

# STAT 5550 project

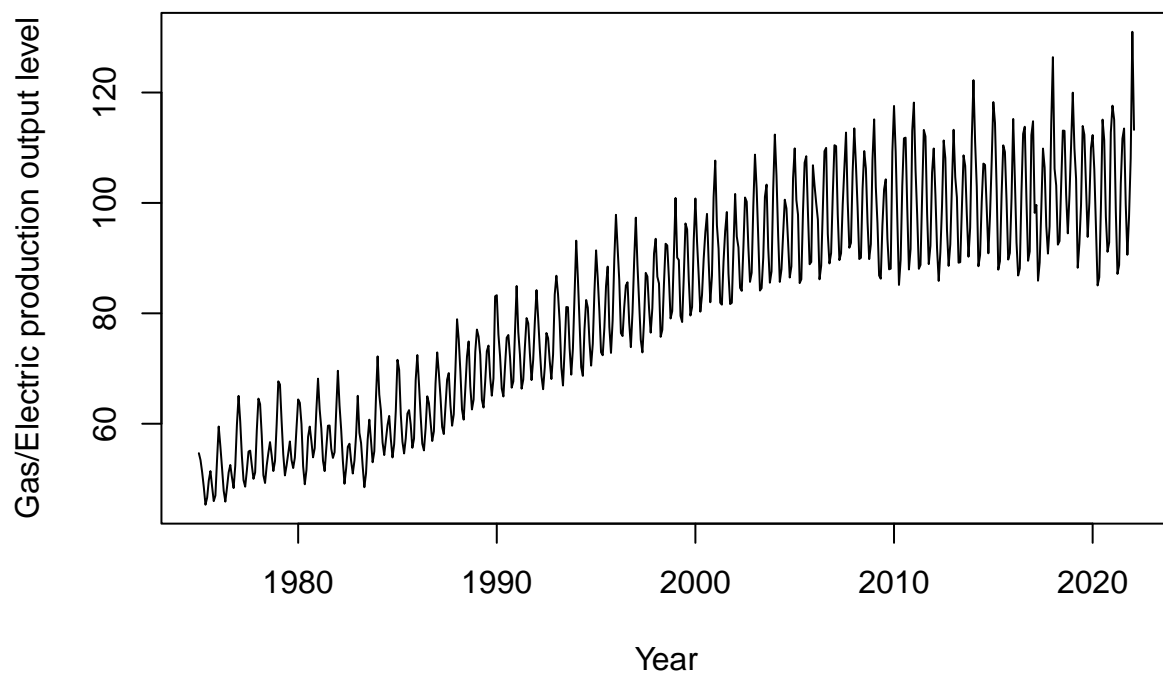
James Laughead

4/13/2022

```
project.data <- read.csv("IPG2211A2N.csv")
df <- (project.data$IPG2211A2N)
df <- df[433:998]

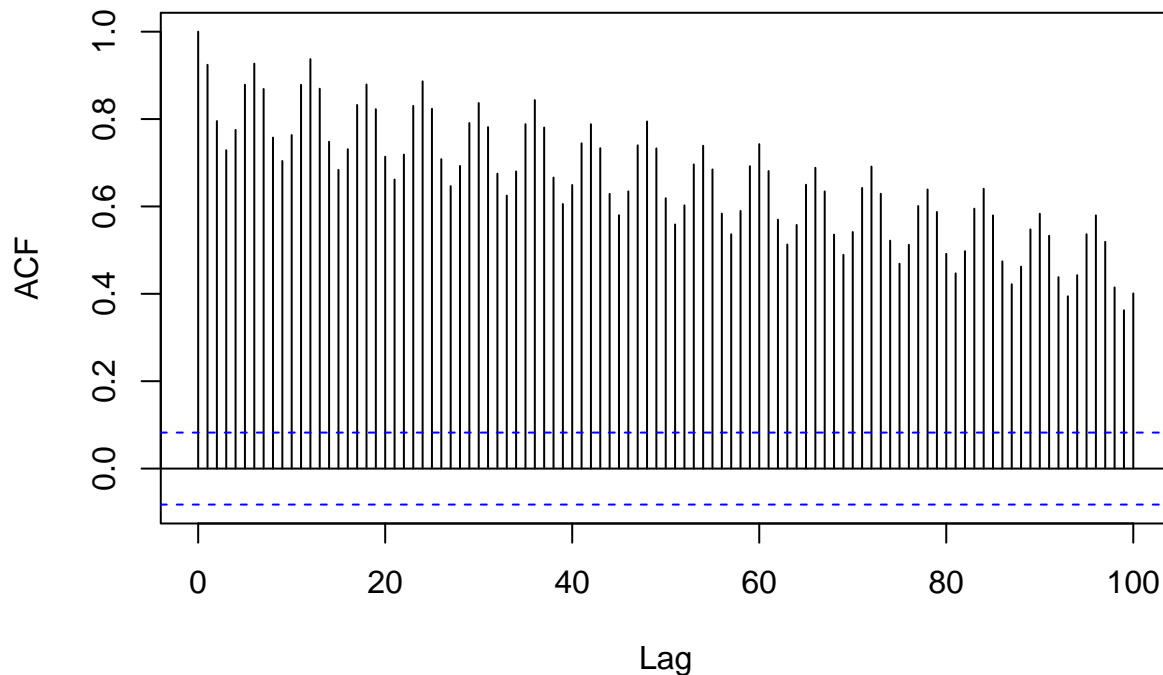
df.ts = ts(df, frequency = 12, start=c(1975, 1))
plot.ts(df.ts, xlab = "Year", ylab = "Gas/Electric production output level", main = "USA utility produ
```

## USA utility production output levels, Jan 1975 – Feb 2022



```
acf(df, lag.max = 100, main = "ACF plot of Gas/Electric production output level data")
```

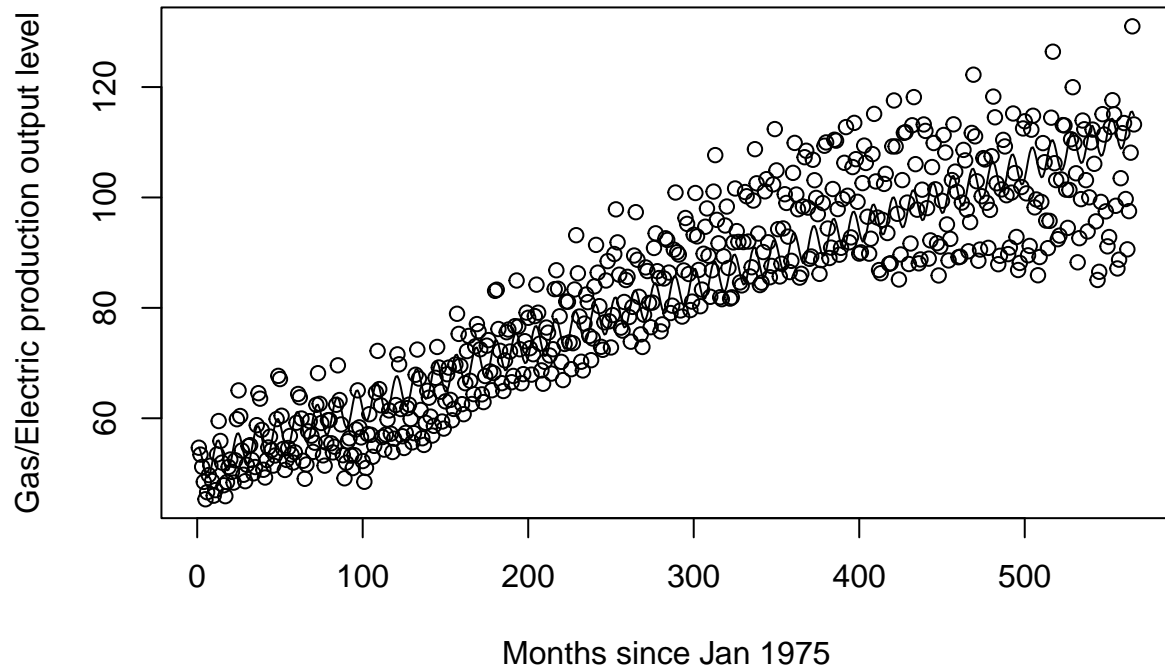
## ACF plot of Gas/Electric production output level data



```
t = 1:length(df)
c = cos(2*pi*t/12)
s = sin(2*pi*t/12)
mod3 = lm(df.ts ~ t+c+s)
summary(mod3)
```

```
##
## Call:
## lm(formula = df.ts ~ t + c + s)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -24.436  -5.351  -0.671   5.823  20.999
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  51.605520   0.694956  74.257 < 2e-16 ***
## t              0.107923   0.002124  50.814 < 2e-16 ***
## c              2.990434   0.490748   6.094 2.05e-09 ***
## s              0.770111   0.490748   1.569   0.117
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.255 on 562 degrees of freedom
## Multiple R-squared:  0.8238, Adjusted R-squared:  0.8228
## F-statistic: 875.6 on 3 and 562 DF, p-value: < 2.2e-16
plot(t, df.ts, main = "Estimated trend and seasonal component", xlab = "Months since Jan 1975", ylab = "Gas/Electric production output level data")
lines(mod3$fit)
```

