

# 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

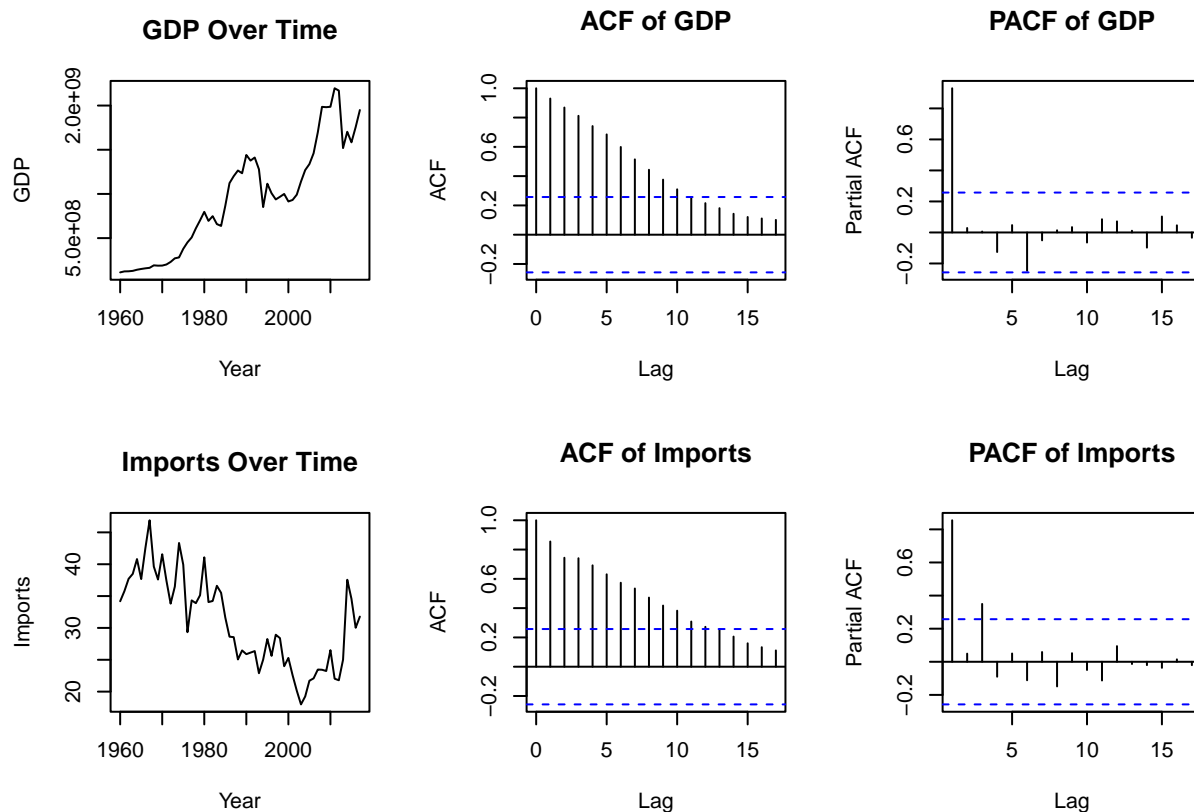
### Raw Data

```
# 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")
```



## Log-Transformed Data

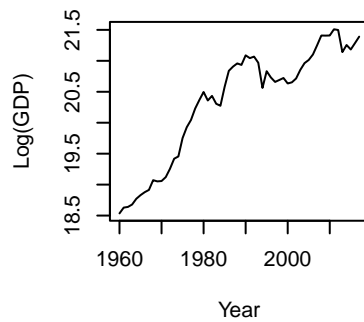
```
# Log-Transformed Data
log.gdp.ts <- ts(log(finalPro_data$GDP), start = 1960, frequency = 1)
log.import.ts <- ts(log(finalPro_data$Imports), start = 1960, frequency = 1)

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

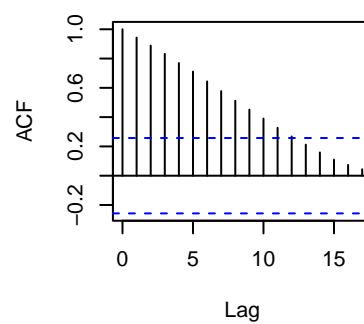
# 1. GDP
plot(log.gdp.ts, main = "Log-Transformed GDP Over Time", ylab = "Log(GDP)", xlab = "Year")
acf(log.gdp.ts, main = "ACF of Log-Transformed GDP")
pacf(log.gdp.ts, main = "PACF of Log-Transformed GDP")

# 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")
```

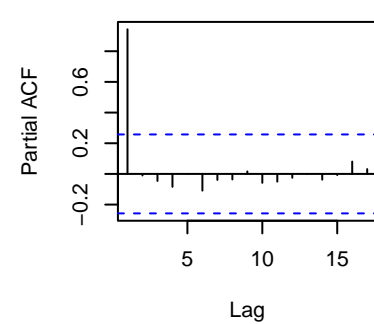
Log-Transformed GDP Over Time



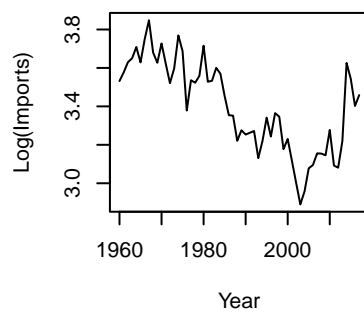
ACF of Log-Transformed GDP



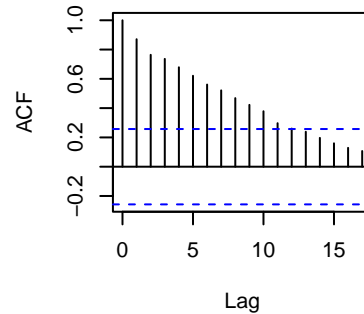
PACF of Log-Transformed GDP



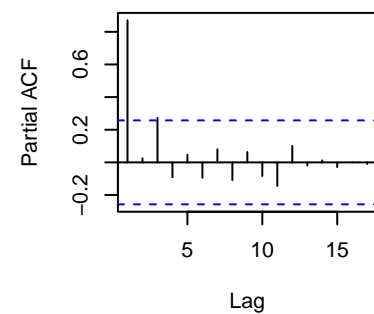
Log-Transformed Imports Over Time



ACF of Log-Transformed Imports



PACF of Log-Transformed Imports



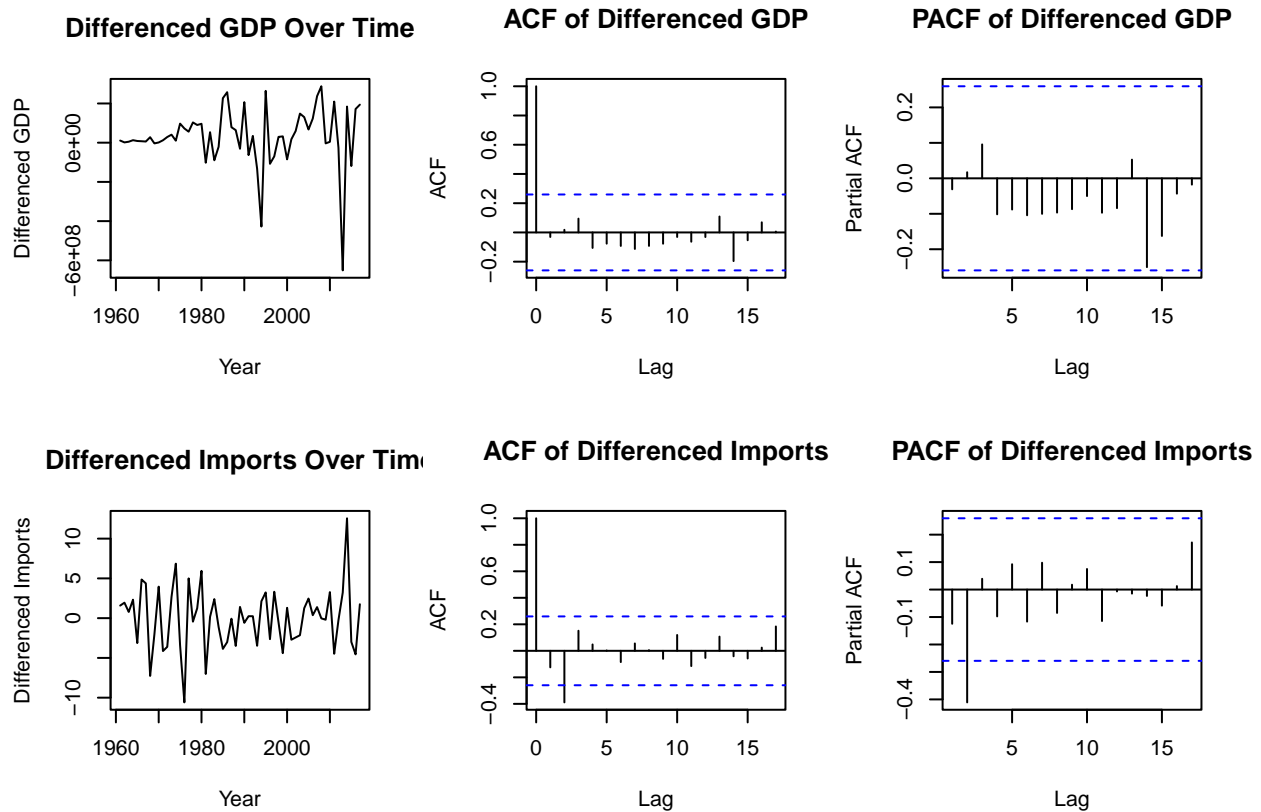
## Differencing Data

```
# Differenced Data (first difference removes first observation, so start at 1961)
diff.gdp.ts <- ts(diff(finalPro_data$GDP), start = 1961, frequency = 1)
diff.import.ts <- ts(diff(finalPro_data$Imports), start = 1961, frequency = 1)

# 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")
```



## 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)

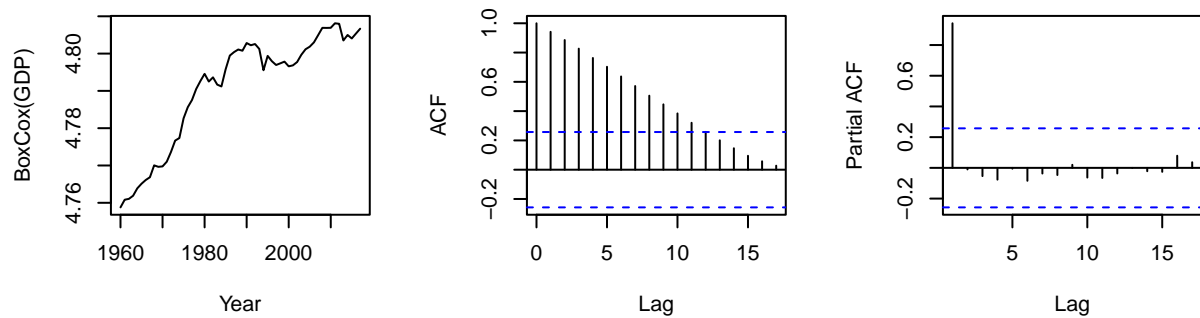
import.lambda <- BoxCox.lambda(finalPro_data$Imports)
boxcox.import.ts <- ts(BoxCox(finalPro_data$Imports, import.lambda), start = 1960, frequency = 1)

