BOGAZICI UNIVERSITY



IE360 Term Project

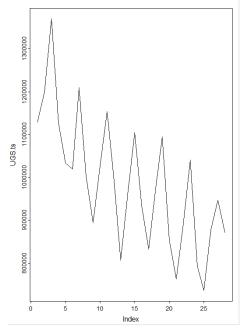
Instructor: Prof. Dr. Refik GÜLLÜ TA: Emre KARA SA: Halis OĞUZ

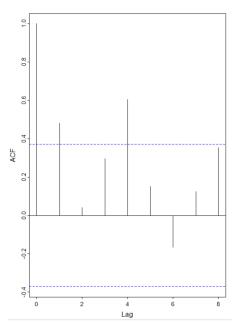
Arda Turan - 2019402075 Burak Berk Bulut - 2019402138 Batuhan Özkan - 2018402033 Mustafa Kutay Almak - 2019402144 Hasan Yağız Kılıç - 2019402147

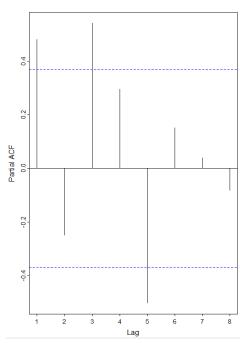
OBSERVATIONS OF THE DATA

UGS

```
UGS.ts <- ts(data$'Unleaded Gasoline Sale (UGS)', frequency=4)
UGS.ts <- UGS.ts[1:(length(UGS.ts)-4)]
plot(UGS.ts,type="I")
acf(UGS.ts,lag=8)
pacf(UGS.ts, lag=8)
plot(UGS.ts, type = "I", main = "Line Graph of UGS Time Series")
```

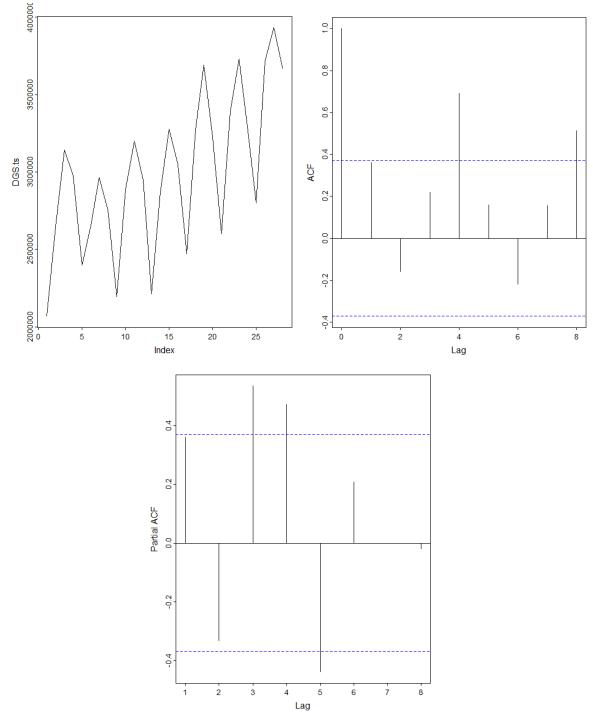






DGS

```
DGS.ts <- ts(data$'Diesel Gasoline Sale (DGS)',frequency=4)
DGS.ts <- DGS.ts[1:(length(DGS.ts)-4)]
plot(DGS.ts,type="I")
acf(DGS.ts,lag=8)
pacf(DGS.ts, lag=8)
plot(DGS.ts, type = "I", main = "Line Graph of UGS Time Series")
```



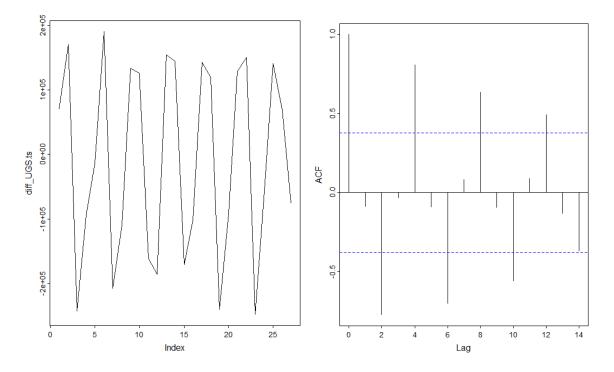
As it can be seen from the plots, both UGS and DGS data are not stationary and they need a preliminary transformation to induce their stationarity.

METHOD A

UGS

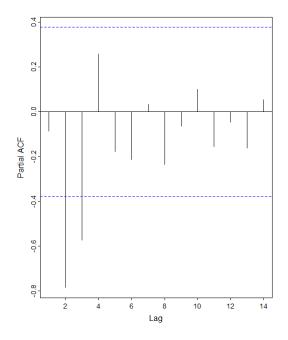
Part 1 and 2

diff_UGS.ts <- diff(UGS.ts)
plot(diff_UGS.ts, type = "I")
acf(diff_UGS.ts)
pacf(diff_UGS.ts)
Perform the Augmented Dickey–Fuller test to check the stationarity.
adf_test <- adf.test(diff_UGS.ts)
print(adf_test)
ADF test says that data is stationary.</pre>



To remove the trend, the difference of the data is taken. According to the ADF test, data is stationary now.

According to the ACF plot, data needs seasonal differencing.



Part 3

```
model <- auto.arima(UGS.ts, seasonal = TRUE)
model
Series: UGS.ts
ARIMA(0,1,0)

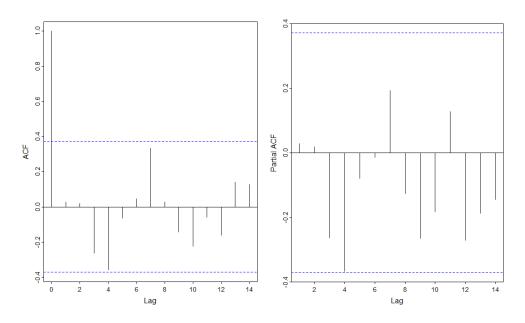
sigma^2 = 2.244e+10: log likelihood = -360.07
AIC=722.15 AICc=722.31 BIC=723.44</pre>
```

Auto.arima function is used to find a reference ARIMA model. Note that the SARIMA model is not considered in this part.

