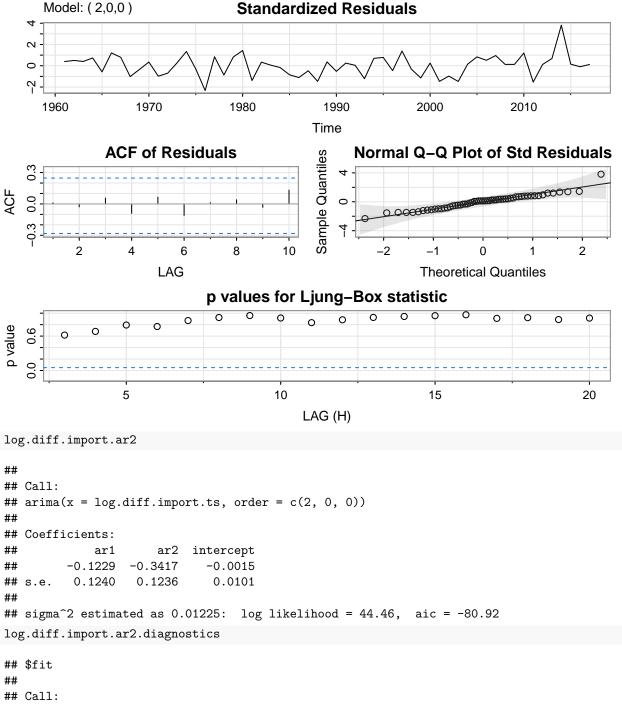
0.0143

```
##
## $degrees_of_freedom
## [1] 55
##
## $ttable
##
                     SE t.value p.value
        Estimate
       -0.0926 0.1312 -0.7056 0.4834
## xmean -0.0014 0.0143 -0.1002 0.9205
##
## $ICs
##
        AIC
                 AICc
                            BIC
## -1.330262 -1.326364 -1.222733
# Model 3: MA(2)
log.diff.import.ma2 \leftarrow arima(log.diff.import.ts, order = c(0,0,2))
log.diff.import.ma2.diagnostics <- sarima(log.diff.import.ts, 0, 0, 2)</pre>
## initial value -2.132354
## iter 2 value -2.199743
## iter 3 value -2.200444
## iter 4 value -2.201470
## iter 5 value -2.202071
## iter 6 value -2.202083
## iter 7 value -2.202083
## iter 7 value -2.202083
## iter 7 value -2.202083
## final value -2.202083
## converged
## initial value -2.200774
## iter 2 value -2.200797
## iter 3 value -2.200822
## iter 4 value -2.200824
## iter 5 value -2.200824
## iter 5 value -2.200824
## iter
         5 value -2.200824
## final value -2.200824
## converged
## <><><><><>
##
## Coefficients:
        Estimate
                     SE t.value p.value
         -0.0465 0.1315 -0.3538 0.7249
## ma1
## ma2
         -0.3865 0.1414 -2.7324 0.0085
## xmean -0.0029 0.0086 -0.3335 0.7401
##
## sigma^2 estimated as 0.01218653 on 54 degrees of freedom
## AIC = -1.423419 AICc = -1.415475 BIC = -1.280047
##
```



```
##
## Call:
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
       xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
##
       optim.control = list(trace = trc, REPORT = 1, reltol = tol))