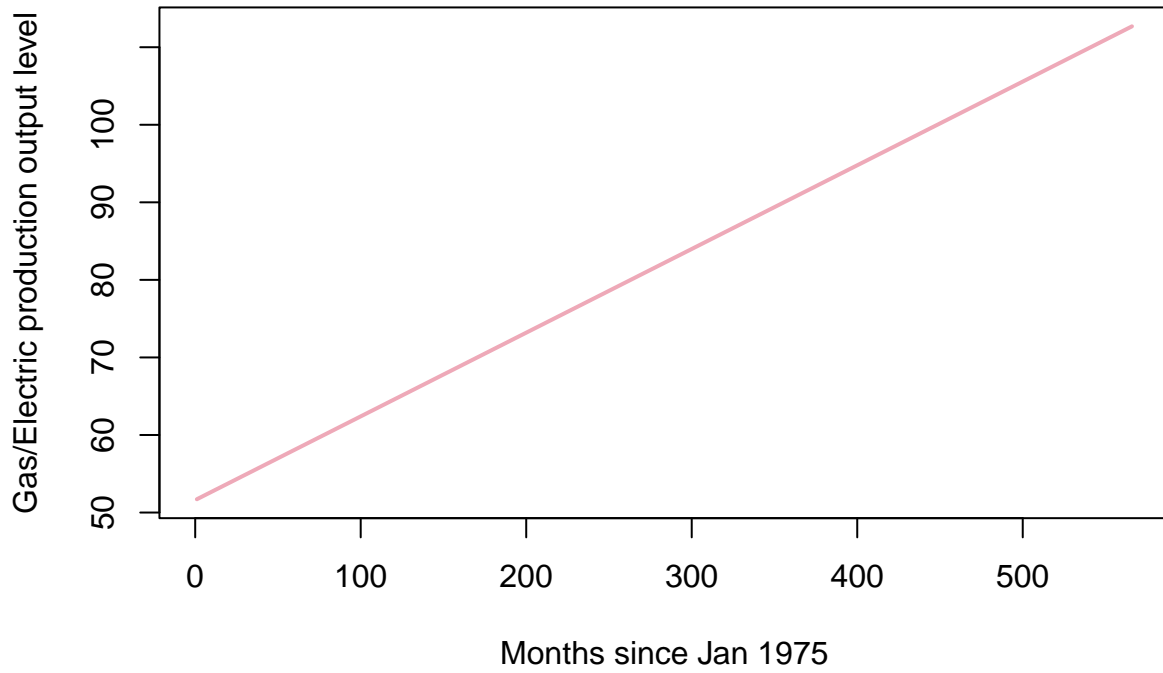
## Estimated trend and seasonal component



```
a <- summary(mod3)$coefficients
ut0 <- a[ 1,1 ]
ut1 <- a[ 2,1 ]
uc1 <- a[ 3,1 ]
uc2 <- a[ 4,1 ]
```

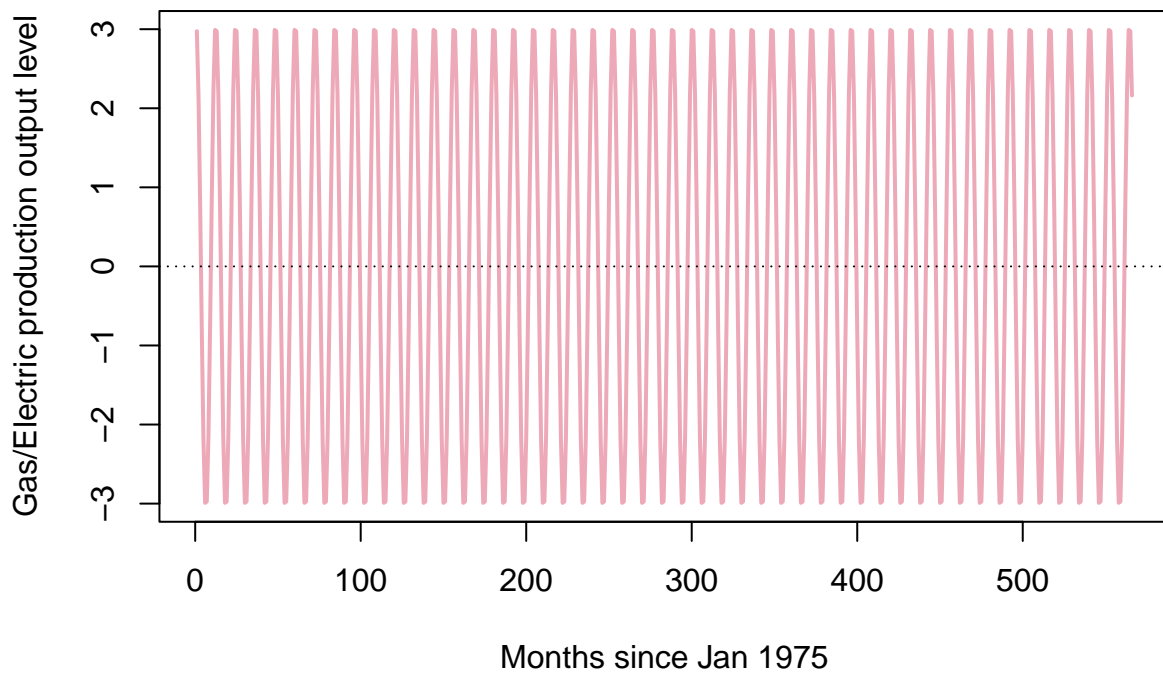
```
plot(time(df), ut0 + (ut1*t), type="l", col="pink2", lwd=2,
     main="Estimated Trend", xlab = "Months since Jan 1975", ylab = "Gas/Electric production output level")
```

## Estimated Trend



```
plot(time(df), (2.990434*(c) + 0.77011133*(s)), type="l", col="pink2", lwd=2,  
      main="Estimated Seasonal Component", xlab = "Months since Jan 1975", ylab = "Gas/Electric production output level",  
      abline(h=0, lty=3))
```

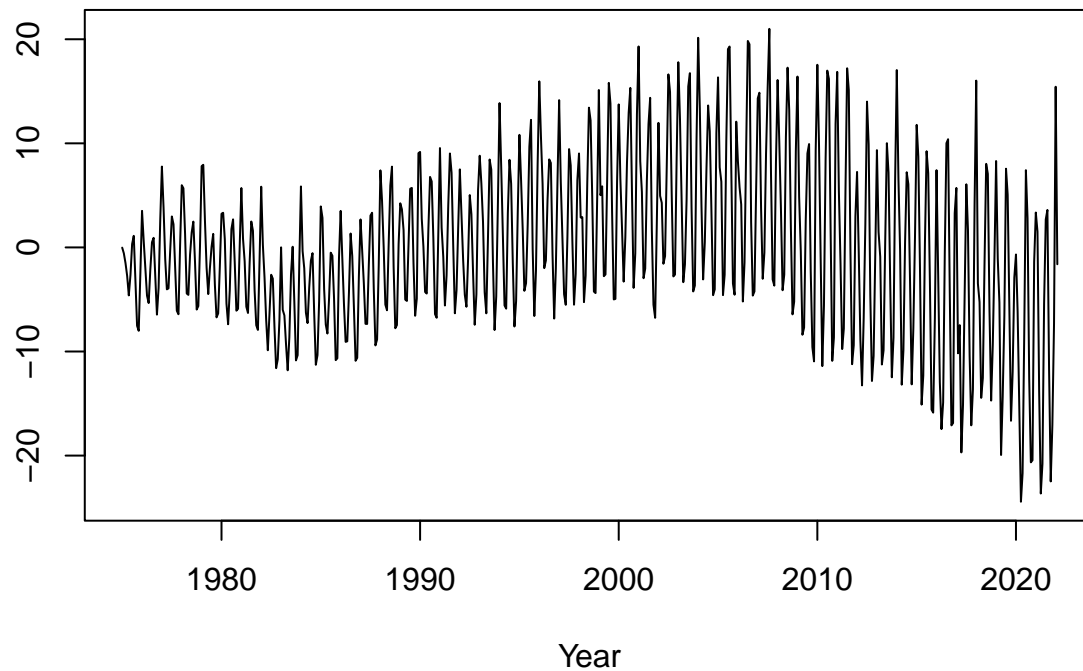
## Estimated Seasonal Component



```
# Make a time series object with the detrended, deseasonalized data:
detrended = df - (ut0 + ut1*t) - ((2.9904347*(c)) + ((0.77011133)*(s)))
detrended = ts(detrended, frequency = 12, start=c(1975, 1))
```

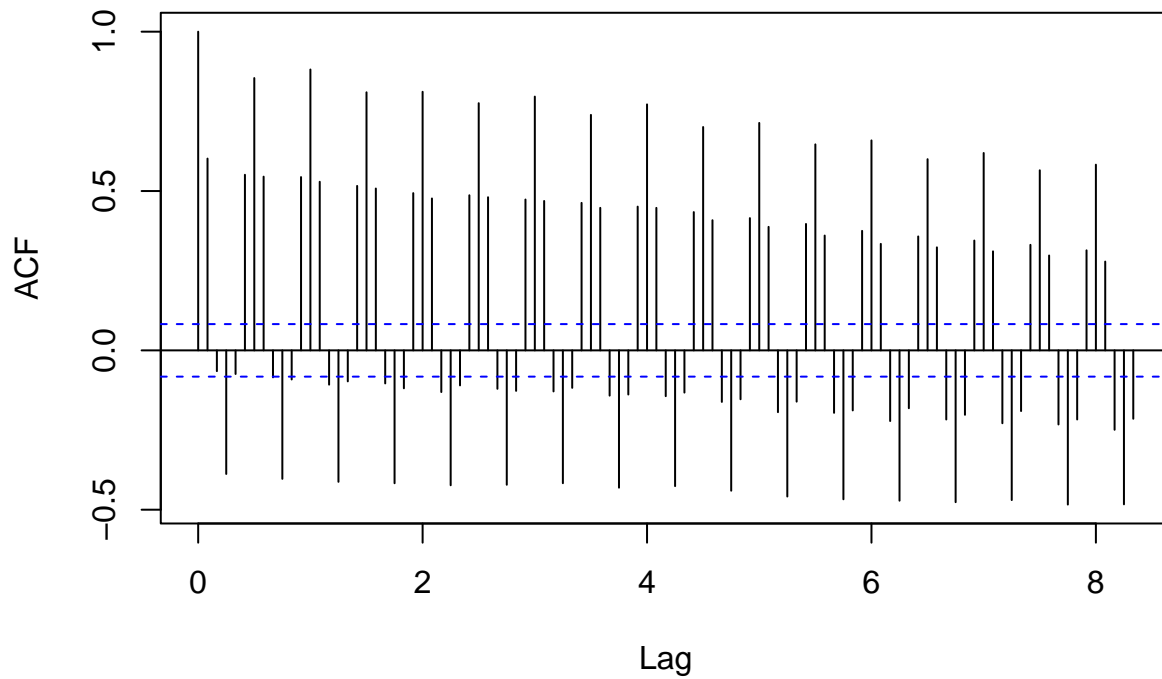
```
plot.ts(detrended, ylab="", main="Detrended, Deseasonalized Data of Gas/Electric production output", xlab="Year")
```