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

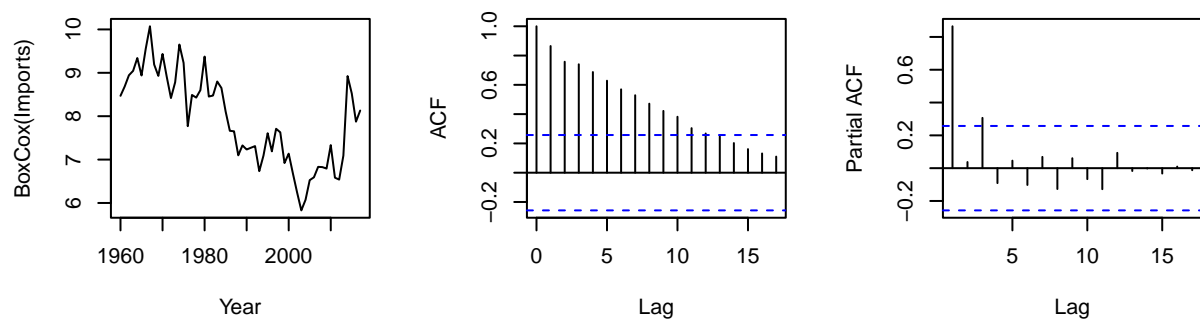
# 1. GDP
plot(boxcox.gdp.ts, main = "Box-Cox Transformed GDP Over Time", ylab = "BoxCox(GDP)", xlab = "Year")
acf(boxcox.gdp.ts, main = "ACF of Box-Cox Transformed GDP")
pacf(boxcox.gdp.ts, main = "PACF of Box-Cox Transformed GDP")

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

### Box-Cox Transformed GDP Over Time · ACF of Box-Cox Transformed GDP · PACF of Box-Cox Transformed GDP



### Box-Cox Transformed Imports Over Time · ACF of Box-Cox Transformed Imports · PACF of Box-Cox Transformed Imports



## 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))
```

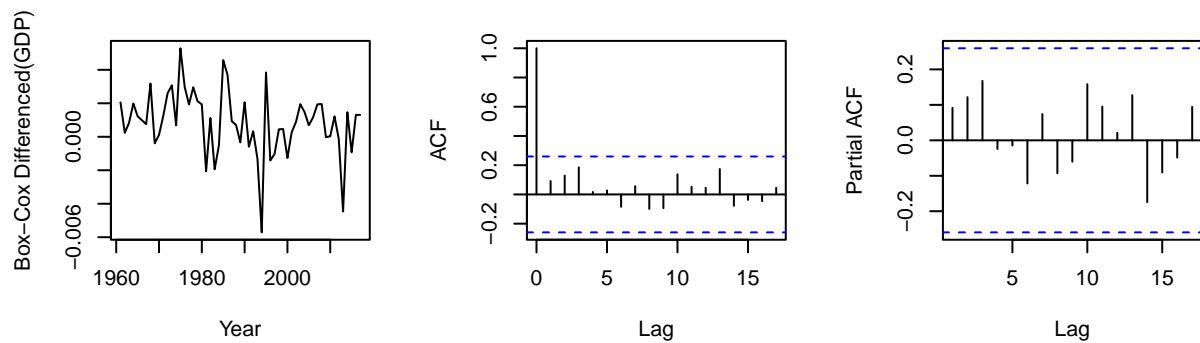
#### # 1. GDP

```
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")
```

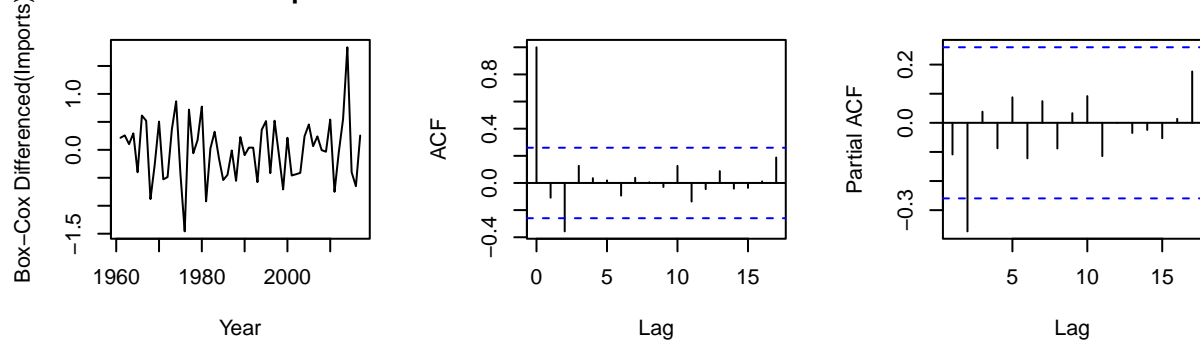
#### # 2. Imports

```
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")
```

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



## Box-Cox Differenced Imports Over ACF of Box-Cox Differenced Imp PACF of Box-Cox Differenced Imp



## Log-Differencing Data (Final)

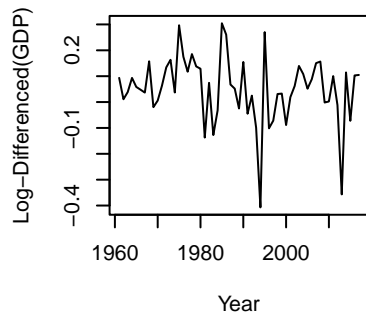
```
# Log-Differenced Data (Final)
log.diff.gdp.ts <- ts(diff(log.gdp.ts), start = 1961, frequency = 1)
log.diff.import.ts <- ts(diff(log.import.ts), start = 1961, frequency = 1)

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

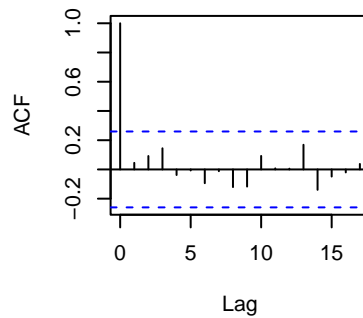
# 1. GDP
plot(log.diff.gdp.ts, main = "Log-Differenced GDP Over Time", ylab = "Log-Differenced(GDP)", xlab = "Year")
acf(log.diff.gdp.ts, main = "ACF of Log-Differenced GDP")
pacf(log.diff.gdp.ts, main = "PACF of Log-Differenced GDP")

# 2. Imports
plot(log.diff.import.ts, main = "Log-Differenced Imports Over Time", ylab = "Log-Differenced(Imports)", xlab = "Year")
acf(log.diff.import.ts, main = "ACF of Log-Differenced Imports")
pacf(log.diff.import.ts, main = "PACF of Log-Differenced Imports")
```

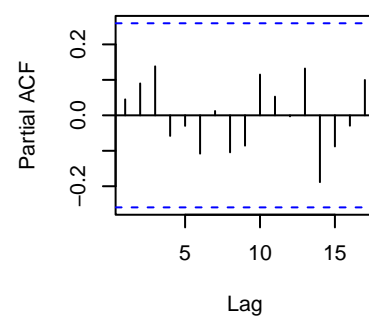
### Log-Differenced GDP Over Tin



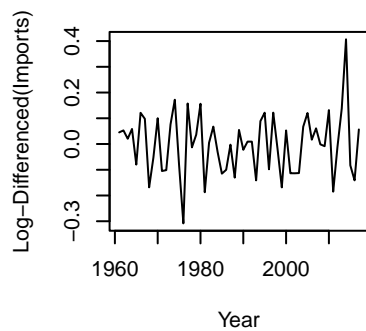
### ACF of Log-Differenced GDP



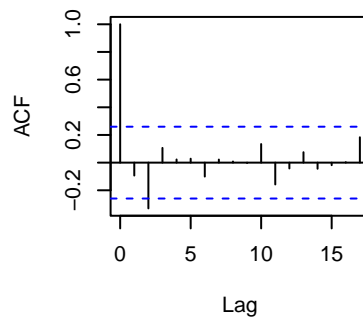
### PACF of Log-Differenced GDF



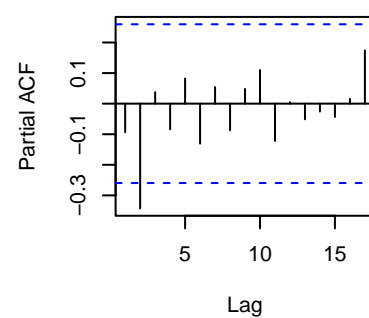
### Log-Differenced Imports Over Time



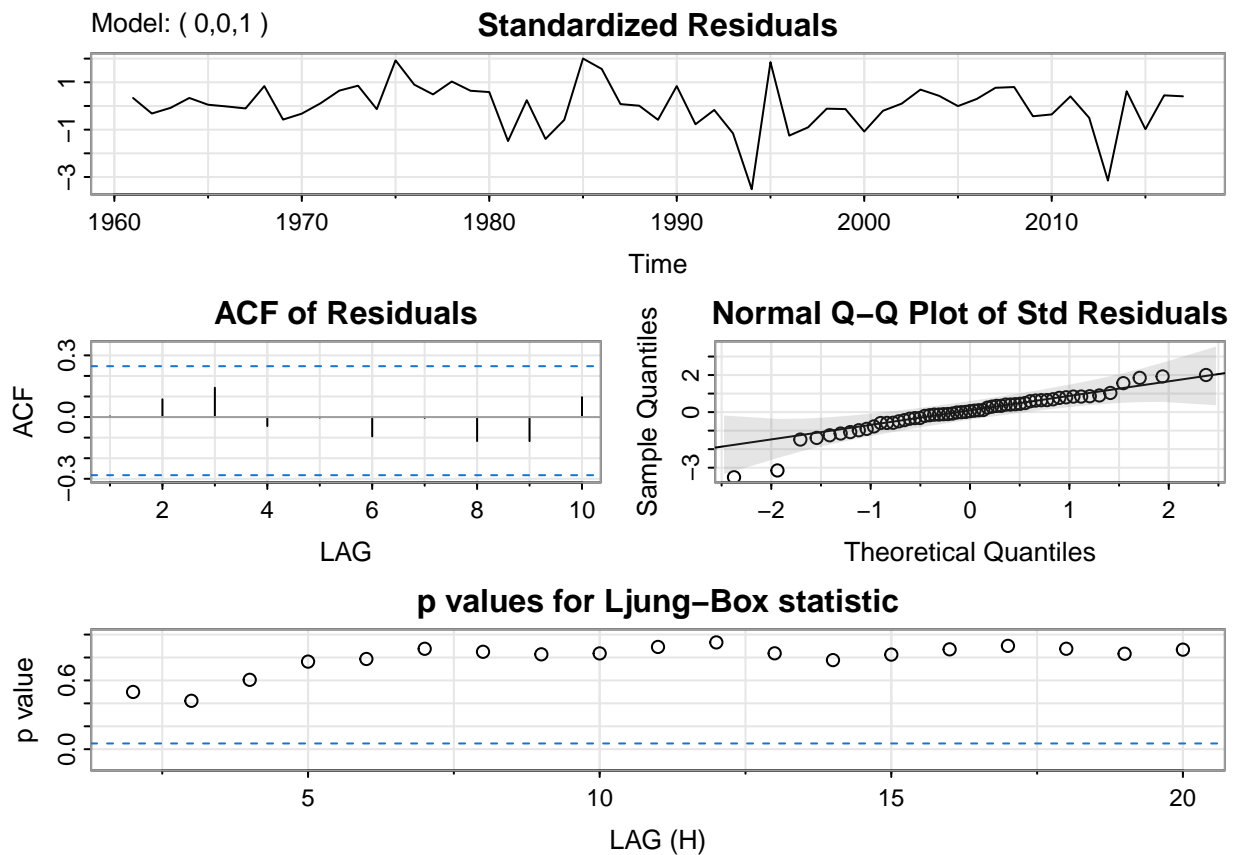
### ACF of Log-Differenced Imports



### PACF of Log-Differenced Impoi