##
## Coefficients:
##
             ma1
                      ma2
                             xmean
                  -0.3865
                           -0.0029
##
         -0.0465
                            0.0086
## s.e.
          0.1315
                   0.1414
##
## sigma^2 estimated as 0.01219: log likelihood = 44.57, aic = -81.13
```

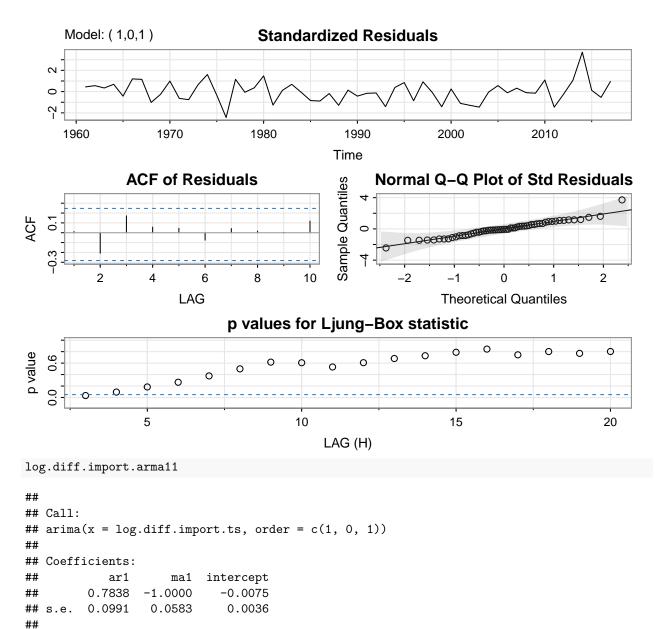
```
##
## $degrees_of_freedom
## [1] 54
##
## $ttable
##
        Estimate
                     SE t.value p.value
       -0.0465 0.1315 -0.3538 0.7249
## ma1
         -0.3865 0.1414 -2.7324 0.0085
## ma2
## xmean -0.0029 0.0086 -0.3335 0.7401
##
## $ICs
##
        AIC
                 AICc
## -1.423419 -1.415475 -1.280047
# Model 4: AR(2)
log.diff.import.ar2 <- arima(log.diff.import.ts, order = c(2,0,0))
log.diff.import.ar2.diagnostics <- sarima(log.diff.import.ts, 2, 0, 0)</pre>
## initial value -2.117685
## iter 2 value -2.186025
## iter 3 value -2.186679
## iter 4 value -2.186969
## iter 5 value -2.187031
## iter 6 value -2.187032
## iter 6 value -2.187032
## iter 6 value -2.187032
## final value -2.187032
## converged
## initial value -2.198846
## iter 2 value -2.198923
## iter 3 value -2.198958
## iter 4 value -2.198960
## iter 5 value -2.198962
## iter 5 value -2.198962
## iter 5 value -2.198962
## final value -2.198962
## converged
## <><><><><>
##
## Coefficients:
        Estimate
                     SE t.value p.value
         -0.1229 0.1240 -0.9912 0.3260
## ar1
## ar2
         -0.3417 0.1236 -2.7637 0.0078
## xmean -0.0015 0.0101 -0.1499 0.8814
##
## sigma^2 estimated as 0.01224758 on 54 degrees of freedom
## AIC = -1.419696 AICc = -1.411751 BIC = -1.276324
##
```



```
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
       xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
##
       optim.control = list(trace = trc, REPORT = 1, reltol = tol))
##
## Coefficients:
##
             ar1
                      ar2
                             xmean
         -0.1229
                           -0.0015
##
                  -0.3417
                   0.1236
                            0.0101
## s.e.
          0.1240
##
## sigma^2 estimated as 0.01225: log likelihood = 44.46, aic = -80.92
```

```
##
## $degrees_of_freedom
  [1] 54
##
## $ttable
##
         Estimate
                      SE t.value p.value
         -0.1229 0.1240 -0.9912 0.3260
## ar1
          -0.3417 0.1236 -2.7637 0.0078
## ar2
## xmean -0.0015 0.0101 -0.1499 0.8814
##
## $ICs
##
         AIC
                  AICc
                             BIC
## -1.419696 -1.411751 -1.276324
# Model 5: ARMA(1,1) (final)
log.diff.import.arma11 \leftarrow arima(log.diff.import.ts, order = c(1,0,1))
log.diff.import.arma11.diagnostics <- sarima(log.diff.import.ts, 1, 0, 1)</pre>
## initial value -2.124853
## iter
         2 value -2.131198
          3 value -2.133231
## iter
        4 value -2.137431
## iter
## iter
         5 value -2.149287
## iter
         6 value -2.160505
## iter
        7 value -2.163489
        8 value -2.164322
## iter
## iter
         9 value -2.164496
## iter 10 value -2.164541
## iter 11 value -2.164592
## iter
        12 value -2.164651
## iter 13 value -2.164733
## iter
        14 value -2.164745
## iter 15 value -2.164927
## iter
        16 value -2.165021
## iter
        17 value -2.165204
## iter
        18 value -2.166863
        19 value -2.167436
## iter
        20 value -2.176018
## iter
## iter
        21 value -2.178987
        22 value -2.192394
## iter
## iter 23 value -2.194201
## iter 24 value -2.200608
## iter 25 value -2.203958
## iter 26 value -2.210785
## iter 27 value -2.211433
## iter 28 value -2.219517
## iter 29 value -2.220264
## iter 30 value -2.226721
## iter 31 value -2.239600
## iter 32 value -2.253113
## iter 33 value -2.260094
## iter 34 value -2.260647
## iter
        35 value -2.263222
## iter 36 value -2.264100
## iter 36 value -2.264100
```

```
## iter 37 value -2.264274
## iter 37 value -2.264274
## iter 38 value -2.264308
## iter 38 value -2.264308
## iter 39 value -2.264316
## iter 39 value -2.264316
## iter 40 value -2.264317
## iter 40 value -2.264317
## iter 41 value -2.264318
## iter 41 value -2.264318
## iter 42 value -2.264318
## iter 42 value -2.264318
## iter 42 value -2.264318
## final value -2.264318
## converged
## initial value -2.151965
## iter
        2 value -2.159581
## iter 3 value -2.165801
## iter 4 value -2.167893
## iter 5 value -2.168446
## iter 6 value -2.168480
## iter 7 value -2.168526
## iter 8 value -2.168538
## iter
        9 value -2.168538