## Detrended, Deseasonalized Data of Gas/Electric production output



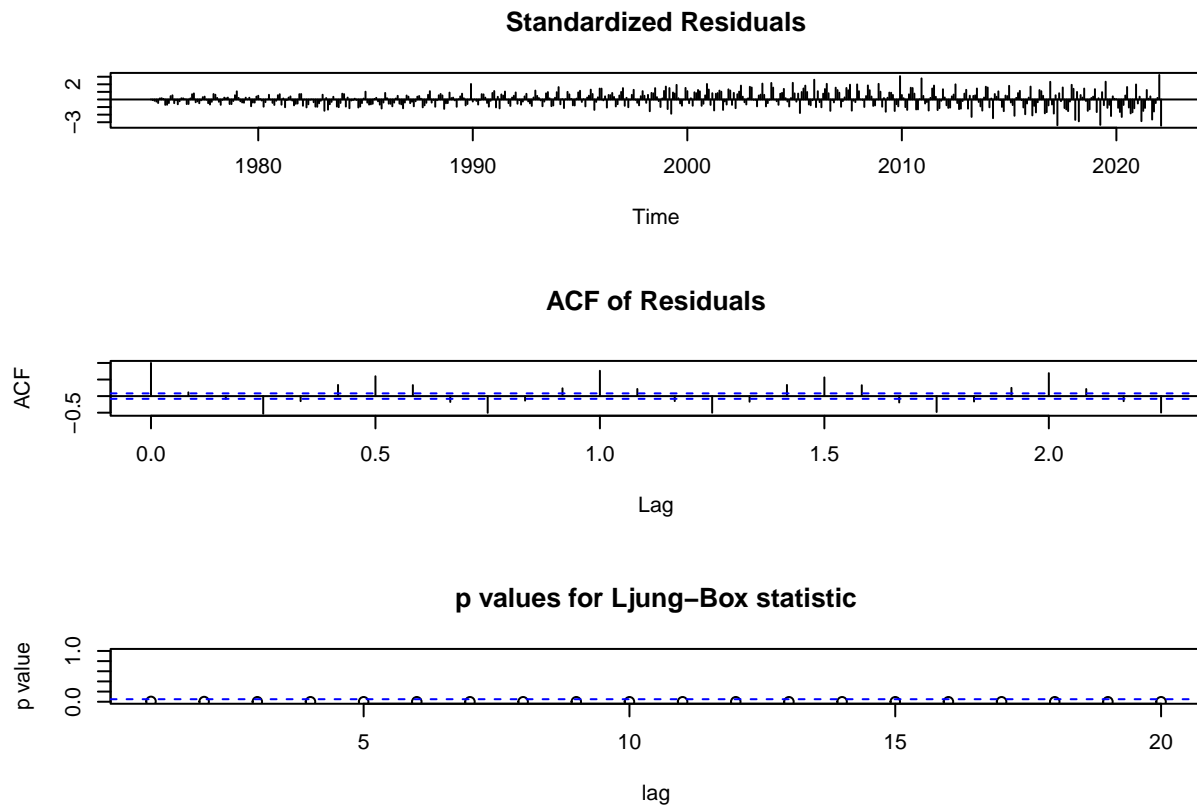
```
acf(detrended, lag.max = 100)
```

## Series detrended



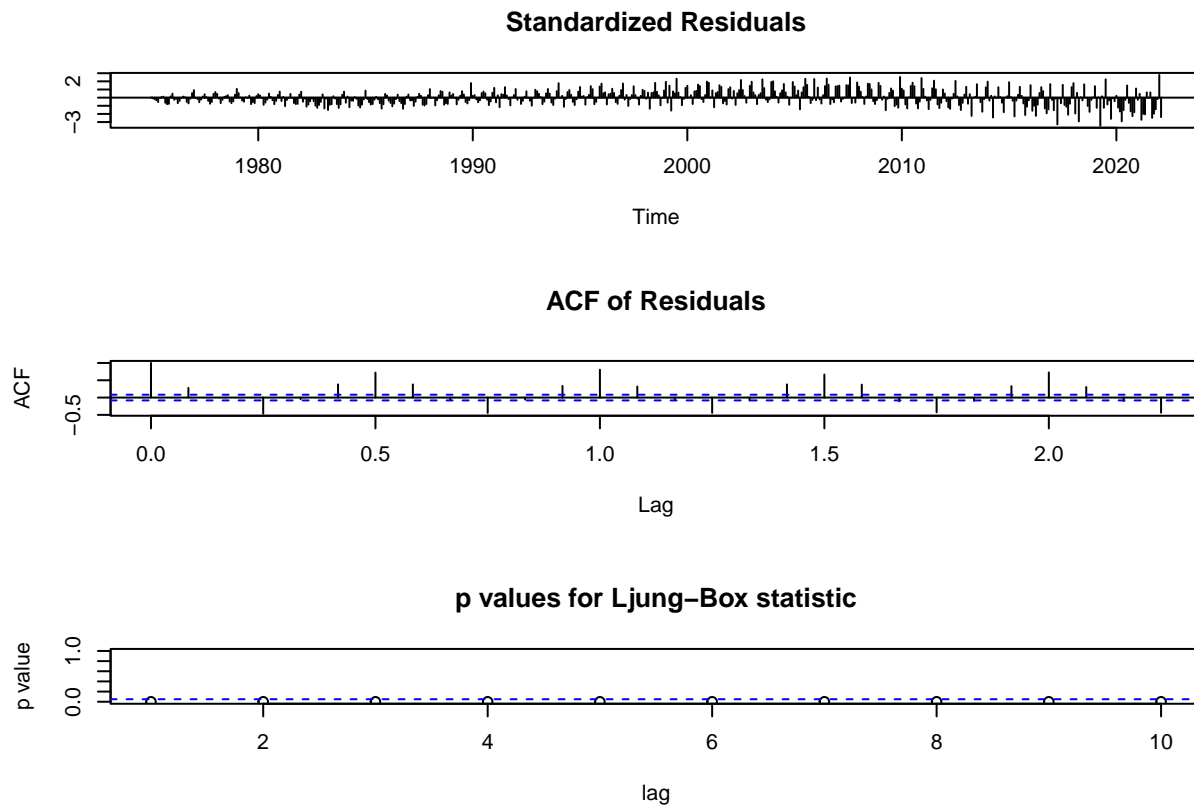
```
varve.arma = arima(detrended, order = c(1, 0, 1))
varve.arma
```

```
##
## Call:
## arima(x = detrended, order = c(1, 0, 1))
##
## Coefficients:
##          ar1      ma1  intercept
##          0.4002  0.6304   -0.0377
## s.e.    0.0435  0.0303    0.6242
##
## sigma^2 estimated as 29.95:  log likelihood = -1765.79,  aic = 3539.58
tsdiag(varve.arma, gof.lag=20)
```



```
varve.arma = arima(detrended, order = c(0, 0, 1))
varve.arma

##
## Call:
## arima(x = detrended, order = c(0, 0, 1))
##
## Coefficients:
##          ma1  intercept
##          0.7583   -0.0198
## s.e.  0.0196    0.4303
##
## sigma^2 estimated as 33.95:  log likelihood = -1801.12,  aic = 3608.24
tsdiag(varve.arma, gof.lag=10)
```



```
varve.arma = arima(detrended, order = c(7, 0, 12))
```

```
## Warning in arima(detrended, order = c(7, 0, 12)): possible convergence problem:
## optim gave code = 1
```

```
varve.arma
```

```
##
```

```
## Call:
```

```
## arima(x = detrended, order = c(7, 0, 12))
```

```
##
```