Part 4 and 5

```
model1<-sarima(UGS.ts,1,1,0,0,1,0,S=4,details=F)
model1
$fit
Call:
arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q),
period = S),
    include.mean = !no.constant, transform.pars = trans, fixed = fixed,
optim.control = list(trace = trc,
       REPORT = 1, reltol = tol))
Coefficients:
         ar1
      -0.5877
s.e. 0.2145
sigma^2 estimated as 3.133e+09: log likelihood = -284.3, aic = 572.59
$degrees_of_freedom
[1] 22
$ttable
                SE t.value p.value
   Estimate
ar1 -0.5877 0.2145 -2.7399 0.012
```

```
$AIC
[1] 24.89537
$AICc
[1] 24.90365
$BIC
[1] 24.99411
acf(resid(model1$fit), main = "ACF of residuals")
pacf(resid(model1$fit), main = "PACF of residuals")
```



model2<-sarima(UGS.ts,0,1,1,0,1,0,S=4,details=F) model2

\$AIC

[1] 24.92816

```
$fit
Call:
arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q),
period = S),
   include.mean = !no.constant, transform.pars = trans, fixed = fixed,
optim.control = list(trace = trc,
       REPORT = 1, reltol = tol))
Coefficients:
        ma1
     -0.5810
s.e.
    0.2207
sigma^2 = -284.67, aic = 573.35
$degrees_of_freedom
[1] 22
$ttable
              SE t.value p.value
   Estimate
     -0.581 0.2207 -2.6327 0.0152
ma1
```

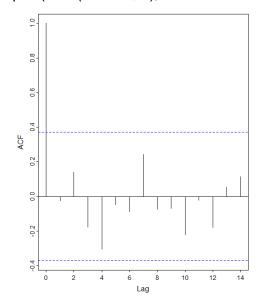
```
$AICc
```

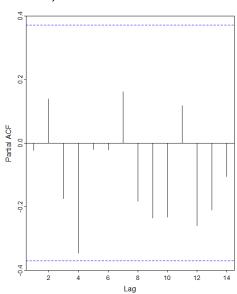
[1] 24.93644

\$BIC

[1] 25.0269

acf(resid(model2\$fit), main = "ACF of residuals")
pacf(resid(model2\$fit), main = "PACF of residuals")





model3<-sarima(UGS.ts,1,1,1,0,1,0,S=4,details=F) model3

\$fit

Call:

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

include.mean = !no.constant, transform.pars = trans, fixed = fixed,
optim.control = list(trace = trc,

REPORT = 1, reltol = tol))

Coefficients:

ar1 ma1 -0.6182 0.0412 s.e. 0.3302 0.3580

 $sigma^2 = -284.29$, aic = 574.58

\$degrees_of_freedom

[1] 21

\$ttable

Estimate SE t.value p.value ar1 -0.6182 0.3302 -1.8720 0.0752 ma1 0.0412 0.3580 0.1151 0.9095

\$AIC

[1] 24.98178

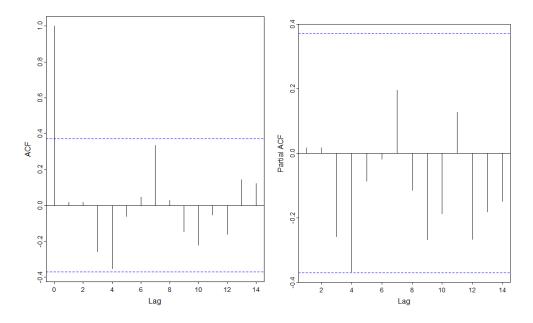
\$AICc

```
[1] 25.00787
```

\$BIC

[1] 25.12989

acf(resid(model3\$fit), main = "ACF of residuals")
pacf(resid(model3\$fit), main = "PACF of residuals")



model4<-sarima(UGS.ts,0,1,0,1,1,0,S=4,details=F) model4

\$fit

```
Call:
```

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

include.mean = !no.constant, transform.pars = trans, fixed = fixed,
optim.control = list(trace = trc,

REPORT = 1, reltol = tol))

Coefficients:

sar1

-0.2872

s.e. 0.2383

 $sigma^2 = -286.66$, aic = 577.32

\$degrees of freedom

[1] 22

\$ttable

Estimate SE t.value p.value sar1 -0.2872 0.2383 -1.2052 0.2409

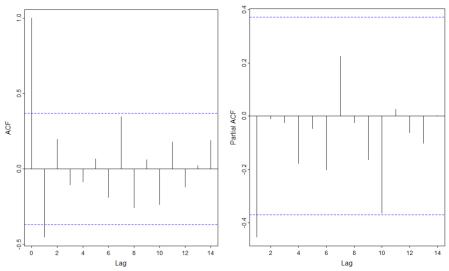
\$AIC

[1] 25.1008

\$AICc

[1] 25.10908

```
$BIC
[1] 25.19954
acf(resid(model4$fit), main = "ACF of residuals")
pacf(resid(model4$fit), main = "PACF of residuals")
```



model5<-sarima(UGS.ts,1,1,0,1,1,0,S=4,details=F) model5

\$fit

```
Call:
```

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

Coefficients:

ar1 sar1 -0.7037 -0.5098 s.e. 0.1940 0.2131

 $sigma^2 = -282.12$, aic = 570.24

\$degrees_of_freedom
[1] 21

\$ttable

Estimate SE t.value p.value ar1 -0.7037 0.1940 -3.6272 0.0016 sar1 -0.5098 0.2131 -2.3919 0.0262

\$AIC

[1] 24.79293

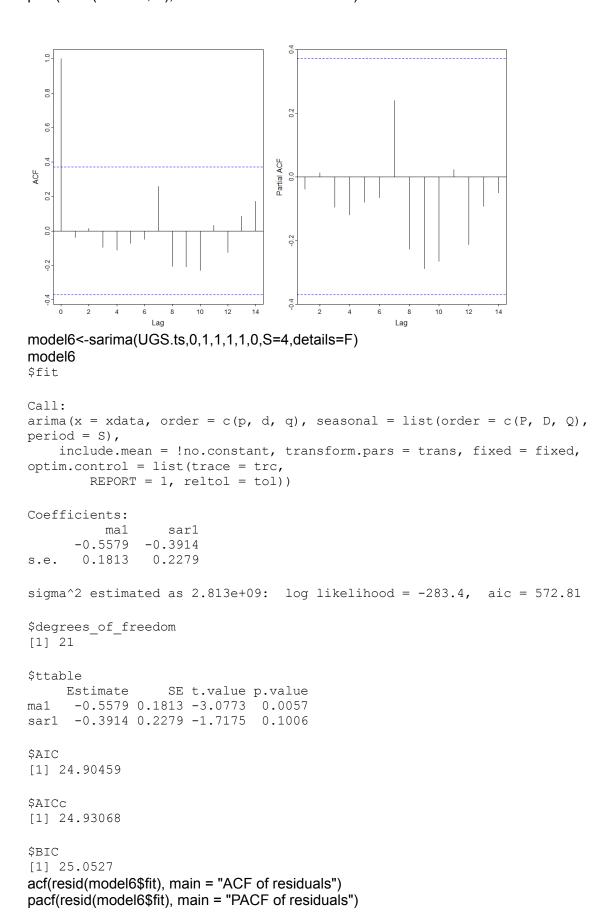
\$AICc

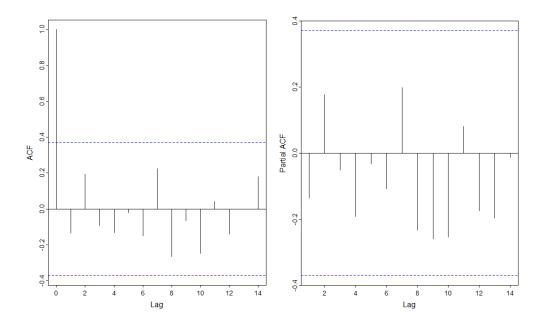
[1] 24.81901

\$BIC

[1] 24.94104

acf(resid(model5\$fit), main = "ACF of residuals")

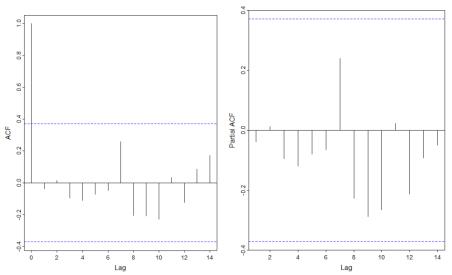




6 different SARIMA models have been tried with different (p,d,q)(P,D,Q) values and different AIC and AICc values have been obtained.