## iter 10 value -2.168539
## iter 11 value -2.168542
## iter 12 value -2.168542
## iter 12 value -2.168542
## iter 12 value -2.168542
## final value -2.168542
## converged
## <><><><>
##
## Coefficients:
       Estimate
                     SE t.value p.value
## ar1
        0.7838 0.0991
                        7.9080
                                  0.000
## ma1
         -1.0000 0.0583 -17.1426
                                  0.000
## xmean -0.0075 0.0036 -2.0729
                                  0.043
##
## sigma^2 estimated as 0.0126088 on 54 degrees of freedom
## AIC = -1.358857 AICc = -1.350913 BIC = -1.215485
```



```
## sigma^2 estimated as 0.01261: log likelihood = 42.73, aic = -77.45
log.diff.import.arma11.diagnostics
```

```
## $fit
##
## Call:
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
       xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
##
       optim.control = list(trace = trc, REPORT = 1, reltol = tol))
##
## Coefficients:
##
            ar1
                     ma1
                            xmean
         0.7838
                 -1.0000
                          -0.0075
##
                  0.0583
                           0.0036
## s.e.
        0.0991
##
## sigma^2 estimated as 0.01261: log likelihood = 42.73, aic = -77.45
```

```
##
## $degrees_of_freedom
## [1] 54
##
## $ttable
##
        Estimate
                     SE t.value p.value
         0.7838 0.0991
                         7.9080
## ar1
## ma1
         -1.0000 0.0583 -17.1426
                                   0.000
## xmean -0.0075 0.0036 -2.0729
##
## $ICs
##
         AIC
                 AICc
                            BIC
## -1.358857 -1.350913 -1.215485
# Model 6: ARMA(1,2)
log.diff.import.arma12 <- arima(log.diff.import.ts, order = c(1,0,2))</pre>
log.diff.import.arma12.diagnostics <- sarima(log.diff.import.ts, 1, 0, 2)</pre>
## initial value -2.124853
        2 value -2.190526
## iter
        3 value -2.196281
## iter
       4 value -2.197046
## iter
## iter
        5 value -2.198253
## iter 6 value -2.199359
## iter
       7 value -2.201100
        8 value -2.202005
## iter
## iter
        9 value -2.202359
## iter 10 value -2.202363
## iter 11 value -2.202364
## iter 12 value -2.202364
## iter 13 value -2.202364
## iter 14 value -2.202364
## iter 15 value -2.202364
## iter 15 value -2.202364
## iter 15 value -2.202364
## final value -2.202364
## converged
## initial value -2.209843
## iter
        2 value -2.209920
## iter
        3 value -2.210058
       4 value -2.210177
## iter
## iter
        5 value -2.210240
## iter 6 value -2.210268
## iter
        7 value -2.210269
## iter
         7 value -2.210269
## iter
         7 value -2.210269
## final value -2.210269
## converged
## <><><><><>
##
## Coefficients:
##
                     SE t.value p.value
        Estimate
## ar1
         -0.3312 0.2902 -1.1412 0.2589
## ma1
          0.2433 0.2720 0.8944 0.3751
## ma2
         -0.4033 0.1267 -3.1820 0.0024
```

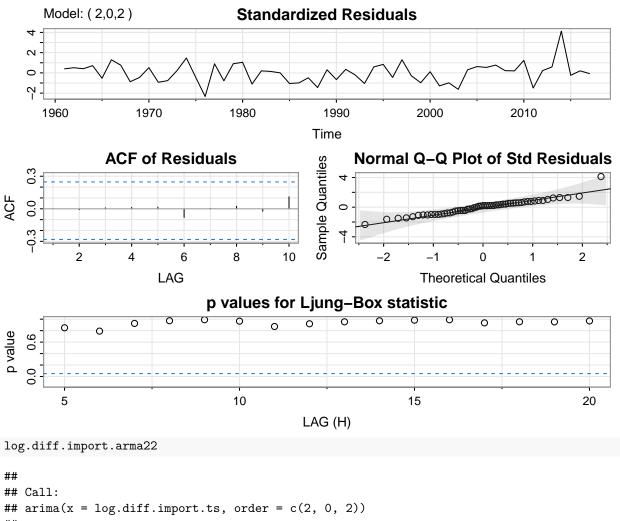
```
## xmean -0.0025 0.0093 -0.2631 0.7935
##
## sigma^2 estimated as 0.01194652 on 53 degrees of freedom
## AIC = -1.407221 AICc = -1.393726 BIC = -1.228006
##
      Model: (1,0,2)
                                   Standardized Residuals
    4
   \alpha
   0
   7
      1960
                    1970
                                  1980
                                                 1990
                                                               2000
                                                                             2010
                                               Time
               ACF of Residuals
                                                     Normal Q-Q Plot of Std Residuals
                                               Sample Quantiles
   0.3
   0.0
                                          10
                           6
                                   8
                                                         -2
                                                                                       2
                       LAG
                                                               Theoretical Quantiles
                               p values for Ljung-Box statistic
                  0
p value
   9.0
   0.0
                                      10
             5
                                                               15
                                                                                        20
                                             LAG (H)