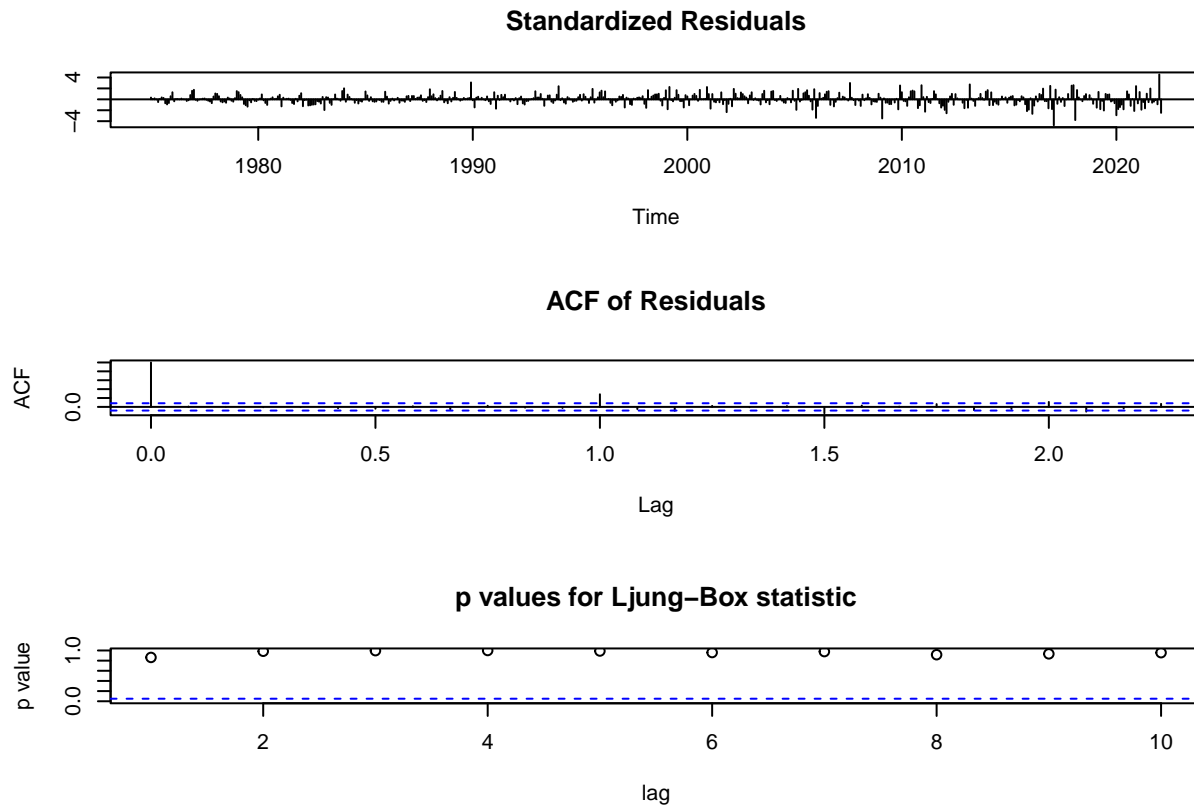
```
## Coefficients:
```

```
## Warning in sqrt(diag(x$var.coef)): NaNs produced
```

```
##      ar1      ar2      ar3      ar4      ar5      ar6      ar7      ma1      ma2
##      0.564 -0.0631 -0.5427  0.5793 -0.0374  0.4309  0.0399 -0.1234  0.0162
## s.e.    NaN      NaN      NaN      NaN      NaN      NaN      NaN      NaN      NaN
##      ma3      ma4      ma5      ma6      ma7      ma8      ma9      ma10
##      0.7135 -0.2863  0.0266 -0.374  -0.1988  0.0555 -0.1785 -0.0052
## s.e.    NaN      NaN      NaN      NaN      NaN      0.0280      NaN      0.0380
##      ma11      ma12      intercept
##      0.0630  0.0473      -2.7035
## s.e.  0.0266      NaN      3.5611
##
## sigma^2 estimated as 7.094:  log likelihood = -1363.1,  aic = 2768.19
```



```
tsdiag(varve.arma, gof.lag=10)
```



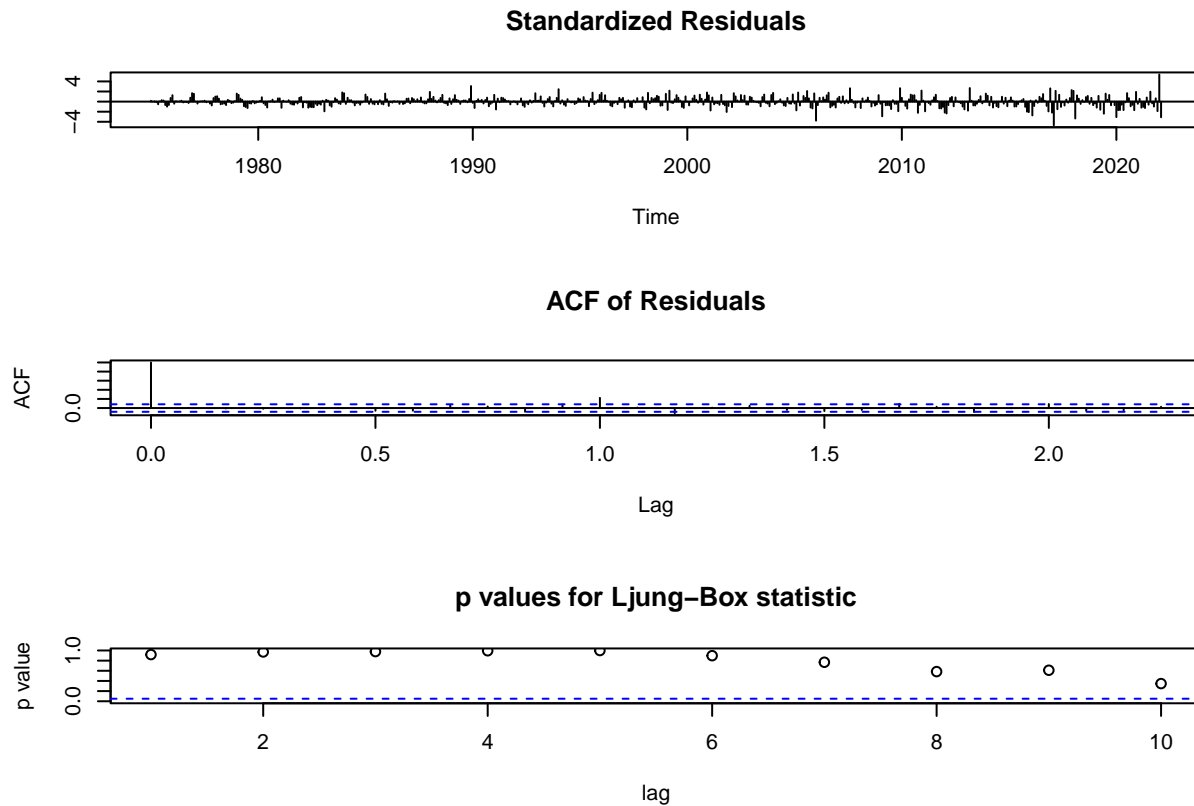
```
varve.arma = arima(detrended, order = c(6, 0, 11))
```

```
## Warning in arima(detrended, order = c(6, 0, 11)): possible convergence problem:
## optim gave code = 1
```

```
varve.arma
```

```
##
## Call:
## arima(x = detrended, order = c(6, 0, 11))
##
## Coefficients:
##      ar1      ar2      ar3      ar4      ar5      ar6      ma1      ma2
##    -0.5157  0.7063 -0.1784 -0.5597  0.7537  0.7750  0.9979 -0.2561
## s.e.   0.1425  0.0492  0.1482  0.1578  0.0449  0.1352  0.1639  0.1288
##      ma3      ma4      ma5      ma6      ma7      ma8      ma9      ma10
##     0.2036  0.9016 -0.3348 -0.9430 -0.1437 -0.0364 -0.3570 -0.0843
## s.e.   0.1179  0.1462  0.1864  0.1247  0.1760  0.1130  0.1162  0.1511
##      ma11 intercept
##     0.0938    -1.7701
## s.e.   0.2049     4.1497
##
## sigma^2 estimated as 6.86:  log likelihood = -1354.74,  aic = 2747.48
```

```
tsdiag(varve.arma, gof.lag=10)
```



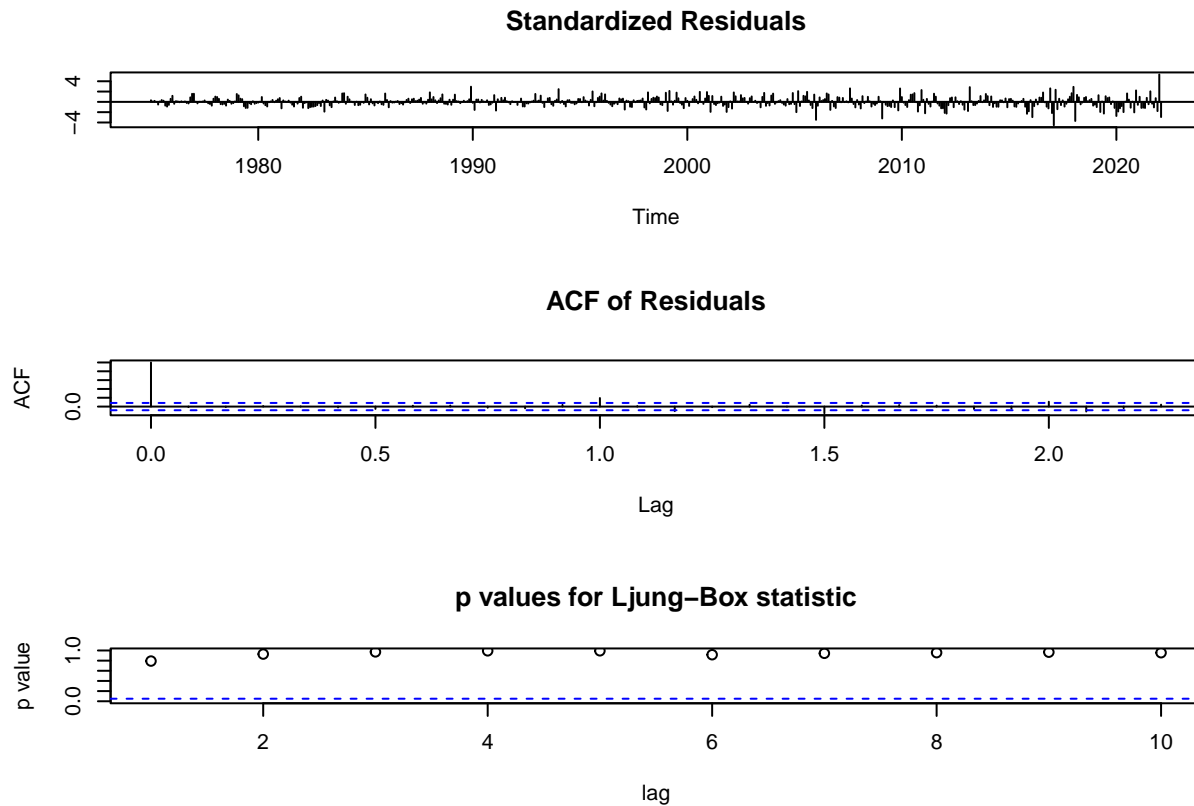
```
varve.arma = arima(detrended, order = c(7, 0, 13))
```

```
## Warning in arima(detrended, order = c(7, 0, 13)): possible convergence problem:
## optim gave code = 1
```

```
varve.arma
```

```
##
## Call:
## arima(x = detrended, order = c(7, 0, 13))
##
## Coefficients:
##      ar1      ar2      ar3      ar4      ar5      ar6      ar7      ma1      ma2
##    -0.6789  0.2744 -0.0565 -0.2584  0.3066  0.9127  0.4488  1.1704  0.2933
## s.e.   0.1170  0.0809  0.0834  0.0988  0.0915  0.0799  0.1317  0.1123  0.1424
##      ma3      ma4      ma5      ma6      ma7      ma8      ma9      ma10
##    0.2877  0.5759  0.1428 -0.7627 -0.7347 -0.2719 -0.1299 -0.0845
## s.e.   0.0928  0.1024  0.1158  0.0958  0.1243  0.0887  0.0838  0.0821
##      ma11     ma12     ma13  intercept
##    -0.1225  0.1434  0.1890    -2.8537
## s.e.   0.0885  0.0733  0.0388     3.5748
##
## sigma^2 estimated as 6.783:  log likelihood = -1350.85,  aic = 2745.69
```

```
tsdiag(varve.arma, gof.lag=10)
```



```
arma(detrended, order = c(1, 0, 1))
```

```
##
## Call:
## arma(x = detrended, order = c(1, 0, 1))
##
## Coefficients:
##      ar1      ma1  intercept
##    0.4002  0.6304   -0.0377
## s.e.  0.0435  0.0303    0.6242
##
## sigma^2 estimated as 29.95:  log likelihood = -1765.79,  aic = 3539.58
```