Part 6

model5 has the lowest AIC
AIC of model 5 is 24.79293
best_model <- model5
acf(resid(best_model\$fit), main = "ACF of residuals (best model)")
pacf(resid(best_model\$fit), main = "PACF of residuals (best model)")</pre>



Model 5 has the lowest AIC value and it has been chosen as the best model. *Consider that the dataset has few observations, looking the AICc values is more accurate. Since AICc is AIC*(correction factor), it still gives the Model 5 as the best option.

Part 7

The ACF and the PACF of the residuals are analyzed. Addition to them, we constructed more statistical test to show the model's validity.

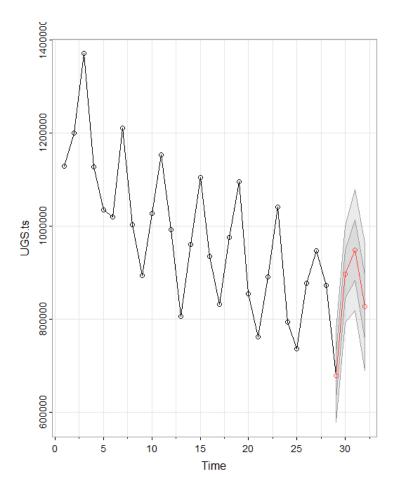
```
# Apply the Ljung-Box test to the residuals
ljung_box_test <- Box.test(resid(best_model$fit), lag = 12, type = "Ljung-Box")</pre>
print(ljung_box_test)
Box-Ljung test
data: resid(best model$fit)
X-squared = 10.75\overline{8}, df = 12, p-value = 0.5497
#p-value is 0.5497. This shows that there is no significant evidence of autocorrelation in the
residuals.
#Applying Box-Pierce
box_pierce_test <- Box.test(resid(best_model$fit), lag = 12, type = "Box-Pierce")
print(box pierce test)
Box-Pierce test
data: resid(best model$fit)
X-squared = 7.0673, df = 12, p-value = 0.8531
#p-value is 0.8531. This shows that there is no significant evidence of autocorrelation in the
residuals.
```

To validate that the model that has been chosen is true, Ljung-Box and Box-Pierce tests have been applied and according to the p-values, the model has passed the tests.

Part 8

```
UGS_forecast <- sarima.for(UGS.ts,4,1,1,0,1,1,0,4)
UGS_forecast
```

```
$pred
Time Series:
Start = 29
End = 32
Frequency = 1
[1] 678288.8 896475.5 948056.1 826669.8
$se
Time Series:
Start = 29
End = 32
Frequency = 1
[1] 49572.18 51702.19 64904.67 68519.26
## Predicted values for 2007 sales:
## 2007 Q1 = 678288.8
## 2007 Q2 = 896475.5
## 2007 Q3 = 948056.1
## 2007 Q4 = 826669.8
```



Forecast of 4 quarters of 2007 is obtained by using the best SARIMA model.

DGS

Part 1 and 2

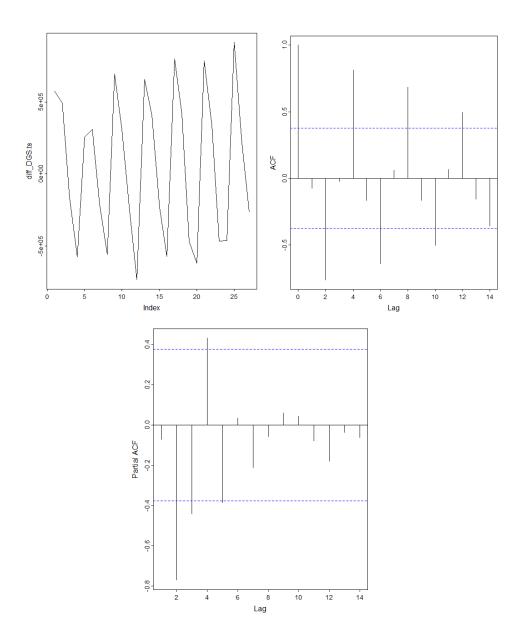
```
diff_DGS.ts <- diff(DGS.ts)
plot(diff_DGS.ts, type = "I")
acf(diff_DGS.ts)
pacf(diff_DGS.ts)
# Perform the ADF test
adf_test <- adf.test(diff_DGS.ts)
print(adf_test)
Augmented Dickey-Fuller Test

data: diff_DGS.ts
Dickey-Fuller = -13.984, Lag order = 2, p-value = 0.01
alternative hypothesis: stationary</pre>
```

ADF test says that data is stationary.

To remove the trend, difference of the data is taken. According to the ADF test, data is stationary now.

According to the ACF plot, data needs seasonal differencing.



Part 3 model <- auto.arima(DGS.ts, seasonal = TRUE) model

Series: DGS.ts

ARIMA(3,1,0) with drift

Coefficients:

sigma^2 = 3.518e+10: log likelihood = -366.89
AIC=743.77 AICc=746.63 BIC=750.25

Auto.arima function is used to find a reference ARIMA model. Note that the SARIMA model is not considered in this part.

Part 4 and 5

```
model1<-sarima(DGS.ts,3,1,0,0,1,0,S=4,details=F)
model1
```

\$fit

```
Call:
```

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

include.mean = !no.constant, transform.pars = trans, fixed = fixed, optim.control = list(trace = trc,

REPORT = 1, reltol = tol))

Coefficients:

ar1 ar2 ar3 -0.3251 0.1629 -0.0854 0.2265 0.2293 0.2291 s.e.

 $sigma^2$ estimated as 2.171e+10: log likelihood = -306.56, aic = 621.11

\$degrees of freedom

[1] 20

\$ttable

Estimate SE t.value p.value ar1 0.1629 0.2265 0.7193 0.4803 ar2 -0.0854 0.2293 -0.3725 0.7134 ar3 -0.3251 0.2291 -1.4190 0.1713

\$AIC

[1] 27.00493

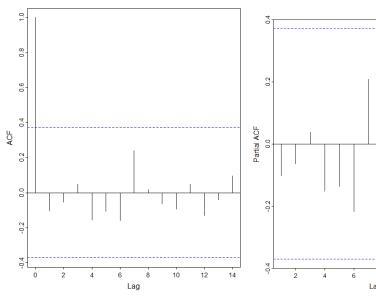
\$AICc

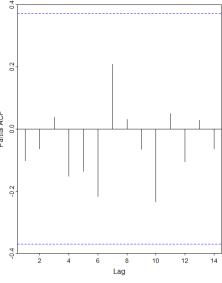
[1] 27.05985

\$BIC

[1] 27.2024

acf(resid(model1\$fit), main = "ACF of residuals") pacf(resid(model1\$fit), main = "PACF of residuals")





$model2 < -sarima(DGS.ts, 3, 1, 1, 0, 1, 0, S=4, details=F) \\ model2$

\$fit

Call:

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

include.mean = !no.constant, transform.pars = trans, fixed = fixed,
optim.control = list(trace = trc,

REPORT = 1, reltol = tol))

Coefficients:

ar1 ar2 ar3 ma1 0.6657 -0.1584 -0.3674 -0.8012 s.e. 0.2735 0.2758 0.2476 0.1835

 $sigma^2$ estimated as 1.674e+10: log likelihood = -304.25, aic = 618.51

\$degrees_of_freedom

[1] 19

\$ttable

Estimate SE t.value p.value ar1 0.6657 0.2735 2.4338 0.0250 ar2 -0.1584 0.2758 -0.5745 0.5724 ar3 -0.3674 0.2476 -1.4838 0.1543 ma1 -0.8012 0.1835 -4.3653 0.0003

\$AIC

[1] 26.89153

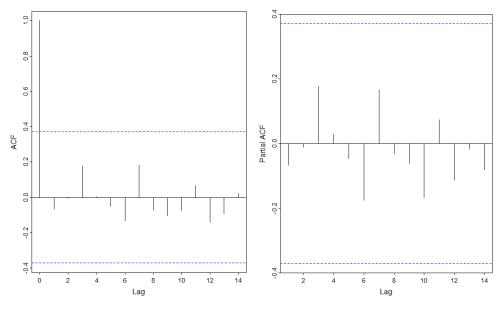
\$AICc

[1] 26.98815

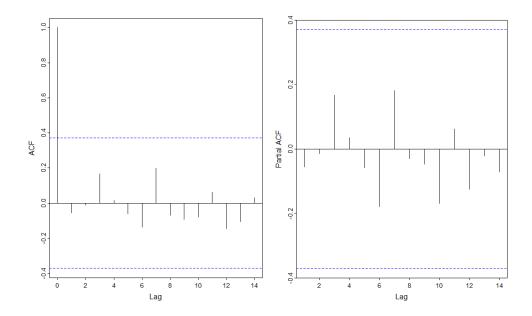
\$BIC

[1] 27.13838

acf(resid(model2\$fit), main = "ACF of residuals")
pacf(resid(model2\$fit), main = "PACF of residuals")



```
model3<-sarima(DGS.ts,3,1,1,0,1,1,S=4,details=F)
model3
$fit
Call:
arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q),
period = S),
    include.mean = !no.constant, transform.pars = trans, fixed = fixed,
optim.control = list(trace = trc,
       REPORT = 1, reltol = tol))
Coefficients:
                 ar2
        ar1
                          ar3
                                  ma1
                                           sma1
      0.6529 -0.1635 -0.3373 -0.7911 -0.0597
s.e. 0.2961 0.2784 0.3487 0.2060 0.4519
sigma^2 estimated as 1.677e+10: log likelihood = -304.24, aic = 620.49
$degrees_of_freedom
[1] 18
$ttable
               SE t.value p.value
    Estimate
arl 0.6529 0.2961 2.2053 0.0407
     -0.1635 0.2784 -0.5873 0.5643
ar2
ar3 -0.3373 0.3487 -0.9673 0.3462
ma1 -0.7911 0.2060 -3.8402 0.0012
sma1 -0.0597 0.4519 -0.1320 0.8964
$AIC
[1] 26.9777
$AICc
[1] 27.13115
$BIC
[1] 27.27392
acf(resid(model3$fit), main = "ACF of residuals")
pacf(resid(model3$fit), main = "PACF of residuals")
```



model4<-sarima(DGS.ts,3,1,0,1,1,0,S=4,details=F) model4

\$fit

```
Call:
```

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

Coefficients:

 $sigma^2$ estimated as 1.824e+10: log likelihood = -305.1, aic = 620.19

\$degrees_of_freedom [1] 19

\$ttable

Estimate SE t.value p.value ar1 -0.1565 0.2647 -0.5913 0.5613 ar2 -0.2894 0.2463 -1.1750 0.2545 ar3 -0.2031 0.2386 -0.8513 0.4052 sar1 -0.5636 0.2447 -2.3030 0.0327

\$AIC

[1] 26.96483

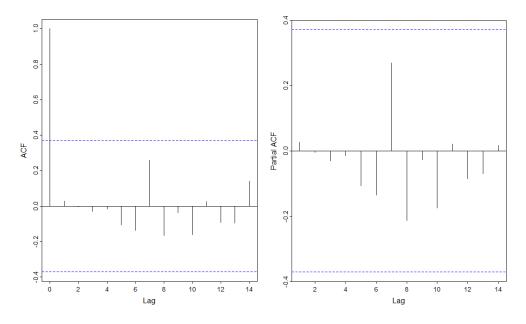
\$AICc

[1] 27.06145

\$BIC

[1] 27.21167

acf(resid(model4\$fit), main = "ACF of residuals")
pacf(resid(model4\$fit), main = "PACF of residuals")



model5<-sarima(DGS.ts,3,1,0,1,1,1,S=4,details=F) model5

\$fit

```
Call:
```

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

include.mean = !no.constant, transform.pars = trans, fixed = fixed,
optim.control = list(trace = trc,

REPORT = 1, reltol = tol))

Coefficients:

 $sigma^2 = -304.91$, aic = 621.81

\$degrees_of_freedom [1] 18

\$ttable

Estimate SE t.value p.value ar1 -0.1447 0.2684 -0.5390 0.5965 ar2 -0.4033 0.2903 -1.3893 0.1817 ar3 -0.1627 0.2472 -0.6581 0.5188 sar1 -0.3352 0.4729 -0.7089 0.4875 sma1 -0.3637 0.4756 -0.7648 0.4543

\$AIC

[1] 27.03524

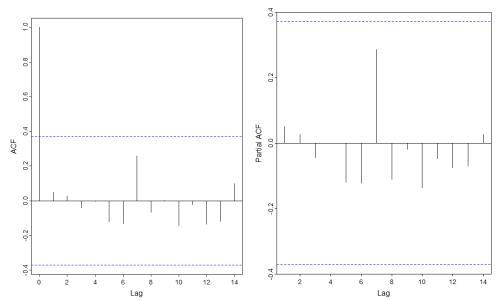
\$AICc

[1] 27.18869

\$BIC

[1] 27.33145

acf(resid(model5\$fit), main = "ACF of residuals") pacf(resid(model5\$fit), main = "PACF of residuals")



model6<-sarima(DGS.ts,3,1,1,1,1,0,S=4,details=F) model6

\$fit

Call:

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

include.mean = !no.constant, transform.pars = trans, fixed = fixed,
optim.control = list(trace = trc,

REPORT = 1, reltol = tol))

Coefficients:

```
ar1 ar2 ar3 ma1 sar1 
0.6546 -0.1619 -0.3418 -0.7930 -0.0511 
s.e. 0.2954 0.2761 0.3378 0.2015 0.4308
```

 $sigma^2$ estimated as 1.676e+10: log likelihood = -304.24, aic = 620.49

\$degrees_of_freedom [1] 18

\$ttable

Estimate SE t.value p.value ar1 0.6546 0.2954 2.2159 0.0398 ar2 -0.1619 0.2761 -0.5864 0.5649 ar3 -0.3418 0.3378 -1.0119 0.3250 ma1 -0.7930 0.2015 -3.9348 0.0010 sar1 -0.0511 0.4308 -0.1187 0.9069