```
## Coefficients:
##      Estimate      SE t.value p.value
## ma1      0.0382 0.1228  0.3113  0.7568
## xmean     0.0502 0.0177  2.8416  0.0063
##
## sigma^2 estimated as 0.01649574 on 55 degrees of freedom
##
## AIC = -1.161487  AICc = -1.157589  BIC = -1.053958
##
```



```
log.diff.gdp.ma1
```

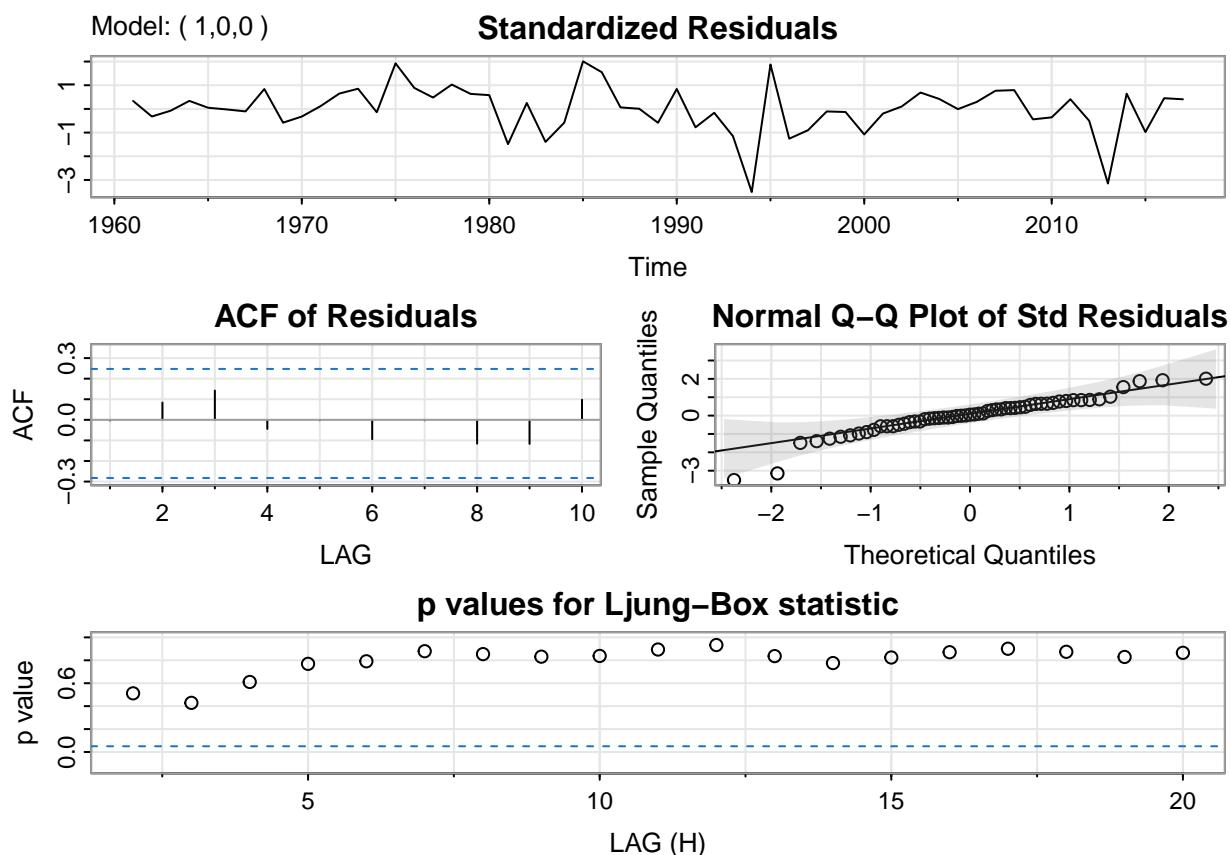
```
##
## Call:
## arima(x = log.diff.gdp.ts, order = c(0, 0, 1))
##
## Coefficients:
##      ma1  intercept
##      0.0382    0.0502
## s.e.  0.1228    0.0177
##
## sigma^2 estimated as 0.0165:  log likelihood = 36.1,  aic = -66.2
```

```
log.diff.gdp.ma1.diagnostics
```

```
## $fit
##
## Call:
```







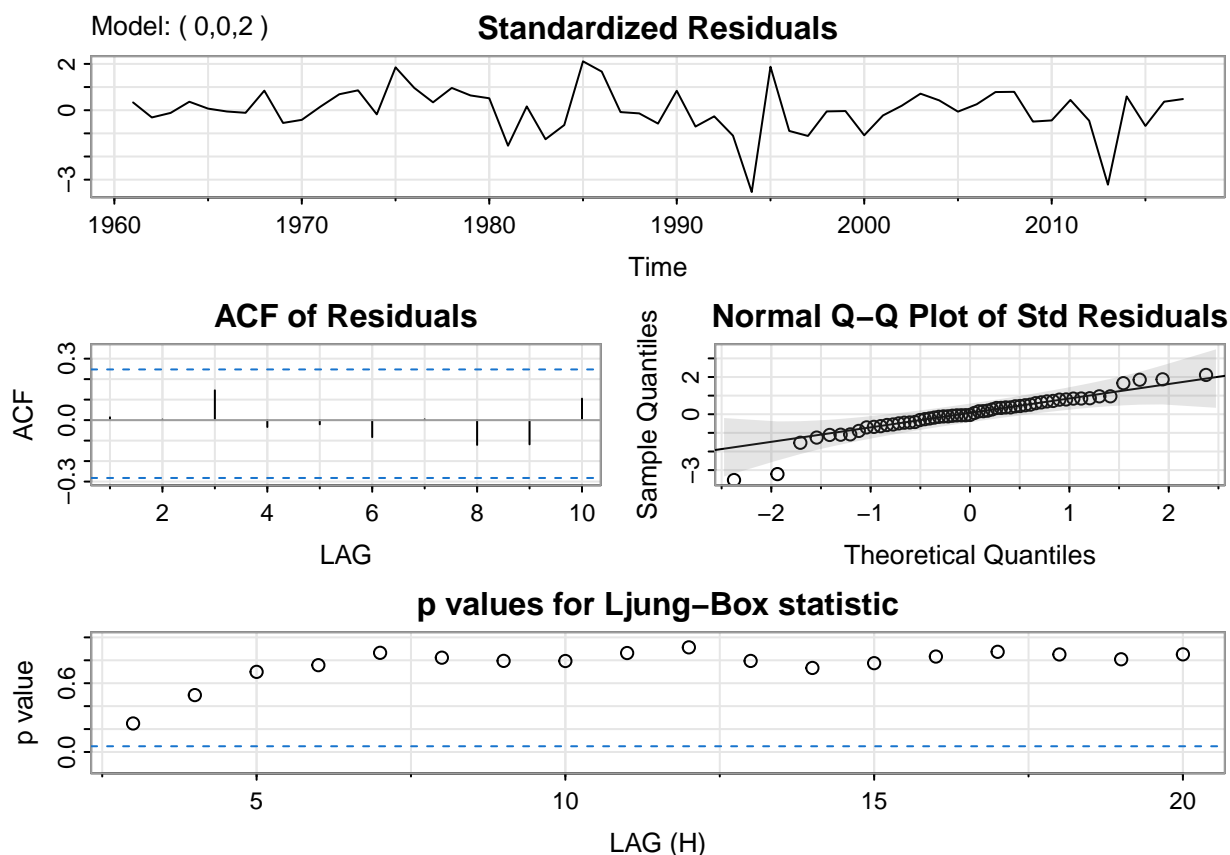
```
log.diff.gdp.ar1
```

```
##
## Call:
## arima(x = log.diff.gdp.ts, order = c(1, 0, 0))
##
## Coefficients:
##          ar1  intercept
##          0.0444    0.0502
## s.e.    0.1315    0.0178
##
## sigma^2 estimated as 0.01649:  log likelihood = 36.11,  aic = -66.22
```

```
log.diff.gdp.ar1.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   xmean
##          0.0444  0.0502
## s.e.    0.1315  0.0178
##
## sigma^2 estimated as 0.01649:  log likelihood = 36.11,  aic = -66.22
```





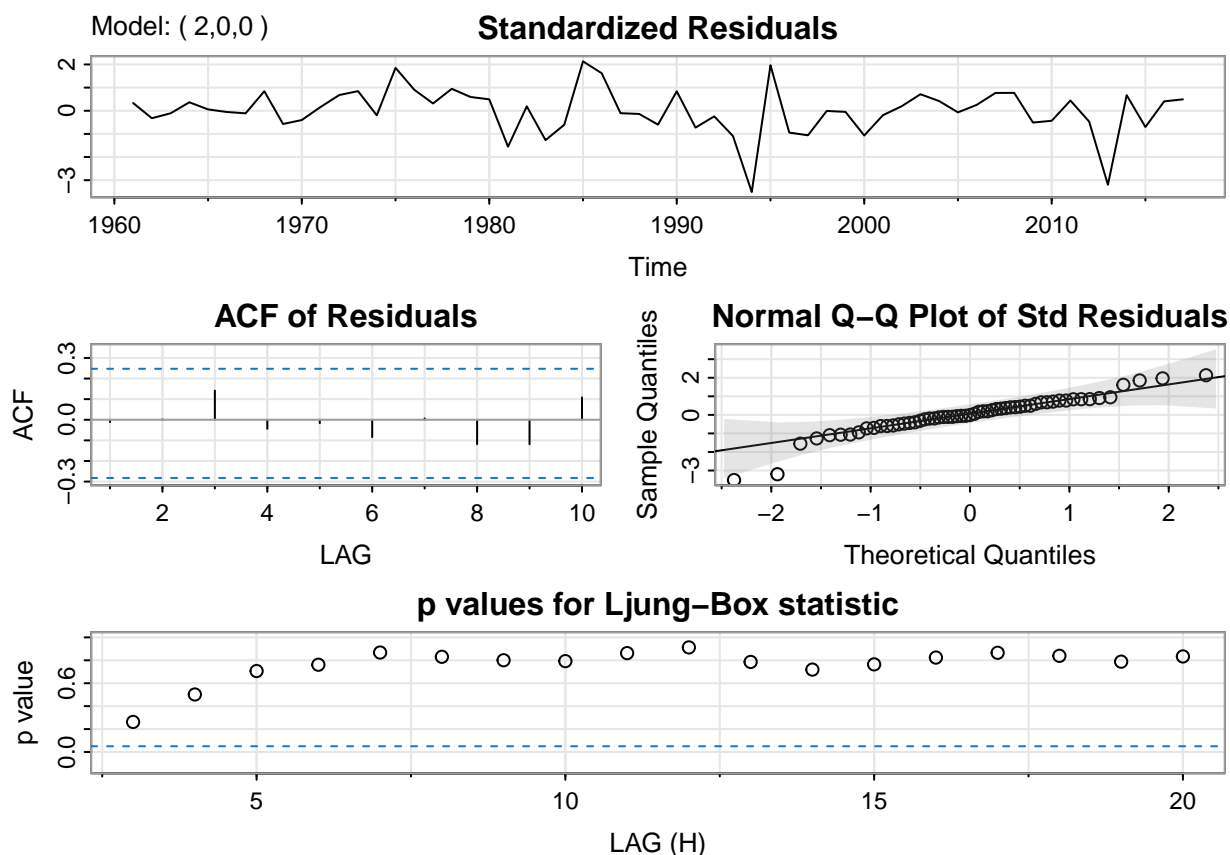
```
log.diff.gdp.ma2
```