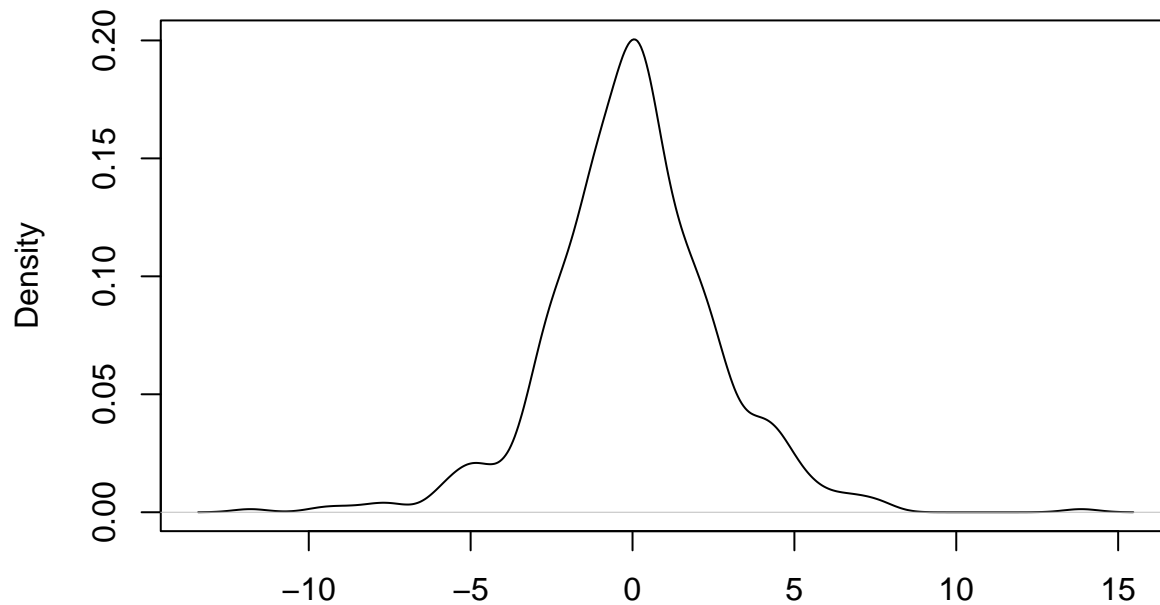
```
a22 <- arma(detrended, order = c(7, 0, 13))
```

```
## Warning in arma(detrended, order = c(7, 0, 13)): possible convergence problem:
## optim gave code = 1
```

```
library(astsa)
```

```
plot(density(a22$residuals), main = "density of Residuals for ARMA(7,13)")
```

### density of Residuals for ARMA(7,13)



N = 566 Bandwidth = 0.5372

```
mod4 = sarima(df, 7,0,13, no.constant=T, details=F)
```

```
## Warning in log(s2): NaNs produced
```

```
## Warning in log(s2): NaNs produced
```

```
## Warning in arima(xdata, order = c(p, d, q), seasonal = list(order = c(P, :  
## possible convergence problem: optim gave code = 1
```

```
## Warning in sqrt(diag(fitit$var.coef)): NaNs produced
```

```
## Warning in sqrt(diag(fitit$var.coef)): NaNs produced
```

```
mod4a = arima(df, order=c(7,0,13), include.mean=F)
```

```
## Warning in log(s2): NaNs produced
```

```
## Warning in log(s2): NaNs produced
```

```
## Warning in arima(df, order = c(7, 0, 13), include.mean = F): possible  
## convergence problem: optim gave code = 1
```

```
mod4.pr = predict(mod4a, n.ahead=2)
```

```
PI.nov = c(mod4.pr$pr[1] - 2*mod4.pr$se[1], mod4.pr$pr[1] + 2*mod4.pr$se[1])  
PI.dec = c(mod4.pr$pr[2] - 2*mod4.pr$se[2], mod4.pr$pr[2] + 2*mod4.pr$se[2])
```

```
mod4.pr = predict(mod4a, n.ahead=45, interval = "pred")
```

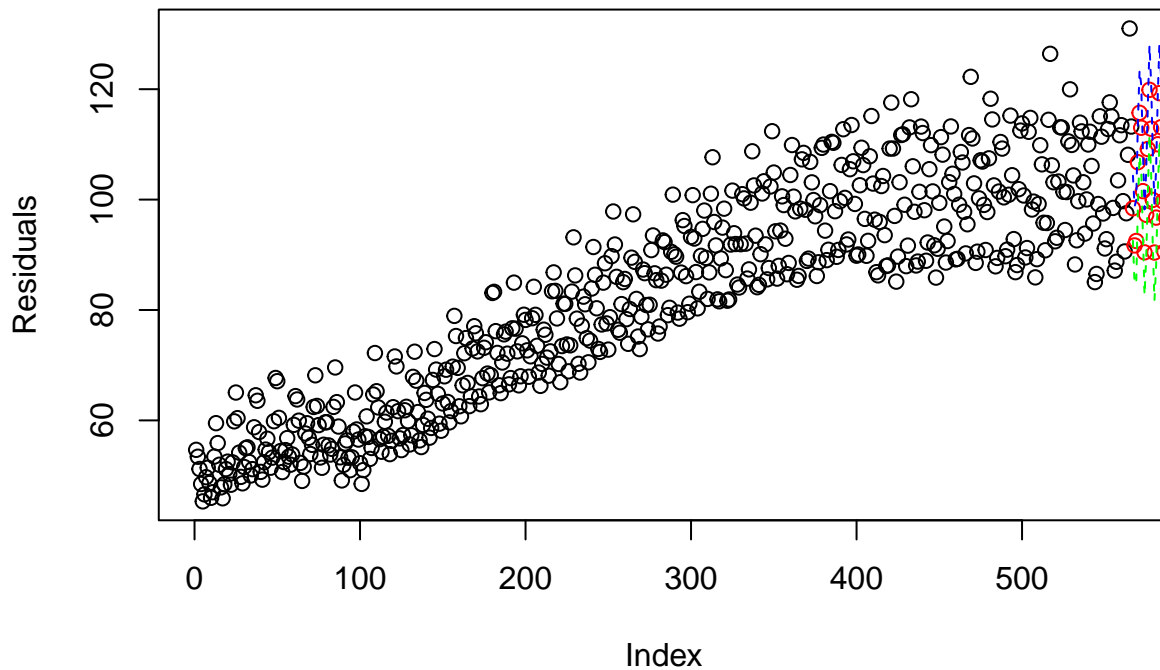
```
zhat = mod4.pr$pr
```

```
pi.z.upper = mod4.pr$pr + 2*mod4.pr$se
```

```
pi.z.lower = mod4.pr$pr - 2*mod4.pr$se
```

```
plot(df, ylab="Residuals", main=expression("Forecasting of USA utility output"))
points(mod4.pr$pred, col="red")
lines(pi.z.upper, lty=2, col="blue")
lines(pi.z.lower, lty=2, col="green")
```