\$AIC

[1] 26.97782

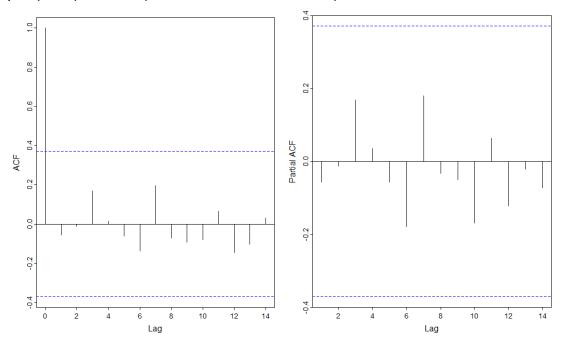
\$AICc

[1] 27.13127

\$BIC

[1] 27.27403

acf(resid(model6\$fit), main = "ACF of residuals")
pacf(resid(model6\$fit), main = "PACF of residuals")

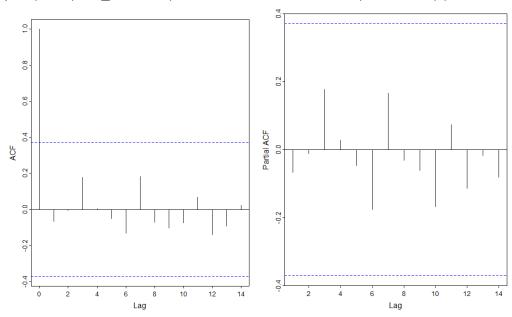


6 different SARIMA models have been tried with different (p,d,q)(P,D,Q) values and different AIC and AICc values have been obtained.

Part 6

model2 has the lowest AIC
AIC of model2 is 26.89153
best_model <- model2
acf(resid(best_model\$fit), main = "ACF of residuals (best model)")</pre>

pacf(resid(best_model\$fit), main = "PACF of residuals (best model)")



Model 5 has the lowest AIC value and it has been chosen as the best model. *Consider that the dataset has few observations, looking the AICc values is more accurate. Since AICc is AIC*(correction factor), it still gives the Model 5 as the best option

Part 7

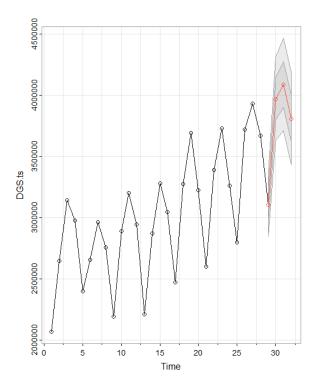
The ACF and the PACF of the residuals are analyzed. Addition to them, we constructed more statistical test to show the model's validity. # Apply the Ljung-Box test to the residuals ljung box test <- Box.test(resid(best model\$fit), lag = 12, type = "Ljung-Box") print(ljung_box_test) Box-Ljung test data: resid(best model\$fit) X-squared = $5.480\overline{5}$, df = 12, p-value = 0.94 #p-value is 0.94. This shows that there is no significant evidence of autocorrelation in the residuals. #Applying Box-Pierce box_pierce_test <- Box.test(resid(best_model\$fit), lag = 12, type = "Box-Pierce") print(box_pierce_test) Box-Pierce test data: resid(best model\$fit) X-squared = 3.765, df = 12, p-value = 0.9873 #p-value is 0.9873. This shows that there is no significant evidence of

To validate that the model that has been chosen is true, Ljung-Box and Box-Pierce tests have been applied and according to the p-values, the model has passed the tests.

Part 8

```
DGS forecast <- sarima.for(DGS.ts,4,3,1,1,0,1,0,4)
DGS_forecast
$pred
Time Series:
Start = 29
End = 32
Frequency = 1
[1] 3105950 3968121 4087278 3808176
$se
Time Series:
Start = 29
End = 32
Frequency = 1
[1] 129382.3 171026.2 188676.0 189157.4
## Predicted values for 2007 sales:
## 2007 Q1 = 3105950
## 2007 Q2 = 3968121
## 2007 Q3 = 4087278
## 2007 Q4 = 3808176
```

autocorrelation in the residuals.



Forecast of 4 quarters of 2007 is obtained by using the best SARIMA model.

METHOD B

<u>UGS</u>

data<-read_excel("IE360-ProjectData.xlsx",col_names = TRUE)
cor(data[sapply(data, is.numeric)], use = "complete.obs")</pre>

	Unleaded Gasoline Sale (UGS)
Unleaded Gasoline Sale (UGS)	1.00000000
Diesel Gasoline Sale (DGS)	0.04670539
RNUV	0.26498942
# LPG Vehicles (NLPG)	-0.62890683
Price of Unleaded Gasoline (PU)	-0.48993252
Price of Diesel Gasoline (PG)	-0.54812079
# Unleaded Gasoline Vehicles (NUGV)	-0.64862885
# of Diesel Gasoline Vehicles (NDGV)	-0.52526929

GNP Agriculture	0.58815962
GNP Commerce	0.20843693
GNP Total	0.19348598
	Diesel Gasoline Sale (DGS) RNUV
Unleaded Gasoline Sale (UGS)	0.04670539 0.26498942
Diesel Gasoline Sale (DGS)	1.00000000 0.30364955
RNUV	0.30364955 1.00000000
# LPG Vehicles (NLPG)	0.63895652 0.18522216
Price of Unleaded Gasoline (PU)	0.49630315 0.04177338
Price of Diesel Gasoline (PG)	0.54026568 0.07378405
# Unleaded Gasoline Vehicles (NUGV)	0.67288637 0.07413199
# of Diesel Gasoline Vehicles (NDGV)	0.65656807 0.30576056
GNP Agriculture	0.60365821 0.16530707
GNP Commerce	0.91741674 0.29698204
GNP Total	0.87962217 0.30260566
	# LPG Vehicles (NLPG)
Unleaded Gasoline Sale (UGS)	-0.62890683
Diesel Gasoline Sale (DGS)	0.63895652
RNUV	0.18522216
# LPG Vehicles (NLPG)	1.00000000
Price of Unleaded Gasoline (PU)	0.67602376
Price of Diesel Gasoline (PG)	0.76400431
# Unleaded Gasoline Vehicles (NUGV)	0.92714875
# of Diesel Gasoline Vehicles (NDGV)	0.90937746
GNP Agriculture	0.03664584
GNP Commerce	0.49970806
GNP Total	0.49188935
	Price of Unleaded Gasoline (PU)
Unleaded Gasoline Sale (UGS)	-0.48993252
Diesel Gasoline Sale (DGS)	0.49630315
RNUV	0.04177338
# LPG Vehicles (NLPG)	0.67602376
Price of Unleaded Gasoline (PU)	1.00000000
Price of Diesel Gasoline (PG)	0.98211135
# Unleaded Gasoline Vehicles (NUGV)	0.73923142