log.diff.import.arma12
##
## Call:
## arima(x = log.diff.import.ts, order = c(1, 0, 2))
##
## Coefficients:
##
                                     intercept
                      ma1
                                ma2
          -0.3312
                   0.2433
                           -0.4033
                                       -0.0025
##
                                        0.0093
          0.2902
                   0.2720
                             0.1267
##
## sigma^2 estimated as 0.01195: log likelihood = 45.11, aic = -80.21
log.diff.import.arma12.diagnostics
## $fit
##
## Call:
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
##
       xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
       optim.control = list(trace = trc, REPORT = 1, reltol = tol))
```

```
##
## Coefficients:
##
                    ma1
                             ma2
                                    xmean
        -0.3312 0.2433 -0.4033 -0.0025
##
## s.e.
         0.2902 0.2720
                         0.1267
                                   0.0093
##
## sigma^2 estimated as 0.01195: log likelihood = 45.11, aic = -80.21
##
## $degrees_of_freedom
## [1] 53
##
## $ttable
        Estimate
                     SE t.value p.value
         -0.3312 0.2902 -1.1412 0.2589
## ar1
## ma1
          0.2433 0.2720 0.8944 0.3751
## ma2
         -0.4033 0.1267 -3.1820 0.0024
## xmean -0.0025 0.0093 -0.2631 0.7935
##
## $ICs
##
        AIC
                 AICc
                            BIC
## -1.407221 -1.393726 -1.228006
# Model 7: ARMA(2,1)
log.diff.import.arma21 <- arima(log.diff.import.ts, order = c(2,0,1))</pre>
log.diff.import.arma21.diagnostics <- sarima(log.diff.import.ts, 2, 0, 1)
## initial value -2.117685
## iter 2 value -2.183588
## iter 3 value -2.185259
## iter 4 value -2.185930
## iter 5 value -2.186463
## iter 6 value -2.186556
        7 value -2.187485
## iter
## iter
       8 value -2.187673
        9 value -2.187720
## iter
## iter 10 value -2.187726
## iter 11 value -2.187727
## iter 11 value -2.187727
## iter 11 value -2.187727
## final value -2.187727
## converged
## initial value -2.199962
## iter 2 value -2.200019
## iter 3 value -2.200062
## iter 4 value -2.200062
## iter 5 value -2.200063
## iter 6 value -2.200064
## iter
       7 value -2.200069
       8 value -2.200072
## iter
## iter
        9 value -2.200074
## iter 10 value -2.200074
## iter 10 value -2.200074
## final value -2.200074
## converged
## <><><><>
```

```
##
## Coefficients:
         Estimate
                       SE t.value p.value
##
          -0.3207 0.7160 -0.4479 0.6561
## ar1
##
   ar2
          -0.3531 0.1222 -2.8890 0.0056
           0.2253 0.8057 0.2797
                                   0.7808
##
   ma1
   xmean -0.0016 0.0108 -0.1486 0.8824
##
## sigma^2 estimated as 0.01221868 on 53 degrees of freedom
##
## AIC = -1.386832 AICc = -1.373337 BIC = -1.207617
##
      Model: (2,0,1)
                                   Standardized Residuals
   4
   ^{\circ}
   0
      1960
                    1970
                                  1980
                                                1990
                                                              2000
                                                                            2010
                                              Time
               ACF of Residuals
                                                     Normal Q-Q Plot of Std Residuals
                                               Sample Quantiles
                                                  0
           2
                   4
                          6
                                  8
                                          10
                                                        -2
                                                                -1
                                                                        0
                                                                               1
                                                                                       2
                       LAG
                                                               Theoretical Quantiles
                              p values for Ljung-Box statistic
p value
             5
                                      10
                                                               15
                                                                                       20
                                             LAG (H)