```
##
## Call:
## arima(x = log.diff.gdp.ts, order = c(0, 0, 2))
##
## Coefficients:
##          ma1      ma2  intercept
##          0.0159 0.0908    0.0503
## s.e.  0.1367 0.1388    0.0187
##
## sigma^2 estimated as 0.01637:  log likelihood = 36.31,  aic = -64.63
```

```
log.diff.gdp.ma2.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
##          0.0159 0.0908 0.0503
## s.e.  0.1367 0.1388 0.0187
##
## sigma^2 estimated as 0.01637:  log likelihood = 36.31,  aic = -64.63
```





```
log.diff.gdp.ar2
```

```
##
## Call:
## arima(x = log.diff.gdp.ts, order = c(2, 0, 0))
##
## Coefficients:
##          ar1      ar2  intercept
##      0.0405  0.0878    0.0504
## s.e.  0.1309  0.1302    0.0194
##
## sigma^2 estimated as 0.01636:  log likelihood = 36.34,  aic = -64.67
```

```
log.diff.gdp.ar2.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   xmean
##      0.0405  0.0878  0.0504
## s.e.  0.1309  0.1302  0.0194
##
## sigma^2 estimated as 0.01636:  log likelihood = 36.34,  aic = -64.67
```

```

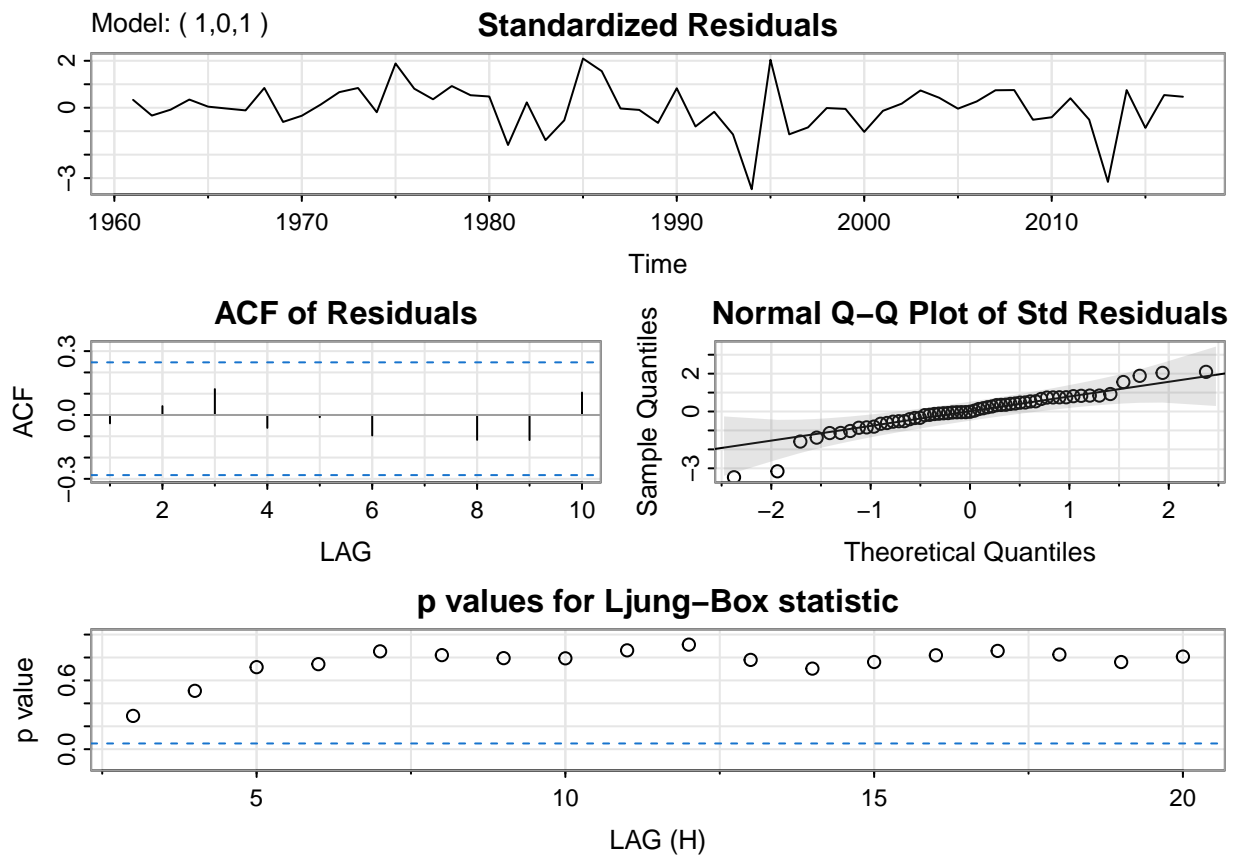
##
## $degrees_of_freedom
## [1] 54
##
## $ttable
##      Estimate      SE t.value p.value
## ar1      0.0405 0.1309  0.3097  0.7580
## ar2      0.0878 0.1302  0.6738  0.5033
## xmean    0.0504 0.0194  2.6026  0.0119
##
## $ICs
##      AIC      AICc      BIC
## -1.134616 -1.126672 -0.991244

# Model 5: ARMA(1,1)
log.diff.gdp.arma11 <- arima(log.diff.gdp.ts, order = c(1,0,1))
log.diff.gdp.arma11.diagnostics <- sarima(log.diff.gdp.ts, 1, 0, 1)

## initial  value -2.043604
## iter    2 value -2.044276
## iter    3 value -2.044596
## iter    4 value -2.044598
## iter    5 value -2.044625
## iter    6 value -2.044786
## iter    7 value -2.044823
## 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
## iter    3 value -2.056124
## 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
## ma1     -0.5615 0.4727 -1.1879  0.2401
## xmean     0.0502 0.0202  2.4854  0.0161
##
## sigma^2 estimated as 0.0163663 on 54 degrees of freedom
##
## AIC = -1.134071  AICc = -1.126126  BIC = -0.9906985
##
```



```
log.diff.gdp.arma11
```

```
##
## Call:
## arima(x = log.diff.gdp.ts, order = c(1, 0, 1))
##
## Coefficients:
##      ar1      ma1  intercept
##      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
```

```
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:
##          ar1          ma1      xmean
##      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 <- arima(log.diff.gdp.ts, order = c(1,0,2))
log.diff.gdp.arma12.diagnostics <- sarima(log.diff.gdp.ts, 1, 0, 2)

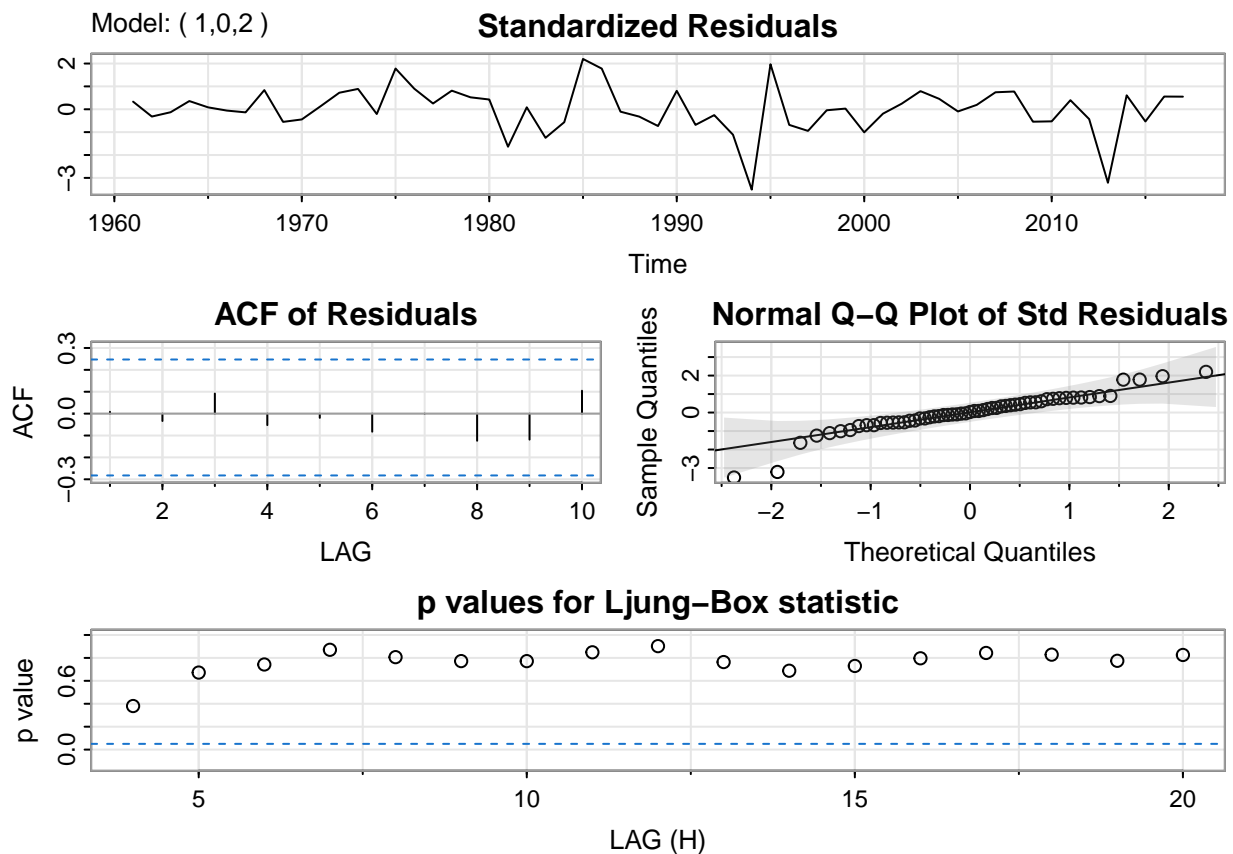
```

```

## initial  value -2.043604
## iter    2 value -2.046177
## iter    3 value -2.048533
## iter    4 value -2.048624
## iter    5 value -2.048815
## iter    6 value -2.049700
## iter    7 value -2.051473
## iter    8 value -2.051869
## iter    9 value -2.052374
## 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
## iter    7 value -2.061138
## iter    8 value -2.061139
## iter    9 value -2.061139
## 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
## ma1     -0.4623 0.4343 -1.0644  0.2920
## ma2      0.1225 0.1510  0.8116  0.4206
## 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
##
```



```
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
## s.e.  0.4350  0.4343  0.1510    0.0212
##
## 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:
##          ar1          ma1          ma2      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
##      Estimate      SE t.value p.value
## 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
## xmean     0.0505 0.0212  2.3799  0.0209
##
## $ICs
##      AIC      AICc      BIC
## -1.1089624 -1.0954671 -0.9297474
```