# of Diesel Gasoline Vehicles (NDGV)	0.68439875
GNP Agriculture	0.11288812
GNP Commerce	0.48075691
GNP Total	0.45775513
	Price of Diesel Gasoline (PG)
Unleaded Gasoline Sale (UGS)	-0.54812079
Diesel Gasoline Sale (DGS)	0.54026568
RNUV	0.07378405
# LPG Vehicles (NLPG)	0.76400431
Price of Unleaded Gasoline (PU)	0.98211135
Price of Diesel Gasoline (PG)	1.0000000
# Unleaded Gasoline Vehicles (NUGV)	0.81214044
# of Diesel Gasoline Vehicles (NDGV)	0.75830059
GNP Agriculture	0.10869816
GNP Commerce	0.51351013
GNP Total	0.49819657
	# Unleaded Gasoline Vehicles (NUGV)
Unleaded Gasoline Sale (UGS)	-0.64862885
Diesel Gasoline Sale (DGS)	0.67288637
RNUV	0.07413199
# LPG Vehicles (NLPG)	0.92714875
Price of Unleaded Gasoline (PU)	0.73923142
Price of Diesel Gasoline (PG)	0.81214044
# Unleaded Gasoline Vehicles (NUGV)	1.0000000
# of Diesel Gasoline Vehicles (NDGV)	0.94565611
GNP Agriculture	0.06599231
GNP Commerce	0.51971136
GNP Total	0.51050156
	# of Diesel Gasoline Vehicles (NDGV)
Unleaded Gasoline Sale (UGS)	-0.52526929
Diesel Gasoline Sale (DGS)	0.65656807
RNUV	0.30576056
# LPG Vehicles (NLPG)	0.90937746
Price of Unleaded Gasoline (PU)	0.68439875
Price of Diesel Gasoline (PG)	0.75830059

# Unleaded Gasoline Vehicles (NUGV)	0.94565611
# of Diesel Gasoline Vehicles (NDGV)	1.00000000
GNP Agriculture	0.06716321
GNP Commerce	0.50529894
GNP Total	0.49132614
	GNP Agriculture GNP Commerce GNP Total
Unleaded Gasoline Sale (UGS)	0.58815962 0.2084369 0.1934860
Diesel Gasoline Sale (DGS)	0.60365821 0.9174167 0.8796222
RNUV	0.16530707 0.2969820 0.3026057
# LPG Vehicles (NLPG)	0.03664584 0.4997081 0.4918894
Price of Unleaded Gasoline (PU)	0.11288812 0.4807569 0.4577551
Price of Diesel Gasoline (PG)	0.10869816 0.5135101 0.4981966
# Unleaded Gasoline Vehicles (NUGV)	0.06599231 0.5197114 0.5105016
# of Diesel Gasoline Vehicles (NDGV)	0.06716321 0.5052989 0.4913261
GNP Agriculture	1.00000000 0.8170954 0.8343358
GNP Commerce	0.81709538 1.0000000 0.9866848
GNP Total	0.83433576 0.9866848 1.0000000

x2<- 1:32 %% 4 == 2

x3<- 1:32 %% 4 == 3

x4<- 1:32 %% 4 == 0

#We created logical vectors x2, x3, and x4 that are used to create dummy variables to identify specific quarters in the dataset. Then we add them to the data.

data1<- data.frame(data,s2 = 1*x2,s3=1*x3,s4=1*x4)

data2<- data.frame(data,Quarters = 1:32,s2 = 1*x2,s3=1*x3,s4=1*x4)

 $Im1 \leftarrow Im(Unleaded.Gasoline.Sale..UGS. \sim s2 + s3 + s4, data = data1)$ summary(Im1)

Call:

lm(formula = Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4, data = data1)

Residuals:

Min 1Q Median 3Q Max

```
Coefficients:
          Estimate Std. Error t value Pr(>|t|)
(Intercept) 884908 47604 18.589 9.36e-16 ***
            108035
                       67323 1.605 0.12163
s2
             246624
54557
                       67323 3.663 0.00123 **
s3
                       67323 0.810 0.42568
s4
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 125900 on 24 degrees of freedom
 (4 observations deleted due to missingness)
Multiple R-squared: 0.3819, Adjusted R-squared: 0.3047
F-statistic: 4.943 on 3 and 24 DF, p-value: 0.008198
lm2<- lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters,data = data2)
summary(lm2)
Call:
lm(formula = Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters,
   data = data2)
Residuals:
  Min 1Q Median 3Q Max
-81167 -31283 -3458 28640 94502
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 1060372 23653 44.830 < 2e-16 ***
                        25987 4.677 0.000104 ***
            121532
s2
            273618
                       26063 10.498 3.03e-10 ***
s3
             95049
                       26189 3.629 0.001405 **
\leq 4
                        1147 -11.764 3.28e-11 ***
Quarters
           -13497
___
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 48570 on 23 degrees of freedom
  (4 observations deleted due to missingness)
Multiple R-squared: 0.9119, Adjusted R-squared: 0.8966
F-statistic: 59.53 on 4 and 23 DF, p-value: 8.446e-12
```

-184749 -86358 -22243 55218 244063

#Check the largest absolute correlation value with UGS from the table of correlation. NUGV is the largest one. First, add the NUGV to the model.

```
Im3<-Im(Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..Unleaded.Gasoline.Vehicles..NUGV., data = data2)
```

summary(Im3)

```
lm(formula = Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Ouarters +
   X..Unleaded.Gasoline.Vehicles..NUGV., data = data2)
Residuals:
          1Q Median 30
  Min
                              Max
-87257 -30446 1812 27878 86183
Coefficients:
                                     Estimate Std. Error t value
                                     7.402e+05 5.981e+05
(Intercept)
                                                          1.238
                                                          4.575
                                     1.209e+05 2.643e+04
s2
                                     2.722e+05 2.661e+04 10.229
s3
                                     9.203e+04 2.720e+04
s4
                                                           3.384
                                    -1.592e+04 4.674e+03 -3.406
Ouarters
                                    6.856e-02 1.280e-01 0.536
X..Unleaded.Gasoline.Vehicles..NUGV.
                                    Pr(>|t|)
(Intercept)
                                    0.228896
                                    0.000148 ***
s2
                                    7.98e-10 ***
s3
                                    0.002673 **
s4
Ouarters
                                    0.002533 **
X.. Unleaded. Gasoline. Vehicles.. NUGV. 0.597494
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 49340 on 22 degrees of freedom
 (4 observations deleted due to missingness)
Multiple R-squared: 0.913, Adjusted R-squared: 0.8933
F-statistic: 46.2 on 5 and 22 DF, p-value: 6.113e-11
```

#Even though NUGV is the largest correlation value with UGS, When we look at the summary NUGV seems unsignificant so try another one with the second largest correlation value which is NLPG

Im4<-Im(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+X..LPG.Vehicles..NLPG.,data = data2)

summary(lm4)

```
lm(formula = Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters +
   X..LPG.Vehicles..NLPG., data = data2)
Residuals:
  Min 1Q Median 3Q
                             Max
-83022 -23287 3235 18555 90847
Coefficients:
                       Estimate Std. Error t value Pr(>|t|)
                       8.082e+05 7.363e+04 10.977 2.16e-10 ***
(Intercept)
                      1.286e+05 2.128e+04 6.044 4.40e-06 ***
s2
                      2.824e+05 2.140e+04 13.200 6.24e-12 ***
s3
                      9.835e+04 2.137e+04 4.601 0.000139 ***
s4
                     -2.422e+04 3.163e+03 -7.657 1.21e-07 ***
Quarters
```

```
X..LPG.Vehicles..NLPG. 3.259e-01 9.184e-02 3.549 0.001800 **
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 39600 on 22 degrees of freedom
  (4 observations deleted due to missingness)
Multiple R-squared: 0.944, Adjusted R-squared: 0.9313
F-statistic: 74.15 on 5 and 22 DF, p-value: 5.077e-13
```