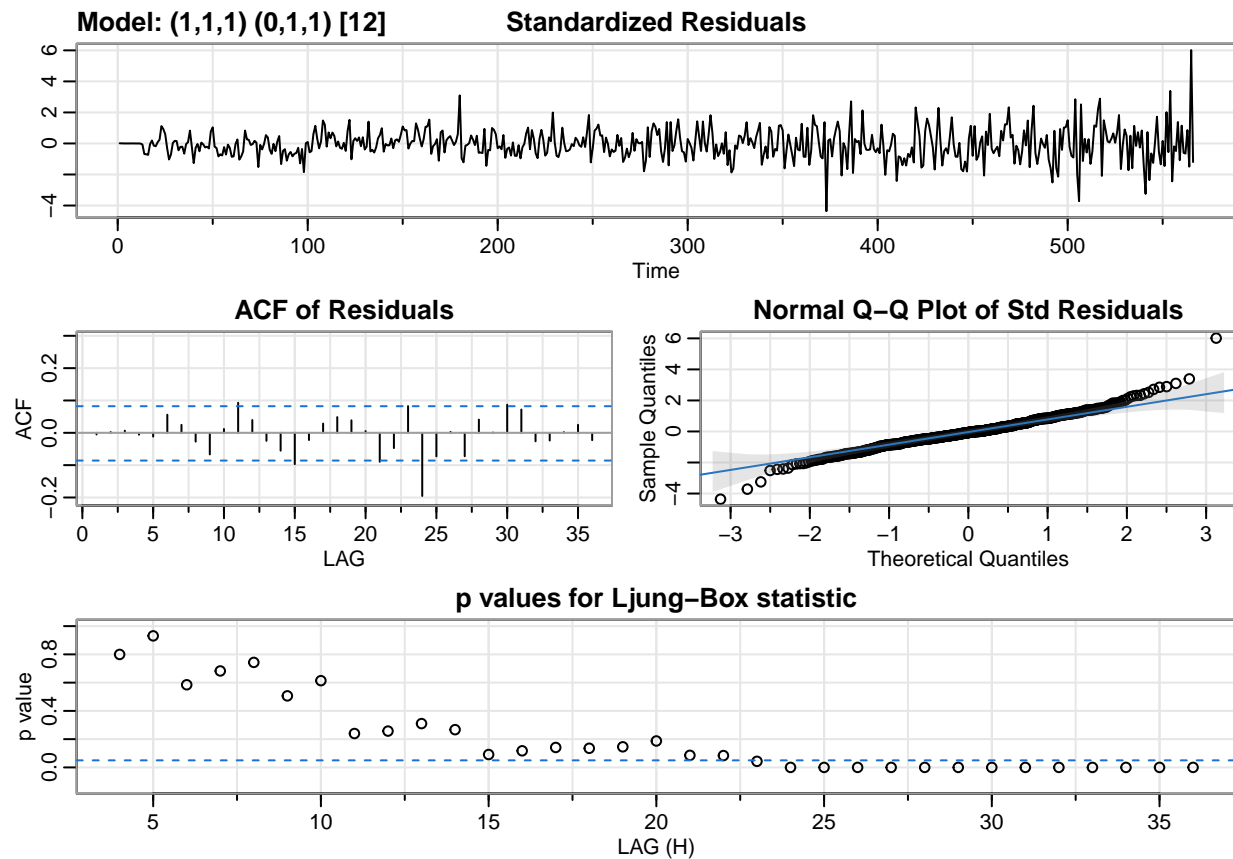
## Forecasting of USA utility output



```
mod4a2 <- sarima(df, 1, 1, 1, P = 0, D = 1, Q = 1, S = 12)
```

```
## initial value 1.196814
## iter 2 value 1.060460
## iter 3 value 0.966577
## iter 4 value 0.959790
## iter 5 value 0.955861
## iter 6 value 0.941389
## iter 7 value 0.930513
## iter 8 value 0.922721
## iter 9 value 0.921147
## iter 10 value 0.904351
## iter 11 value 0.895591
## iter 12 value 0.886471
## iter 13 value 0.885673
## iter 14 value 0.885659
## iter 15 value 0.885601
## iter 16 value 0.885599
## iter 17 value 0.885599
## iter 17 value 0.885599
## iter 17 value 0.885599
## final value 0.885599
```

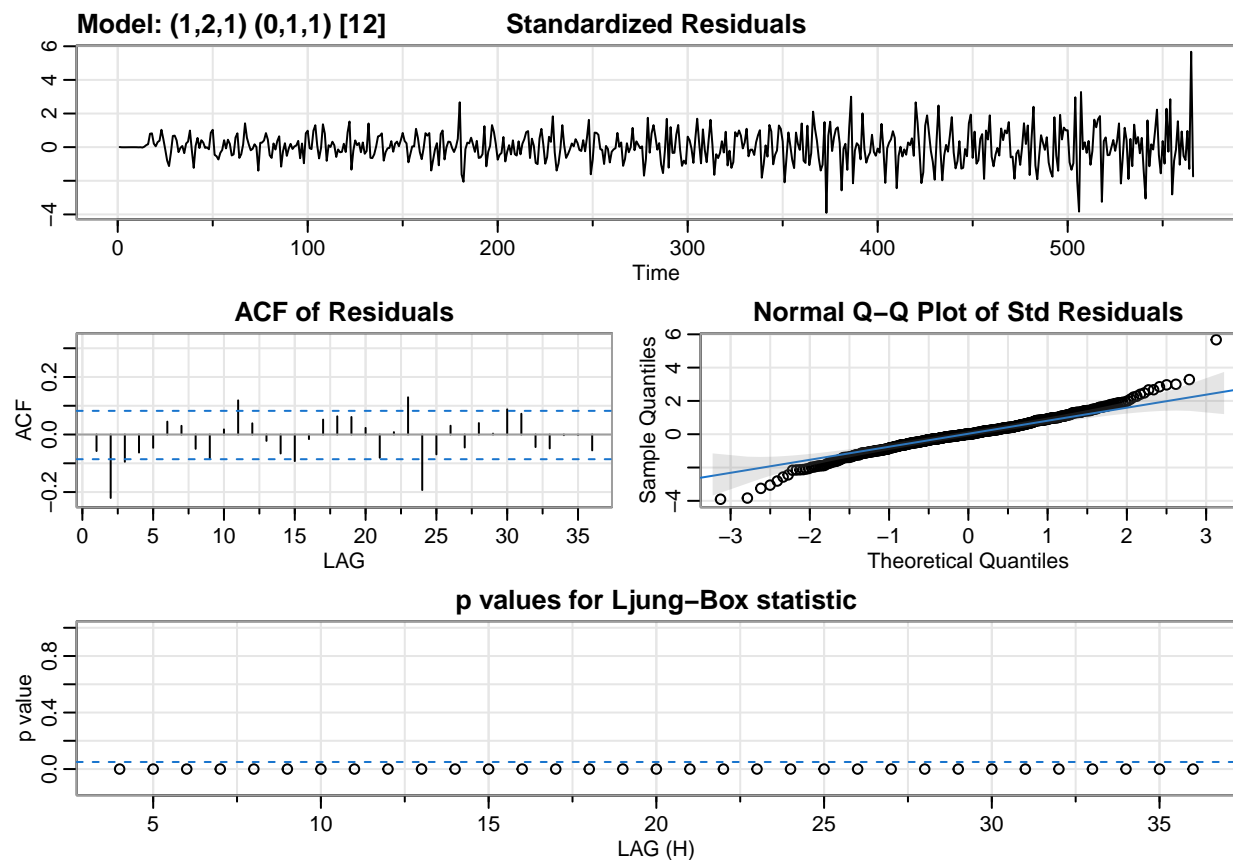
```
## converged
## initial value 0.895738
## iter 2 value 0.895658
## iter 3 value 0.895657
## iter 4 value 0.895604
## iter 5 value 0.895603
## iter 5 value 0.895603
## iter 5 value 0.895603
## final value 0.895603
## converged
```



```
mod4a3 <- sarima(df, 1, 2, 1, P = 0, D = 1, Q = 1, S = 12)
```

```
## initial value 1.664193
## iter 2 value 1.249103
## iter 3 value 1.196355
## iter 4 value 1.030922
## iter 5 value 1.027807
## iter 6 value 1.022797
## iter 7 value 1.019772
## iter 8 value 1.019676
## iter 9 value 1.019633
## iter 10 value 1.019633
## iter 11 value 1.019633
## iter 11 value 1.019633
## iter 11 value 1.019633
## final value 1.019633
```

```
## converged
## initial value 1.015640
## iter 2 value 0.992087
## iter 3 value 0.989338
## iter 4 value 0.988499
## iter 5 value 0.988386
## iter 6 value 0.988363
## iter 7 value 0.988361
## iter 8 value 0.988361
## iter 8 value 0.988361
## iter 8 value 0.988361
## final value 0.988361
## converged
```