log.diff.import.arma21
##
## arima(x = log.diff.import.ts, order = c(2, 0, 1))
## Coefficients:
##
             ar1
                       ar2
                               ma1
                                     intercept
                                       -0.0016
##
         -0.3207
                   -0.3531
                            0.2253
## s.e.
          0.7160
                    0.1222 0.8057
                                        0.0108
## sigma^2 estimated as 0.01222: log likelihood = 44.52, aic = -79.05
```

```
log.diff.import.arma21.diagnostics
## $fit
##
## Call:
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
       xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
       optim.control = list(trace = trc, REPORT = 1, reltol = tol))
##
## Coefficients:
##
            ar1
                     ar2
                             ma1
                                     xmean
                -0.3531 0.2253
##
        -0.3207
                                  -0.0016
        0.7160
                  0.1222 0.8057
                                   0.0108
##
## sigma^2 estimated as 0.01222: log likelihood = 44.52, aic = -79.05
## $degrees_of_freedom
## [1] 53
##
## $ttable
##
                     SE t.value p.value
        Estimate
## ar1
         -0.3207 0.7160 -0.4479 0.6561
## ar2
         -0.3531 0.1222 -2.8890 0.0056
## ma1
          0.2253 0.8057 0.2797 0.7808
## xmean -0.0016 0.0108 -0.1486 0.8824
##
## $ICs
##
        AIC
                 AICc
                             BIC
## -1.386832 -1.373337 -1.207617
# Model 8: ARMA(2,2)
log.diff.import.arma22 <- arima(log.diff.import.ts, order = c(2,0,2))</pre>
log.diff.import.arma22.diagnostics <- sarima(log.diff.import.ts, 2, 0, 2)
## initial value -2.117685
## iter 2 value -2.146070
## iter
        3 value -2.189558
## iter
        4 value -2.191196
## iter
       5 value -2.191658
## iter
        6 value -2.194157
## iter 7 value -2.196115
## iter
        8 value -2.199075
        9 value -2.200842
## iter
## iter 10 value -2.202596
## iter 11 value -2.203318
## iter 12 value -2.203457
## iter 13 value -2.203461
## iter 14 value -2.203464
## iter 15 value -2.203467
## iter 16 value -2.203469
## iter 17 value -2.203470
## iter 18 value -2.203472
## iter 19 value -2.203476
## iter 20 value -2.203477
```

```
## iter 21 value -2.203477
## iter 22 value -2.203478
## iter 23 value -2.203478
## iter 23 value -2.203478
## final value -2.203478
## converged
## initial value -2.209480
## iter 2 value -2.209834
## iter 3 value -2.209908
## iter 4 value -2.210012
## iter 5 value -2.210118
## iter 6 value -2.210246
## iter 7 value -2.210304
## iter 8 value -2.210312
## iter 9 value -2.210329
## iter 10 value -2.210339
## iter 11 value -2.210345
## iter 12 value -2.210345
## iter 12 value -2.210345
## final value -2.210345
## converged
## <><><><>
##
## Coefficients:
##
       Estimate
                    SE t.value p.value
## ar1
        -0.3480 0.3419 -1.0178 0.3135
## ar2
        -0.0325 0.3434 -0.0947 0.9249
        0.2586 0.3225 0.8019 0.4263
## ma1
       -0.3743 0.3343 -1.1197 0.2680
## ma2
## xmean -0.0024 0.0095 -0.2494 0.8040
##
## sigma^2 estimated as 0.0119451 on 52 degrees of freedom
## AIC = -1.372287 AICc = -1.351647 BIC = -1.157229
##
```