#NLPG seems significant so we added NLPG to the model. We will decide what we should add to the model afterwards by doing anova

#We will go with the model which has the smallest p value

```
model<-lm(Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.,data = data2)
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+X..LPG.Vehicles..NLP G.+RNUV,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG. + RNUV

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.4505e+10
2 21 3.1850e+10 1 2654743072 1.7504 0.2001
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.+Price.of.Unleaded.Gasoline..PU.,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG. + Price.of.Unleaded.Gasoline..PU.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.4505e+10
2 21 3.0810e+10 1 3695242120 2.5187 0.1274
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+X..LPG.Vehicles..NLP G.+Price.of.Diesel.Gasoline..PG.,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG. + Price.of.Diesel.Gasoline..PG.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.4505e+10
2 21 3.2381e+10 1 2.124e+09 1.3775 0.2537
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+X..LPG.Vehicles..NLP G.+X..Unleaded.Gasoline.Vehicles..NUGV.,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG. + X..Unleaded.Gasoline.Vehicles..NUGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.4505e+10
2 21 3.4071e+10 1 434605087 0.2679 0.6102
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS. \sim s2+s3+s4+Quarters+X..LPG.Vehicles..NLP G.+X..of.Diesel.Gasoline.Vehicles..NDGV.,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG. + X..of.Diesel.Gasoline.Vehicles..NDGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.4505e+10
2 21 3.1873e+10 1 2632631294 1.7346 0.202
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+NX..LPG.Vehicles..NL PG.LPG+GNP.Agriculture,data = data2))

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+X..LPG.Vehicles..NLP G.+GNP.Commerce,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG. + GNP.Commerce

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.4505e+10
2 21 3.3559e+10 1 945922407 0.5919 0.4502
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+X..LPG.Vehicles..NLP G.+GNP.Tota, data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG. + GNP.Total

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.4505e+10

2 21 3.2818e+10 1 1687207241 1.0796 0.3106
```

#PU has the lowest p value. So we update the model by adding PU to model.

```
model<-lm(Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.+Price.of.Unleaded.Gasoline..PU.,data = data2) summary(model)
```

```
Call:
lm(formula = Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters +
```

```
X..LPG.Vehicles..NLPG. + Price.of.Unleaded.Gasoline..PU.,
   data = data2)
Residuals:
 Min 10 Median 30 Max
-68200 -23810 -3509 13864 90507
Coefficients:
                               Estimate Std. Error t value Pr(>|t|)
                               9.780e+05 1.285e+05 7.610 1.82e-07
(Intercept)
                               1.330e+05 2.077e+04 6.405 2.39e-06
s2
                               2.883e+05 2.102e+04 13.715 5.97e-12
s3
                               9.616e+04 2.072e+04
                                                    4.641 0.00014
s4
                              -2.167e+04 3.456e+03 -6.271 3.21e-06
Quarters
X..LPG.Vehicles..NLPG.
                               3.004e-01 9.027e-02 3.328 0.00319
Price.of.Unleaded.Gasoline..PU. -3.458e+02 2.179e+02 -1.587 0.12745
(Intercept)
                              ***
s2
                              ***
s3
                              ***
s4
                              ***
Quarters
X..LPG.Vehicles..NLPG.
                              **
Price.of.Unleaded.Gasoline..PU.
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 38300 on 21 degrees of freedom
 (4 observations deleted due to missingness)
Multiple R-squared: 0.95, Adjusted R-squared: 0.9357
F-statistic: 66.47 on 6 and 21 DF, p-value: 1.435e-12
```

PU looks like insignificant so we want to remove it from the model. But we should check it by removing NLPG from the model to see whether PU is significant or not.

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+Price.of.Unleaded.Gasoline..PU.,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG. + Price.of.Unleaded.Gasoline..PU.

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Unleaded.Gasoline..PU.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 21 3.081e+10

2 22 4.706e+10 -1 -1.625e+10 11.076 0.003193 **
---

Signif. codes: 0 \***' 0.001 \**' 0.01 \*' 0.05 \'.' 0.1 \' 1
```

#p value is so small so we can neither remove NLPG nor add PU to the model.

#Now we will add the variable with second smallest P value obtained from ANOVA which is RNUV

```
model<-lm(Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG.+RNUV,data = data2) summary(model)
```

```
Call:
lm(formula = Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters +
   X..LPG.Vehicles..NLPG. + RNUV, data = data2)
Residuals:
          1Q Median
  Min
                      3Q.
                             Max
-72932 -18011 1721 18130 86399
Coefficients:
                       Estimate Std. Error t value Pr(>|t|)
                      8.799e+05 9.042e+04 9.731 3.12e-09 ***
(Intercept)
                      1.184e+05 2.231e+04 5.308 2.91e-05 ***
s2
                      2.708e+05 2.281e+04 11.870 8.90e-11 ***
s3
                      8.355e+04 2.381e+04
                                           3.509 0.00209 **
s4
                      -2.049e+04 4.197e+03 -4.882 7.93e-05 ***
Quarters
X..LPG.Vehicles..NLPG. 2.148e-01 1.233e-01
                                            1.742 0.09618 .
                      2.377e+06 1.797e+06 1.323 0.20006
RNUV
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 38940 on 21 degrees of freedom
 (4 observations deleted due to missingness)
Multiple R-squared: 0.9483, Adjusted R-squared: 0.9335
F-statistic: 64.19 on 6 and 21 DF, p-value: 2.027e-12
```

#RNUV seems insignificant so we should check to remove NLPG with anova.

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+RNUV,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + X..LPG.Vehicles..NLPG. + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV Res.Df RSS Df Sum of Sq F Pr(>F)

1 21 3.1850e+10

2 22 3.6452e+10 -1 -4601362376 3.0338 0.09618 . ---

Signif. codes: 0 \***' 0.001 \**' 0.05 \'.' 0.1 \' 1
```

#P value is large enough to remove NLPG.

model<-lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+RNUV,data = data2) summary(model)

```
Analysis of Variance Table
Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters +
X..LPG.Vehicles..NLPG. +
   RNUV
Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV
                                     F Pr(>F)
               RSS Df Sum of Sq
 Res.Df
  21 3.1850e+10
      22 3.6452e+10 -1 -4601362376 3.0338 0.09618 .
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
> model<-lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+RNUV,data = data2)
> summary(model)
lm(formula = Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters +
    RNUV, data = data2)
Residuals:
Min 1Q Median 3Q Max -63231 -15788 -4559 27431 83703
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 1033232.2 21482.5 48.096 < 2e-16 ***
          106743.1
                        22241.4 4.799 8.57e-05 ***
            257206.5 22409.0 11.478 9.30e-11 ***
s3
           69115.6 23330.3 2.962 0.00719 **
-13357.6 962.5 -13.878 2.32e-12 ***
s4
Quarters
           4507922.6 1375124.6 3.278 0.00344 **
RNUV
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 40710 on 22 degrees of freedom
  (4 observations deleted due to missingness)
Multiple R-squared: 0.9408, Adjusted R-squared: 0.9274
F-statistic: 69.95 on 5 and 22 DF, p-value: 9.243e-13
```

#RNUV is significant. We will look the ANOVA results to decide if we update the model with any variable.