```
# Model 7: ARMA(2,1)
```

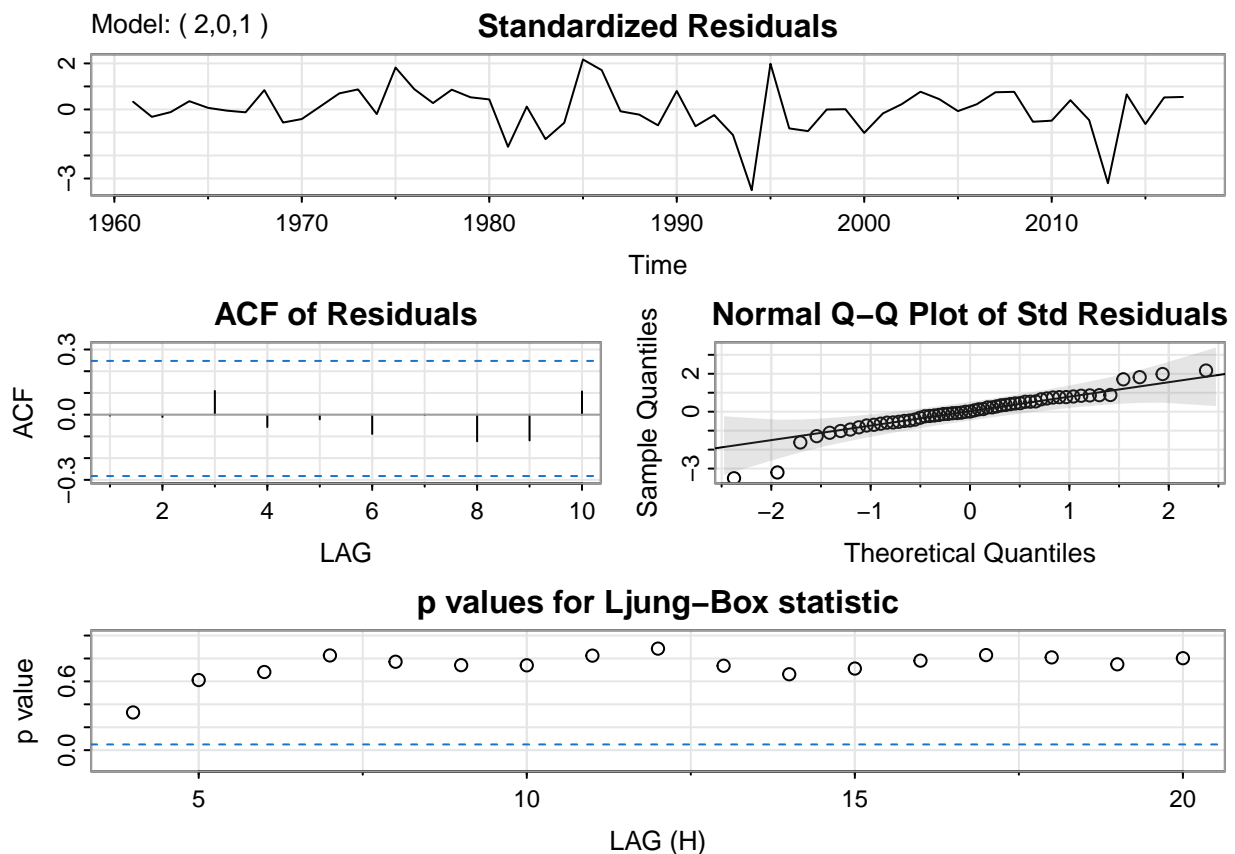
```
log.diff.gdp.arma21 <- arima(log.diff.gdp.ts, order = c(2,0,1))
log.diff.gdp.arma21.diagnostics <- sarima(log.diff.gdp.ts, 2, 0, 1)
```

```
## 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
## iter 9 value -2.043499
## 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
## iter 3 value -2.059466
## iter 4 value -2.059467
## iter 5 value -2.059470
## iter 6 value -2.059475
## iter 7 value -2.059479
## iter 8 value -2.059480
## iter 9 value -2.059480
## iter 9 value -2.059480
## iter 9 value -2.059480
## final value -2.059480
## converged
## <><><><><><><><><><><><><>
##
## Coefficients:
##      Estimate      SE t.value p.value
## ar1      0.4057 0.5161  0.7862  0.4353
## ar2      0.0918 0.1378  0.6661  0.5082
## ma1     -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
##

```



```
log.diff.gdp.arma21
```

```
##
## Call:
## arima(x = log.diff.gdp.ts, order = c(2, 0, 1))
##
## Coefficients:
##          ar1      ar2      ma1  intercept
##          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
```

```
log.diff.gdp.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
##          0.4057  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
##      Estimate      SE t.value p.value
## ar1      0.4057 0.5161  0.7862  0.4353
## ar2      0.0918 0.1378  0.6661  0.5082
## ma1     -0.3718 0.5064 -0.7343  0.4660
## xmean     0.0504 0.0209  2.4092  0.0195
##
## $ICs
##      AIC      AICc      BIC
## -1.1056435 -1.0921482 -0.9264285
```

```
# Model 8: ARMA(2,2) (final)
```

```
log.diff.gdp.arma22 <- arima(log.diff.gdp.ts, order = c(2,0,2))
log.diff.gdp.arma22.diagnostics <- sarima(log.diff.gdp.ts, 2, 0, 2)
```

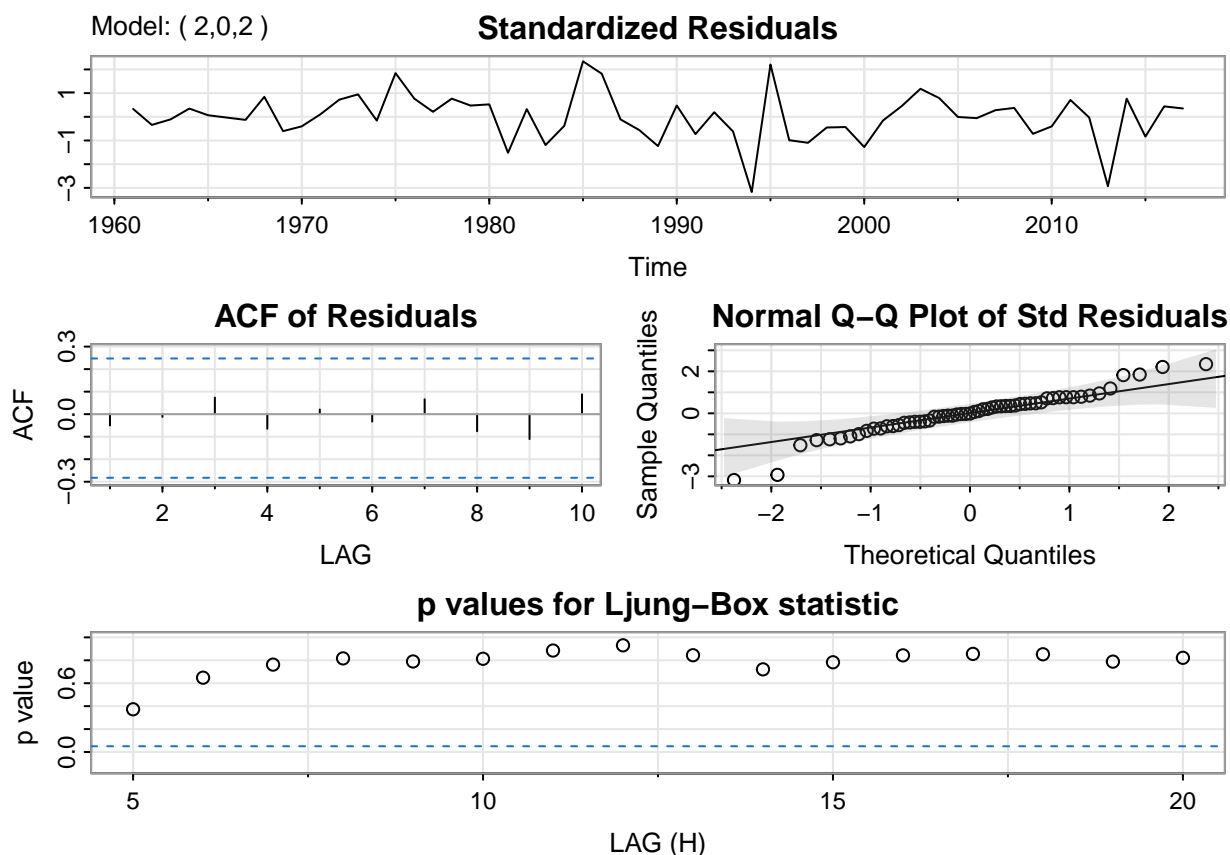
```
## initial  value -2.035413
## iter    2 value -2.038802
## iter    3 value -2.040478
## iter    4 value -2.040641
## iter    5 value -2.040791
## iter    6 value -2.042896
## iter    7 value -2.044919
## iter    8 value -2.044997
```

```
## 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
## iter   16 value -2.047784
## 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
## iter   22 value -2.048189
## iter   23 value -2.048265
## iter   24 value -2.048319
## iter   25 value -2.048351
## iter   26 value -2.048358
## iter   27 value -2.048491
## 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
## iter   36 value -2.059606
## 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
## iter   42 value -2.097030
## 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
## iter   49 value -2.101215
## 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
## s.e.  0.1236  0.0924  0.1006  0.1    0.0196
##
## 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
##      1.5232 -0.9044 -1.5297  1.0  0.0508
## s.e.  0.1236  0.0924  0.1006  0.1  0.0196
##
## sigma^2 estimated as 0.01444:  log likelihood = 38.44,  aic = -64.87
```



```
##
## $degrees_of_freedom
## [1] 52
##
## $ttable
##      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
##
## $ICs
##      AIC      AICc      BIC
## -1.1381564 -1.1175166 -0.9230984
```

## GDP Final Model: ARMA(2,2)

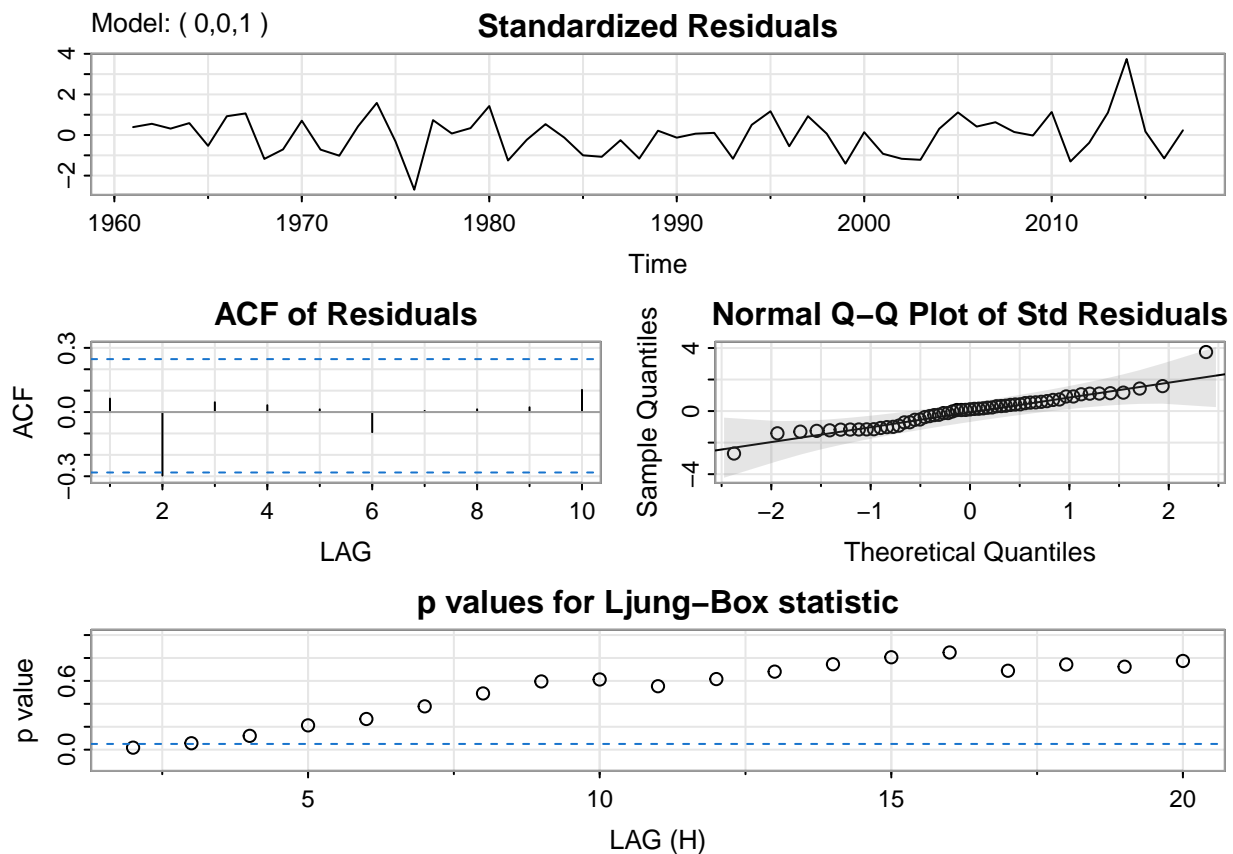
### Imports