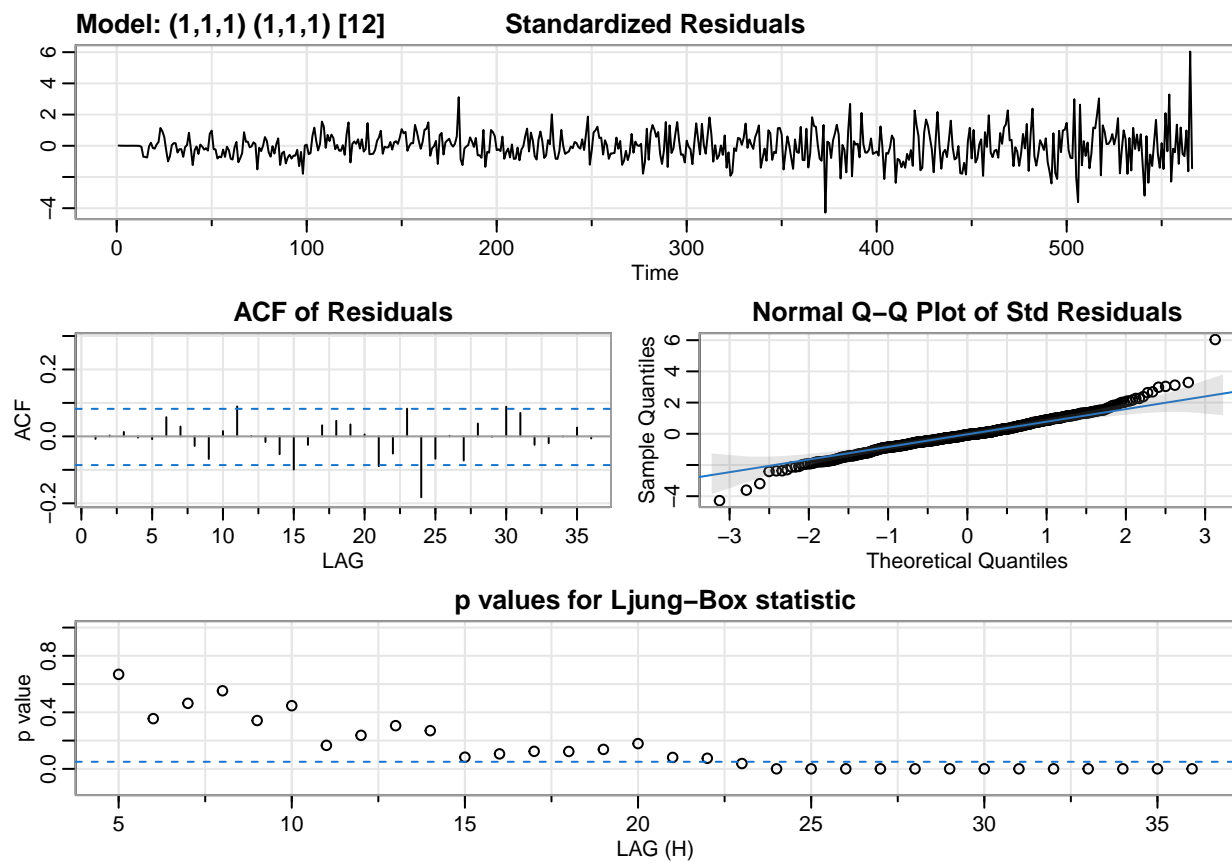
```
mod4a4 <- sarima(df, 1, 1, 1, P = 1, D = 1, Q = 1, S = 12)
```

```
## initial value 1.206077
## iter 2 value 1.061995
## iter 3 value 1.018643
## iter 4 value 0.989856
## iter 5 value 0.975888
## iter 6 value 0.958861
## iter 7 value 0.945745
## iter 8 value 0.941229
## iter 9 value 0.924644
## iter 10 value 0.913846
## iter 11 value 0.910599
```

```

## iter 12 value 0.902243
## iter 13 value 0.900396
## iter 14 value 0.900382
## iter 15 value 0.899479
## iter 16 value 0.899245
## iter 17 value 0.899178
## iter 18 value 0.899177
## iter 19 value 0.899177
## iter 19 value 0.899177
## final value 0.899177
## converged
## initial value 0.895335
## iter 2 value 0.894065
## iter 3 value 0.893668
## iter 4 value 0.893609
## iter 5 value 0.893553
## iter 6 value 0.893538
## iter 7 value 0.893537
## iter 8 value 0.893536
## iter 8 value 0.893536
## iter 8 value 0.893536
## final value 0.893536
## converged

```



```

mod4a5 <- sarima(df, 1, 2, 1, P = 1, D = 2, Q = 1, S = 12)

```

```

## initial value 2.132952

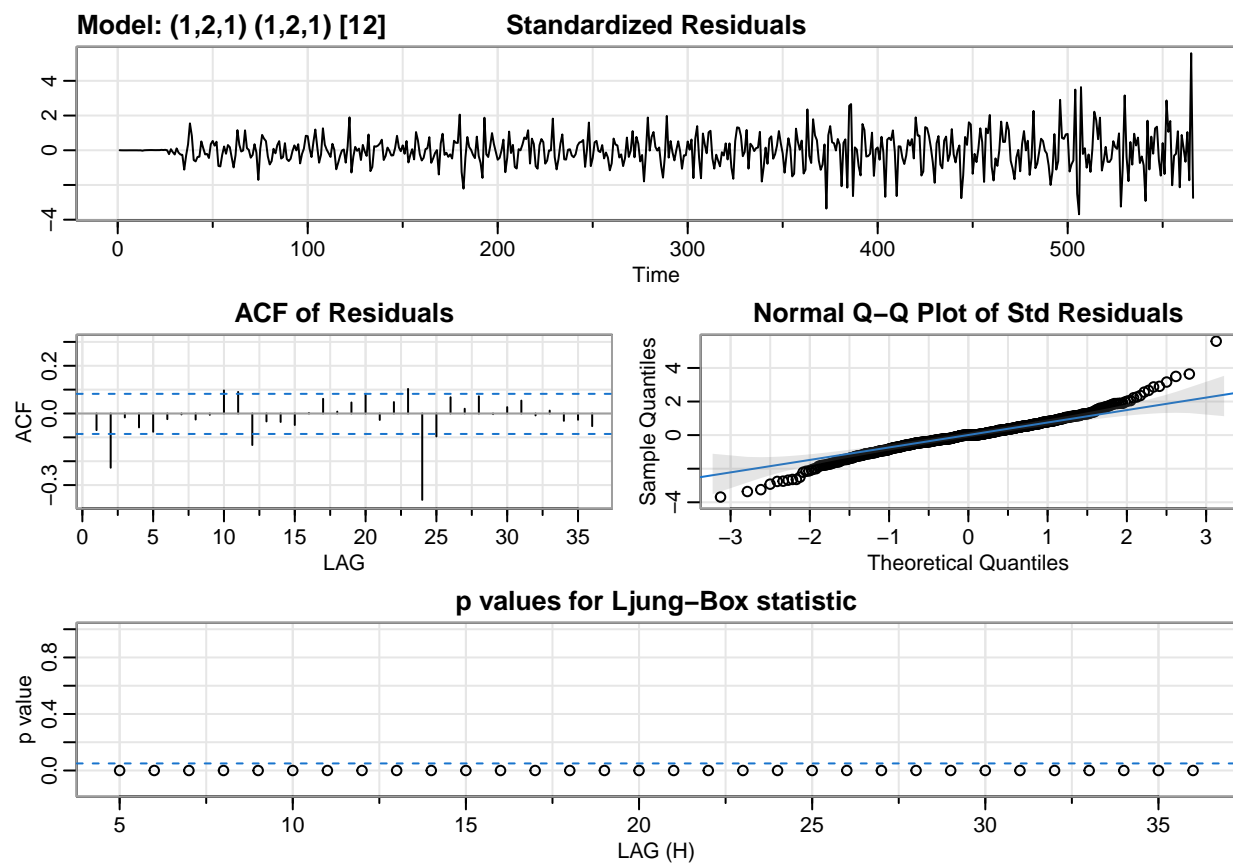
```



```

## iter 2 value 1.506780
## iter 3 value 1.446489
## iter 4 value 1.318096
## iter 5 value 1.260581
## iter 6 value 1.255807
## iter 7 value 1.255537
## iter 8 value 1.255187
## iter 9 value 1.255092
## iter 10 value 1.255035
## iter 11 value 1.255034
## iter 11 value 1.255034
## iter 11 value 1.255034
## final value 1.255034
## converged
## initial value 1.237722
## iter 2 value 1.186550
## iter 3 value 1.165124
## iter 4 value 1.164197
## iter 5 value 1.163324
## iter 6 value 1.162741
## iter 7 value 1.162714
## iter 8 value 1.162713
## iter 8 value 1.162713
## iter 8 value 1.162713
## final value 1.162713
## converged