```
## Coefficients:
##
                      ar2
                              ma1
                                       ma2
                                             intercept
##
         -0.3480
                  -0.0325
                          0.2586
                                   -0.3743
                                               -0.0024
                   0.3434 0.3225
                                    0.3343
                                                0.0095
## s.e.
          0.3419
##
## sigma^2 estimated as 0.01195: log likelihood = 45.11, aic = -78.22
log.diff.import.arma22.diagnostics
```

```
## $fit
##
## Call:
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
       xreg = xmean, include.mean = FALSE, transform.pars = trans, fixed = fixed,
##
##
       optim.control = list(trace = trc, REPORT = 1, reltol = tol))
##
## Coefficients:
##
             ar1
                      ar2
                              ma1
                                        ma2
                                               xmean
                  -0.0325
                           0.2586
                                    -0.3743
                                             -0.0024
##
         -0.3480
                   0.3434
                                     0.3343
                                              0.0095
## s.e.
          0.3419
                          0.3225
##
## sigma^2 estimated as 0.01195: log likelihood = 45.11, aic = -78.22
```

```
##
## $degrees_of_freedom
## [1] 52
##
## $ttable
##
        Estimate
                     SE t.value p.value
## ar1
       -0.3480 0.3419 -1.0178 0.3135
       -0.0325 0.3434 -0.0947 0.9249
## ar2
## ma1
         0.2586 0.3225 0.8019 0.4263
## ma2
         -0.3743 0.3343 -1.1197 0.2680
## xmean -0.0024 0.0095 -0.2494 0.8040
##
## $ICs
##
        AIC
                 AICc
                            BIC
## -1.372287 -1.351647 -1.157229
```

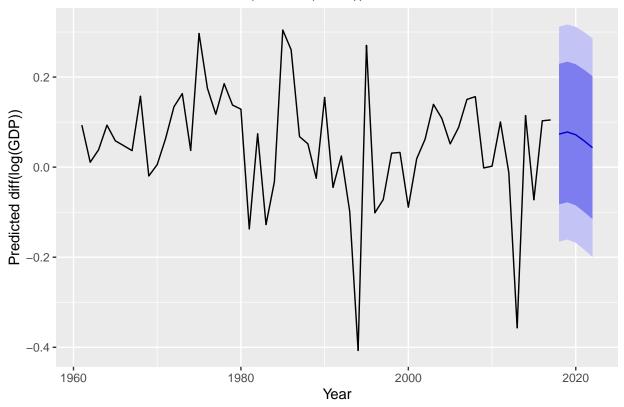
Imports Final Model: ARMA(1,1)

Forecast

GDP

```
# Forecast
gdp_forecast <- forecast(log.diff.gdp.arma22, h = 5)</pre>
gdp_forecast
                             Lo 80
##
       Point Forecast
                                       Hi 80
                                                             Hi 95
                                                   Lo 95
## 2018
           0.07335514 -0.08268150 0.2293918 -0.1652823 0.3119926
## 2019
            0.07804164 -0.07800850 0.2340918 -0.1606165 0.3166997
## 2020
           0.07190733 -0.08476349 0.2285781 -0.1677000 0.3115147
## 2021
            0.05832534 -0.09951227 0.2161630 -0.1830665 0.2997171
            0.04318533 -0.11551828 0.2018889 -0.1995309 0.2859015
autoplot(gdp_forecast) +
 ggtitle("5-Year Forecast of GDP (ARIMA(2,0,2))") +
 xlab("Year") +
 ylab("Predicted diff(log(GDP))")
```

5-Year Forecast of GDP (ARIMA(2,0,2))



Imports

```
# Imports
import_forecast <- forecast(log.diff.import.arma11, h = 5)</pre>
import_forecast
##
        Point Forecast
                            Lo 80
                                       Hi 80
                                                   Lo 95
                                                             Hi 95
## 2018
           -0.06390127 -0.2089035 0.08110099 -0.2856631 0.1578605
           -0.05171036 -0.1995994 0.09617871 -0.2778872 0.1744665
## 2019
## 2020
           -0.04215508 -0.1917900 0.10747989 -0.2710020 0.1866919
## 2021
           -0.03466560 -0.1853631 0.11603193 -0.2651376 0.1958064
## 2022
           -0.02879531 -0.1801419 0.12255130 -0.2602600 0.2026694
autoplot(import_forecast) +
  ggtitle("5-Year Forecast of Imports (ARIMA(1,0,1))") +
 xlab("Year") +
 ylab("Predicted diff(log(Imports))")
```

5-Year Forecast of Imports (ARIMA(1,0,1))