```
# Model Selection on Log-Differenced Imports (stationary)
# Imports

# Model 1: MA(1)
log.diff.import.ma1 <- arima(log.diff.import.ts, order = c(0,0,1))
log.diff.import.ma1.diagnostics <- sarima(log.diff.import.ts, 0, 0, 1)
```

```
## 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
## iter 9 value -2.144361
## iter 9 value -2.144361
## iter 9 value -2.144361
## final value -2.144361
## converged
## initial value -2.143999
## iter 2 value -2.144005
## iter 3 value -2.144005
## iter 4 value -2.144007
## iter 4 value -2.144007
## iter 4 value -2.144007
## final value -2.144007
## converged
## <><><><><><><><><><><><><><>
##
## Coefficients:
##      Estimate      SE t.value p.value
## ma1     -0.2314 0.1836 -1.2604 0.2128
## 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
##
```



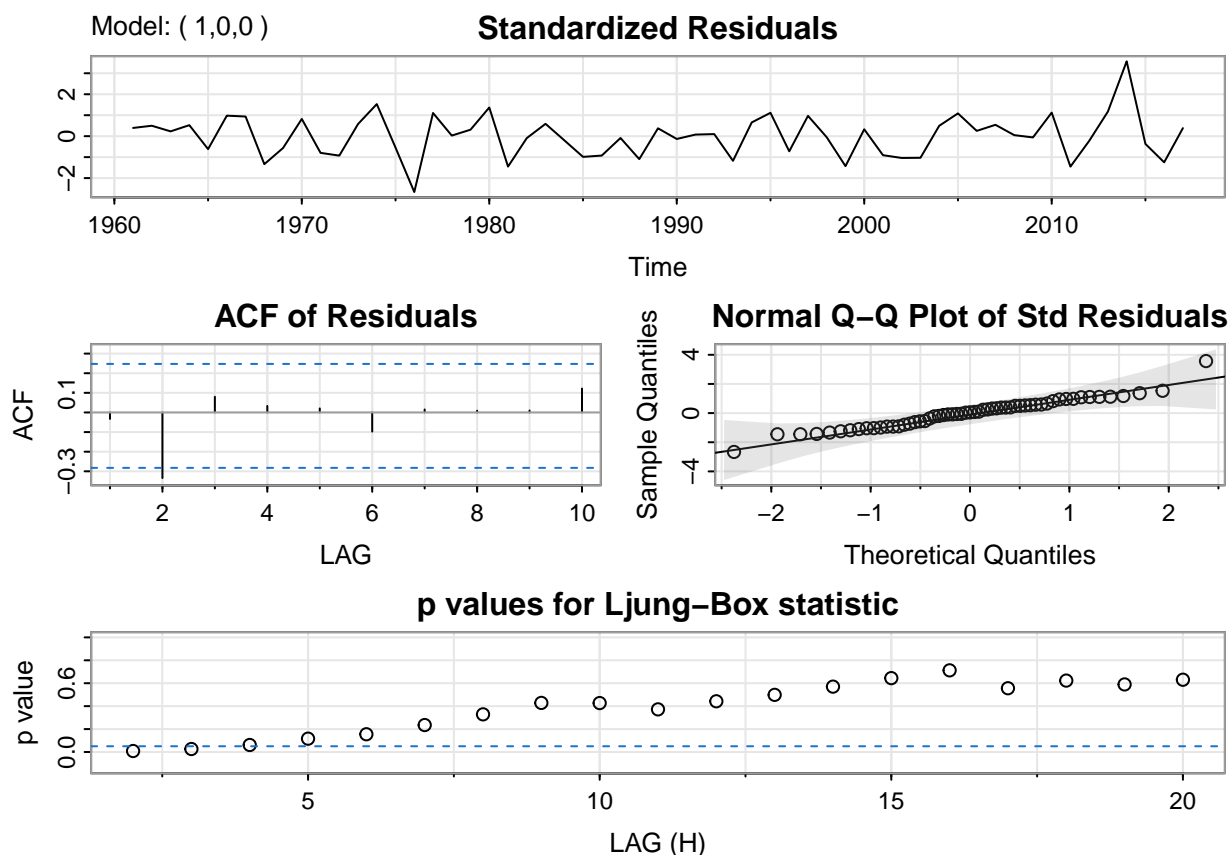
```
log.diff.import.ma1
```

```
##
## 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
##
## 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))
##
```





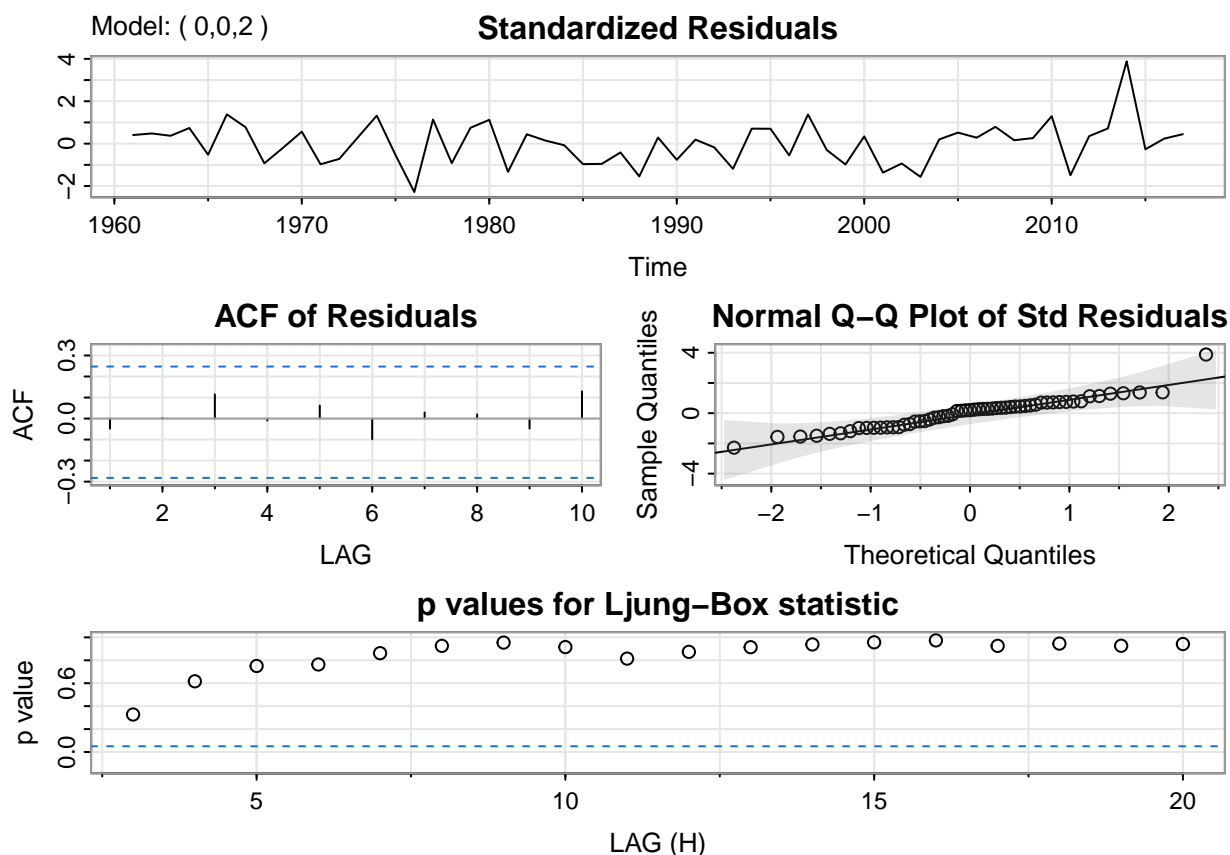
```
log.diff.import.ar1
```

```
##
## Call:
## arima(x = log.diff.import.ts, order = c(1, 0, 0))
##
## Coefficients:
##          ar1  intercept
##      -0.0926  -0.0014
## s.e.   0.1312   0.0143
##
## sigma^2 estimated as 0.01393:  log likelihood = 40.91,  aic = -75.82
```

```
log.diff.import.ar1.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    xmean
##      -0.0926  -0.0014
## s.e.   0.1312   0.0143
##
## sigma^2 estimated as 0.01393:  log likelihood = 40.91,  aic = -75.82
```





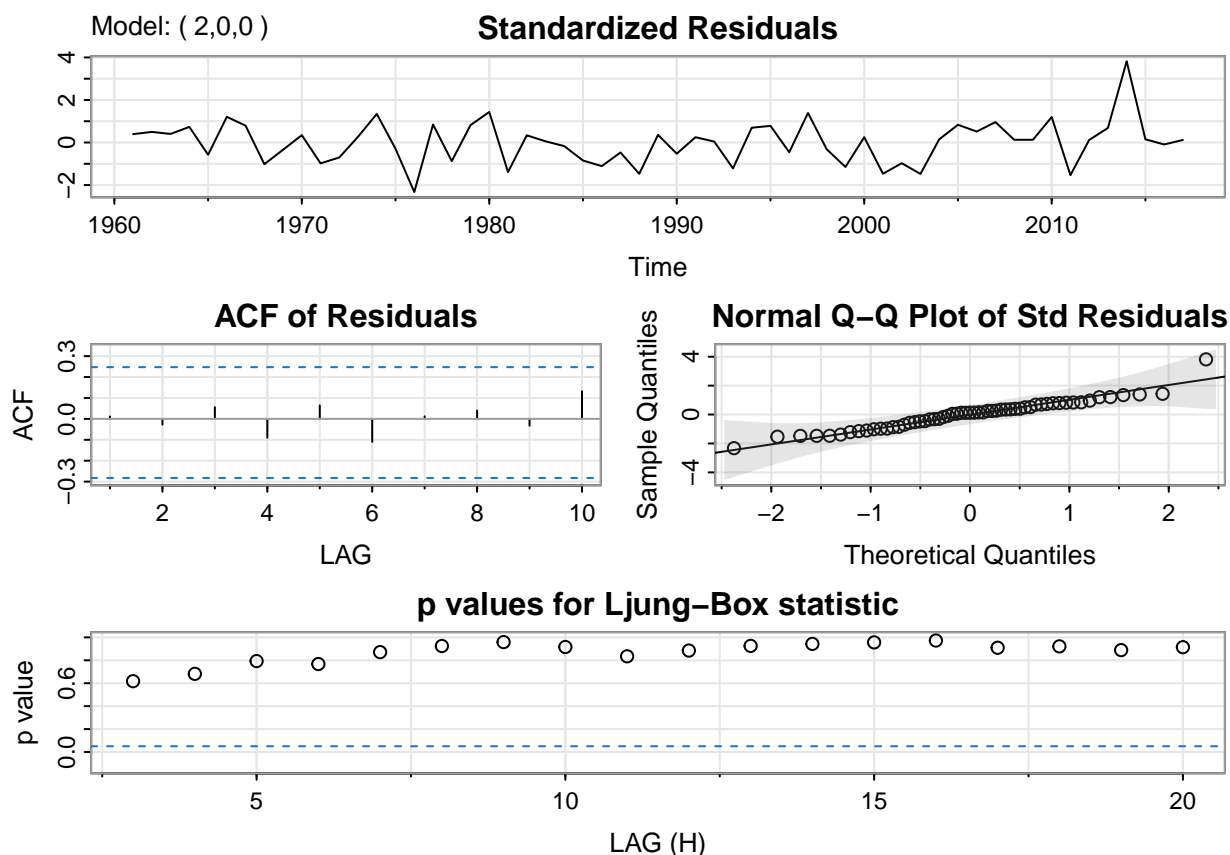
```
log.diff.import.ma2
```