```

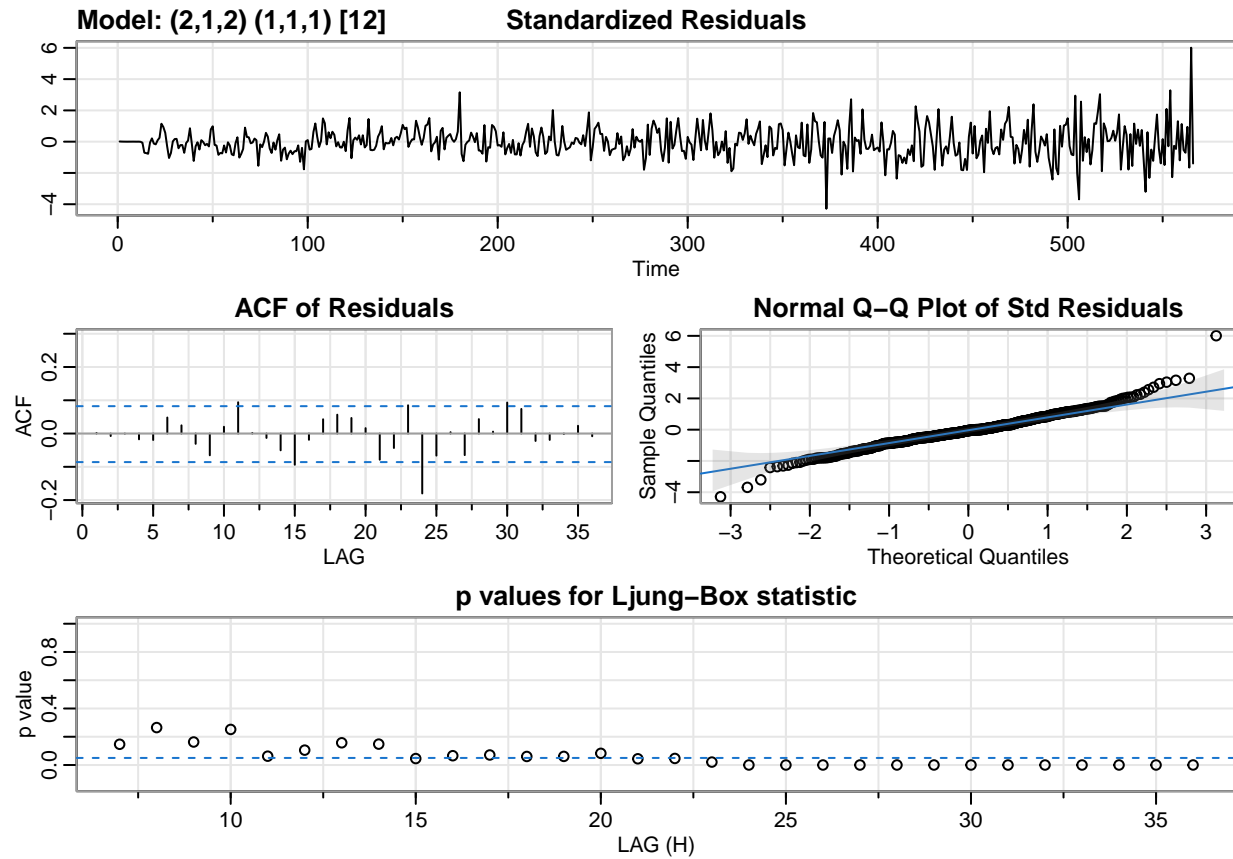


```
mod4a6 <- sarima(df, 2, 1, 2, P = 1, D = 1, Q = 1, S = 12)
```

```
## initial value 1.206557
## iter 2 value 1.030166
## iter 3 value 0.987616
## iter 4 value 0.961544
## iter 5 value 0.939853
## iter 6 value 0.921408
## iter 7 value 0.902978
## iter 8 value 0.896040
## iter 9 value 0.894071
## iter 10 value 0.893116
## iter 11 value 0.892428
## iter 12 value 0.892360
## iter 13 value 0.892351
## iter 14 value 0.892344
## iter 15 value 0.892343
## iter 16 value 0.892339
## iter 17 value 0.892328
## iter 18 value 0.892278
## iter 19 value 0.892188
## iter 20 value 0.892050
## iter 21 value 0.892023
## iter 22 value 0.892019
## iter 23 value 0.892019
## iter 23 value 0.892019
## final value 0.892019
## converged
## initial value 0.893879
## iter 2 value 0.893708
## iter 3 value 0.893600
## iter 4 value 0.893554
## iter 5 value 0.893488
## iter 6 value 0.893487
## iter 7 value 0.893487
## iter 8 value 0.893486
## iter 9 value 0.893486
## iter 10 value 0.893486
## iter 11 value 0.893486
## iter 12 value 0.893486
## iter 13 value 0.893485
## iter 14 value 0.893483
## iter 15 value 0.893482
## iter 16 value 0.893481
## iter 17 value 0.893480
## iter 18 value 0.893480
## iter 19 value 0.893480
## iter 20 value 0.893480
## iter 21 value 0.893480
## iter 22 value 0.893480
## iter 23 value 0.893479
## iter 24 value 0.893478
## iter 25 value 0.893472
```

```
## iter 26 value 0.893466
## iter 27 value 0.893431
## iter 28 value 0.893409
## iter 29 value 0.893321
## iter 30 value 0.893313
## iter 31 value 0.893305
## iter 32 value 0.893302
## iter 33 value 0.893302
## iter 34 value 0.893302
## iter 35 value 0.893302
## iter 36 value 0.893300
## iter 37 value 0.893297
## iter 38 value 0.893287
## iter 39 value 0.893273
## iter 40 value 0.893271
## iter 41 value 0.893261
## iter 42 value 0.893252
## iter 43 value 0.893249
## iter 44 value 0.893249
## iter 45 value 0.893248
## iter 46 value 0.893247
## iter 47 value 0.893243
## iter 48 value 0.893234
## iter 49 value 0.893214
## iter 50 value 0.893186
## iter 51 value 0.893141
## iter 52 value 0.893087
## iter 53 value 0.893085
## iter 54 value 0.893054
## iter 55 value 0.893047
## iter 56 value 0.893042
## iter 57 value 0.893041
## iter 58 value 0.893040
## iter 59 value 0.893036
## iter 60 value 0.893025
## iter 61 value 0.892991
## iter 62 value 0.892970
## iter 63 value 0.892944
## iter 64 value 0.892940
## iter 65 value 0.892859
## iter 66 value 0.892859
## iter 67 value 0.892850
## iter 68 value 0.892836
## iter 69 value 0.892835
## iter 70 value 0.892834
## iter 71 value 0.892832
## iter 72 value 0.892828
## iter 73 value 0.892816
## iter 74 value 0.892788
## iter 75 value 0.892761
## iter 76 value 0.892741
## iter 77 value 0.892733
## iter 77 value 0.892733
## iter 78 value 0.892733
```

```
## iter 79 value 0.892733
## iter 79 value 0.892733
## iter 79 value 0.892733
## final value 0.892733
## converged
```



```
mod4a2$AIC
```

```
## [1] 4.64355
```

```
mod4a3$AIC
```

```
## [1] 4.832716
```

```
mod4a4$AIC
```

```
## [1] 4.643033
```

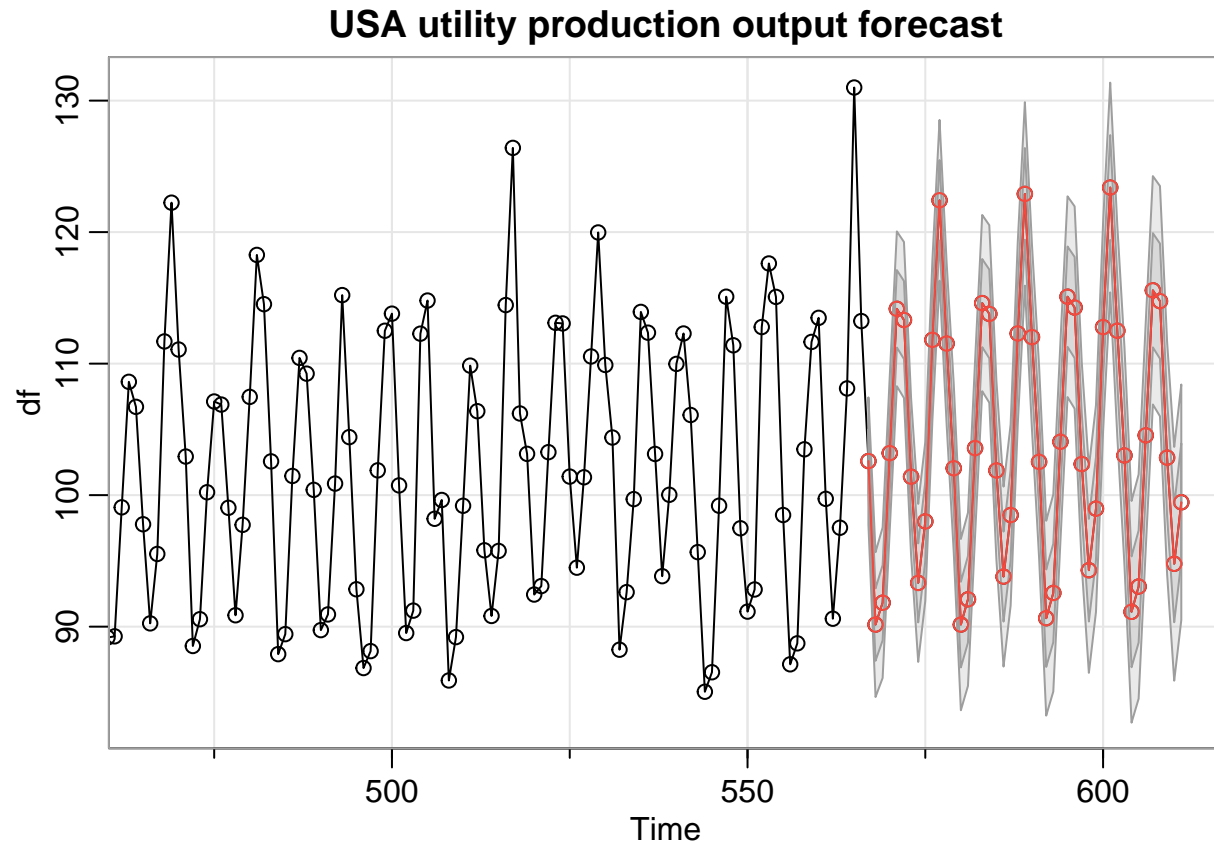
```
mod4a5$AIC
```

```
## [1] 5.181823
```

```
mod4a6$AIC
```

```
## [1] 4.648659
```

```
mod4a <- sarima.for(df,n.ahead = 45,1,1,1,P=0,D=1,Q=1,S=12, main = "USA utility production output forecast")
```



```
#mod4.pr = predict(mod4a, n.ahead=45, interval = "pred")
```

```
mod4.pr = mod4a
```

```
PI.nov = c(mod4.pr$pr[1] - 2*mod4.pr$se[1], mod4.pr$pr[1] + 2*mod4.pr$se[1])
```

```
PI.dec = c(mod4.pr$pr[2] - 2*mod4.pr$se[2], mod4.pr$pr[2] + 2*mod4.pr$se[2])
```

```
zhat = mod4.pr$pr
```

```
pi.z.upper = mod4.pr$pr + 2*mod4.pr$se
```

```
pi.z.lower = mod4.pr$pr - 2*mod4.pr$se
```

```
plot(df, ylab="Residuals", main=expression("Forecasting of USA utility output"))
```

```
points(mod4.pr$pred, col="brown")
```

```
lines(pi.z.upper, lty=2, col="blue")
```

```
lines(pi.z.lower, lty=2, col="orange")
```

Forecasting of USA utility output