```
##
## Call:
## arima(x = log.diff.import.ts, order = c(0, 0, 2))
##
## Coefficients:
##          ma1          ma2  intercept
##      -0.0465  -0.3865   -0.0029
## s.e.   0.1315   0.1414    0.0086
##
## sigma^2 estimated as 0.01219:  log likelihood = 44.57,  aic = -81.13
```

```
log.diff.import.ma2.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
##      -0.0465  -0.3865   -0.0029
## s.e.   0.1315   0.1414    0.0086
##
## sigma^2 estimated as 0.01219:  log likelihood = 44.57,  aic = -81.13
```





```
log.diff.import.ar2
```

```
##
## Call:
## arima(x = log.diff.import.ts, order = c(2, 0, 0))
##
## Coefficients:
##          ar1      ar2  intercept
##      -0.1229 -0.3417   -0.0015
## s.e.   0.1240   0.1236    0.0101
##
## sigma^2 estimated as 0.01225:  log likelihood = 44.46,  aic = -80.92
```

```
log.diff.import.ar2.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    xmean
##      -0.1229 -0.3417  -0.0015
## s.e.   0.1240   0.1236   0.0101
##
## 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
## ar1      -0.1229 0.1240 -0.9912 0.3260
## ar2      -0.3417 0.1236 -2.7637 0.0078
## 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 <- arima(log.diff.import.ts, order = c(1,0,1))
log.diff.import.arma11.diagnostics <- sarima(log.diff.import.ts, 1, 0, 1)

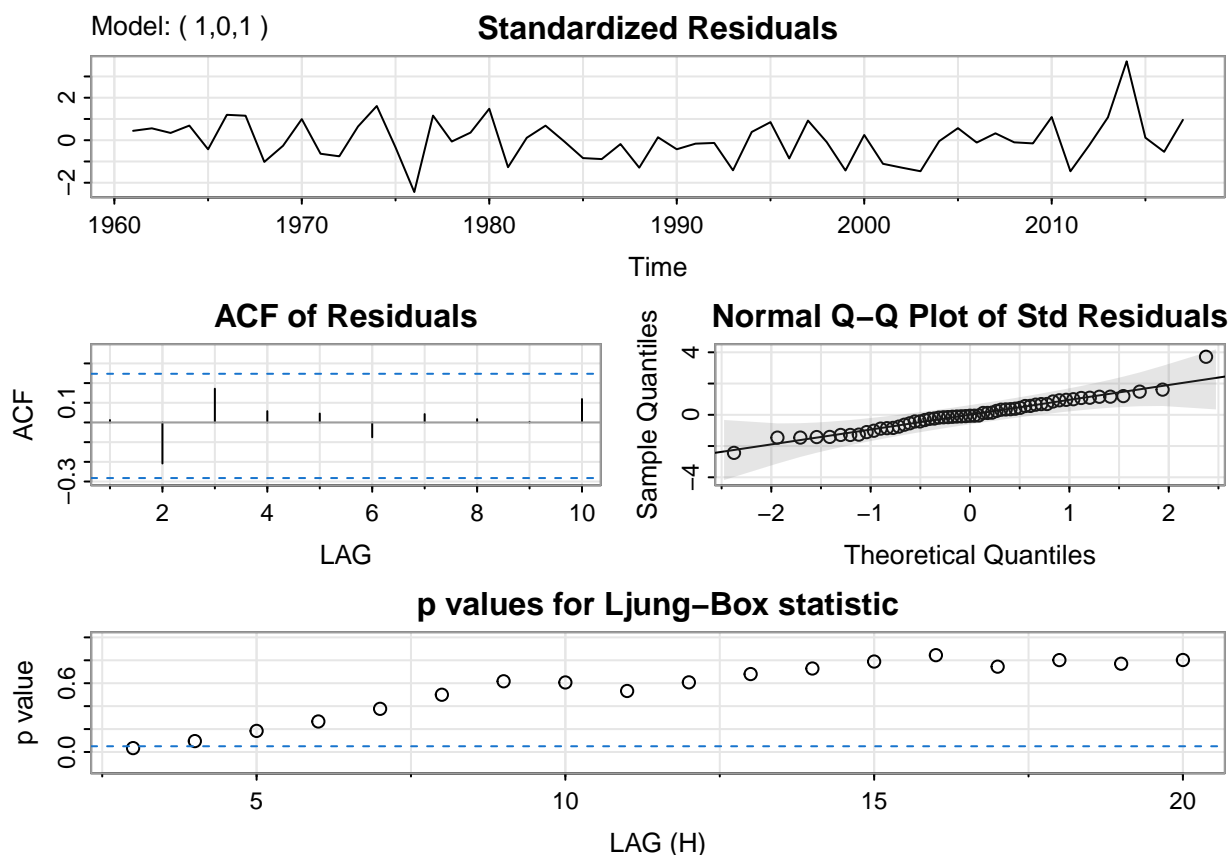
## initial  value -2.124853
## iter    2 value -2.131198
## iter    3 value -2.133231
## iter    4 value -2.137431
## iter    5 value -2.149287
## iter    6 value -2.160505
## iter    7 value -2.163489
## iter    8 value -2.164322
## iter    9 value -2.164496
## iter   10 value -2.164541
## iter   11 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
## iter   19 value -2.167436
## iter   20 value -2.176018
## iter   21 value -2.178987
## iter   22 value -2.192394
## 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
##

```



```
log.diff.import.arma11
```

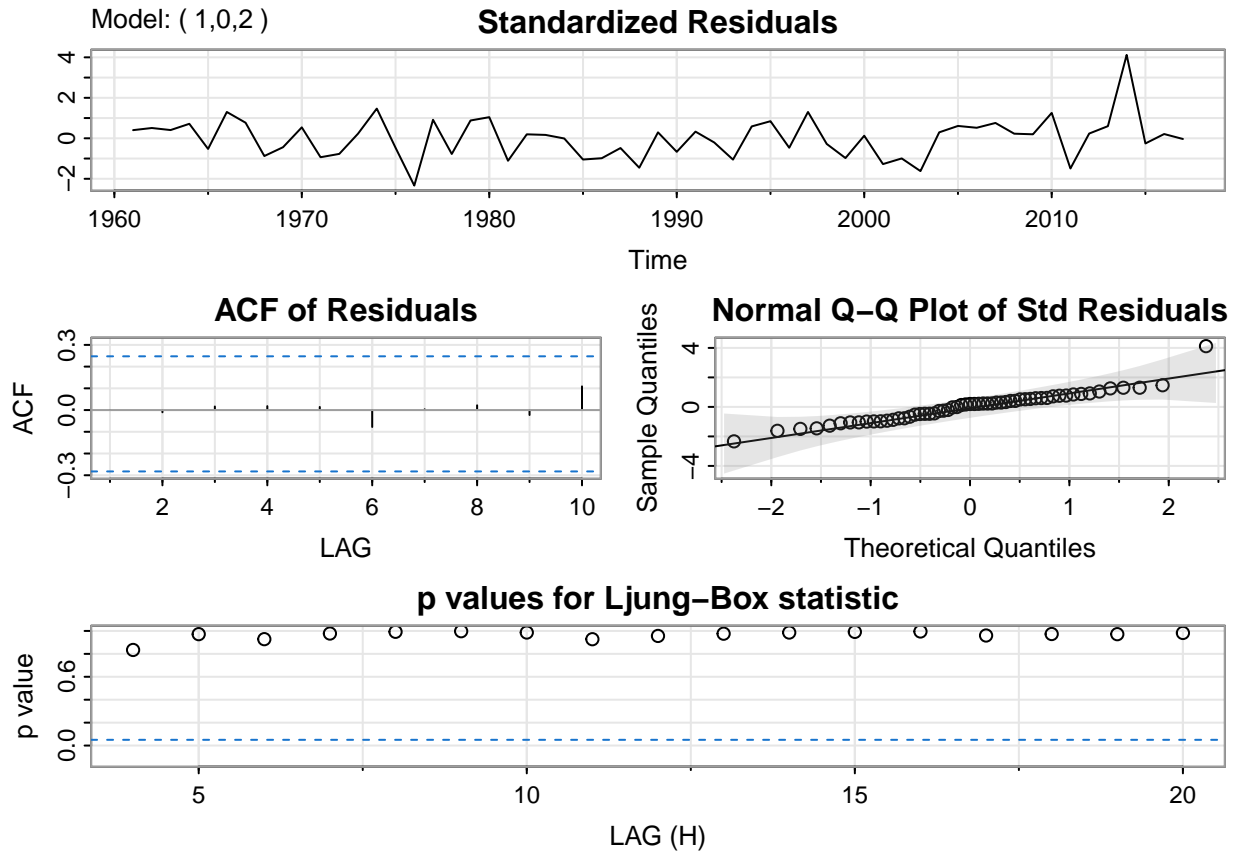
```
##
## Call:
## arima(x = log.diff.import.ts, order = c(1, 0, 1))
##
## Coefficients:
##          ar1          ma1  intercept
##          0.7838      -1.0000      -0.0075
## s.e.  0.0991    0.0583    0.0036
##
## sigma^2 estimated as 0.01261:  log likelihood = 42.73,  aic = -77.45
```

```
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
## s.e.  0.0991    0.0583    0.0036
##
## sigma^2 estimated as 0.01261:  log likelihood = 42.73,  aic = -77.45
```



```
## 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
##
```



```
log.diff.import.arma12
```

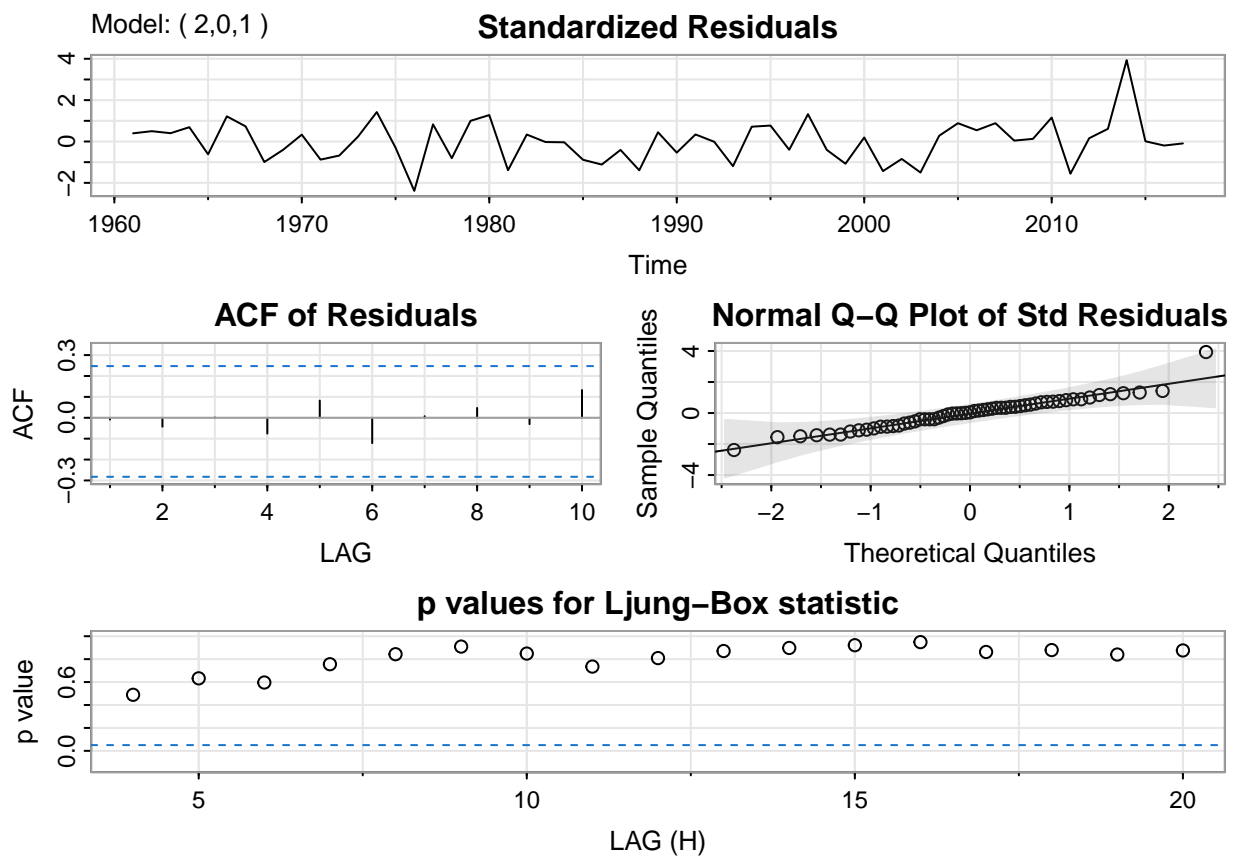
```
##
## Call:
## arima(x = log.diff.import.ts, order = c(1, 0, 2))
##
## Coefficients:
##      ar1      ma1      ma2 intercept
##    -0.3312  0.2433 -0.4033   -0.0025
## s.e.    0.2902  0.2720  0.1267    0.0093
##
## sigma^2 estimated as 0.01195: log likelihood = 45.11, aic = -80.21
```

```
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:
##      Estimate      SE t.value p.value
## 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
##
## sigma^2 estimated as 0.01221868 on 53 degrees of freedom
##
## AIC = -1.386832  AICc = -1.373337  BIC = -1.207617
##
```



```
log.diff.import.arma21
```

```
##
## Call:
## arima(x = log.diff.import.ts, order = c(2, 0, 1))
##
## Coefficients:
##      ar1      ar2      ma1  intercept
##    -0.3207 -0.3531  0.2253   -0.0016
## 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
##      -0.3207  -0.3531   0.2253  -0.0016
## s.e.    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
##      Estimate      SE t.value p.value
## 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))
log.diff.import.arma22.diagnostics <- sarima(log.diff.import.ts, 2, 0, 2)
```

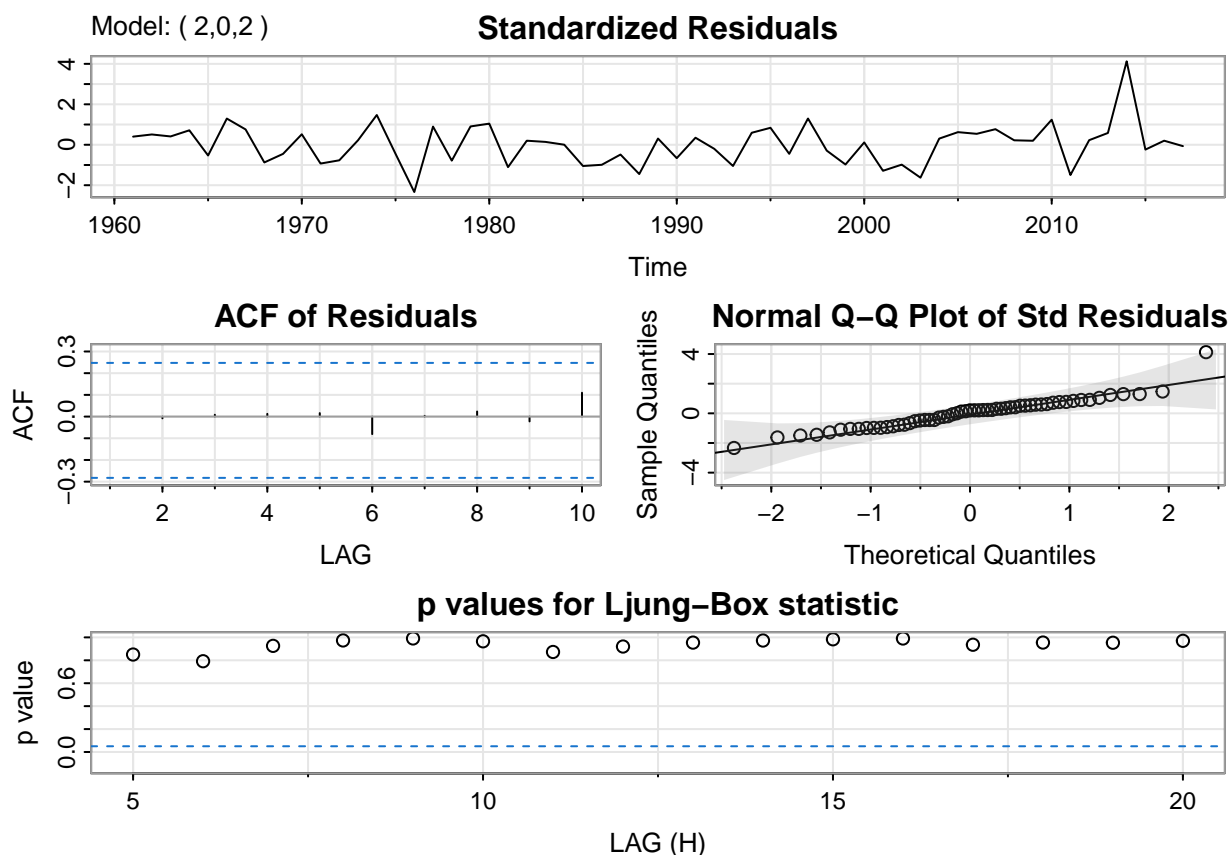
```
## initial value -2.117685
## iter 2 value -2.146070
## iter 3 value -2.189558
## iter 4 value -2.191196
## iter 5 value -2.191658
## iter 6 value -2.194157
## iter 7 value -2.196115
## iter 8 value -2.199075
## iter 9 value -2.200842
## iter 10 value -2.202596
## iter 11 value -2.203318
## iter 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
## 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
##
## sigma^2 estimated as 0.0119451 on 52 degrees of freedom
##
## AIC = -1.372287 AICc = -1.351647 BIC = -1.157229
##

```



```
log.diff.import.arma22
```

```
##
## Call:
## arima(x = log.diff.import.ts, order = c(2, 0, 2))
##
## Coefficients:
##      ar1      ar2      ma1      ma2  intercept
##    -0.3480 -0.0325  0.2586 -0.3743  -0.0024
## s.e.   0.3419   0.3434  0.3225   0.3343   0.0095
##
## 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
##    -0.3480 -0.0325  0.2586 -0.3743 -0.0024
## s.e.   0.3419   0.3434  0.3225   0.3343  0.0095
##
## 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
## ar2    -0.0325 0.3434 -0.0947 0.9249
## 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)

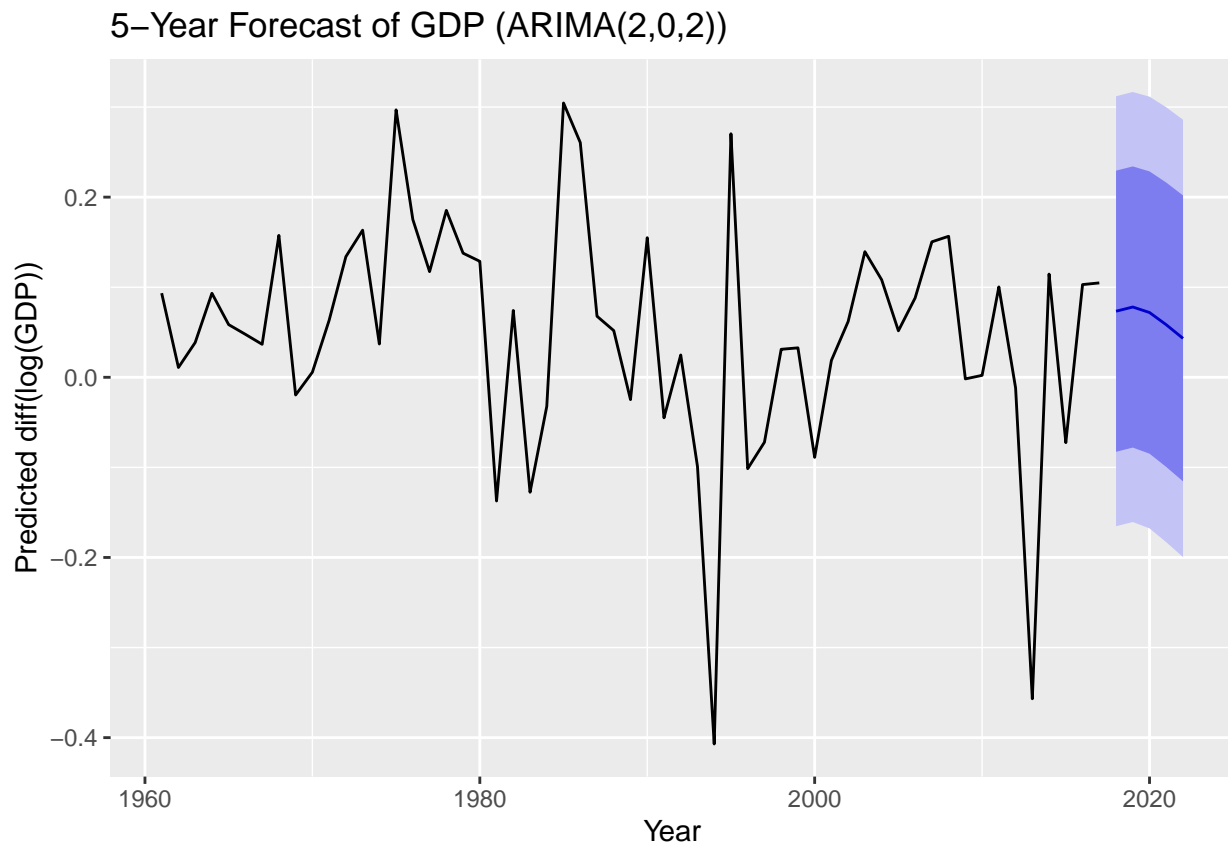
## Forecast

## GDP

```
# Forecast
# GDP
gdp_forecast <- forecast(log.diff.gdp.arma22, h = 5)
gdp_forecast
```

```
##      Point Forecast      Lo 80      Hi 80      Lo 95      Hi 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
## 2022    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))")
```



## Imports

```
# Imports
import_forecast <- forecast(log.diff.import.arma11, h = 5)
import_forecast
```

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 2018	-0.06390127	-0.2089035	0.08110099	-0.2856631	0.1578605
## 2019	-0.05171036	-0.1995994	0.09617871	-0.2778872	0.1744665
## 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))")
```