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+RNUV+Price.of.Diesel .Gasoline..PG.,data = data2))

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+RNUV+X..Unleaded.G asoline.Vehicles..NUGV. ,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV + X..Unleaded.Gasoline.Vehicles..NUGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.6452e+10
2 21 3.6431e+10 1 20673819 0.0119 0.9141
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+RNUV+X..of.Diesel.G asoline.Vehicles..NDGV.,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV + X..of.Diesel.Gasoline.Vehicles..NDGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.6452e+10

2 21 3.4784e+10 1 1668140319 1.0071 0.327
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+RNUV+GNP.Agricultur e,data = data2))

```
Analysis of Variance Table
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+RNUV+GNP.Commer ce,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV + GNP.Commerce

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.6452e+10
2 21 3.4978e+10 1 1473618126 0.8847 0.3576
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+RNUV+GNP.Total,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV + GNP.Total

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 3.6452e+10

2 21 3.5938e+10 1 513355673 0.3 0.5897
```

#PG has the Smallest p value. So we should add the PG to the model

```
model<-lm(Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG.+RNUV,data = data2) summary(model)
```

```
Call:
lm(formula = Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters +
   Price.of.Diesel.Gasoline..PG. + RNUV, data = data2)
Residuals:
          1Q Median
                      30
  Min
                           Max
-50232 -21999 -259 12073 81102
Coefficients:
                             Estimate Std. Error t value Pr(>|t|)
                            1240074.8 88870.8 13.954 4.30e-12 ***
(Intercept)
                             109011.7
                                        20214.8
                                                 5.393 2.39e-05 ***
s2
                             263712.8 20526.6 12.847 2.05e-11 ***
s3
                                       21306.2
                              63616.1
                                                 2.986 0.00705 **
s4
                                       1607.7 -6.306 2.97e-06 ***
                             -10138.3
Quarters
                                          275.1 -2.386 0.02655 *
                              -656.3
Price.of.Diesel.Gasoline..PG.
                            4952340.7 1262268.7
                                                 3.923 0.00078 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 36960 on 21 degrees of freedom
 (4 observations deleted due to missingness)
Multiple R-squared: 0.9534, Adjusted R-squared: 0.9401
F-statistic: 71.67 on 6 and 21 DF, p-value: 6.808e-13
```

#PG is significant but we still want to try removing RNUV from the model to see if it is going better.

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+Price.of.Diesel.Gasoline..PG.,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 21 2.8679e+10

2 22 4.9701e+10 -1 -2.1022e+10 15.393 0.0007802 ***

---

Signif. codes: 0 \***' 0.001 \**' 0.01 \*' 0.05 \'.' 0.1 \' 1
```

#P value is so small so we can not remove RNUV. Now we do anova to see if we can add any more variables to the model.

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+Price.of.Diesel.Gasoline.PG.+RNUV+X..Unleaded.Gasoline.Vehicles..NUGV.,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline.PG. + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline.PG. + RNUV + X..Unleaded.Gasoline.Vehicles..NUGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 21 2.8679e+10
2 20 2.8679e+10 1 432.15 0 0.9996
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+Price.of.Diesel.Gasoline.PG.+RNUV+X..of.Diesel.Gasoline.Vehicles..NDGV.,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV + X..of.Diesel.Gasoline.Vehicles..NDGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 21 2.8679e+10

2 20 2.6662e+10 1 2017243108 1.5132 0.2329
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+Price.of.Diesel.Gasoline..PG.+RNUV+GNP.Agriculture,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV + GNP.Agriculture

Res.Df RSS Df Sum of Sq F Pr(>F)

1 21 2.8679e+10
2 20 2.8326e+10 1 353147552 0.2493 0.623
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+Price.of.Diesel.Gasoline..PG.+RNUV+GNP.Commerce,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV + GNP.Commerce Res.Df RSS Df Sum of Sq F Pr(>F)

1 21 2.8679e+10
2 20 2.6347e+10 1 2332493838 1.7706 0.1983
```

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+Price.of.Diesel.Gasoline..PG.+RNUV+GNP.Total ,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV + GNP.Total
Res.Df RSS Df Sum of Sq F Pr(>F)

1 21 2.8679e+10
2 20 2.8115e+10 1 564508508 0.4016 0.5335
```

#We added the variable with smallest p value, GNPC to the model.

```
model<-lm(Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG.+RNUV+GNP.Commerce,data = data2) summary(model)
```

```
Call:
lm(formula = Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters +
    Price.of.Diesel.Gasoline..PG. + RNUV + GNP.Commerce, data = data2)

Residuals:
    Min    1Q Median    3Q    Max
-55218 -18758    2045    14876    78354

Coefficients:
```

```
Estimate Std. Error t value Pr(>|t|)
                             1.074e+06 1.521e+05 7.062 7.56e-07 ***
(Intercept)
                             6.306e+04 3.984e+04 1.583 0.129115
s2
                             1.277e+05 1.042e+05 1.226 0.234388
s3
                            -3.233e+03 5.442e+04 -0.059 0.953214
s4
                             -1.321e+04 2.800e+03 -4.720 0.000131 ***
Quarters
Price.of.Diesel.Gasoline..PG. -6.957e+02 2.718e+02 -2.560 0.018684 *
                              3.652e+06 1.578e+06 2.314 0.031424 *
                              5.851e-02 4.397e-02 1.331 0.198281
GNP.Commerce
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 36300 on 20 degrees of freedom
 (4 observations deleted due to missingness)
Multiple R-squared: 0.9572, Adjusted R-squared: 0.9423
F-statistic: 63.94 on 7 and 20 DF, p-value: 2.633e-12
```

#GNPC is insignificant yet we still want to check if removing the others will make any difference

#Removing RNUV

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+Price.of.Diesel.Gasoline..PG.+GNP.Commerce,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV + GNP.Commerce

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + GNP.Commerce

Res.Df RSS Df Sum of Sq F Pr(>F)

1 20 2.6347e+10

2 21 3.3401e+10 -1 -7053624722 5.3544 0.03142 * ---

Signif. codes: 0 \***' 0.001 \**' 0.05 \'.' 0.1 \' 1
```

we can not remove RNUV from the model with this smallest p value

#Removing PG

Analysis of Variance Table

we can not remove PG from the model with this smallest p value

#Removing both

anova(model,lm(Unleaded.Gasoline.Sale..UGS.~s2+s3+s4+Quarters+GNP.Commerce,data = data2))

```
Analysis of Variance Table

Model 1: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + Price.of.Diesel.Gasoline..PG. + RNUV + GNP.Commerce

Model 2: Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + GNP.Commerce

Res.Df RSS Df Sum of Sq F Pr(>F)

1 20 2.6347e+10

2 22 4.1322e+10 -2 -1.4975e+10 5.6836 0.01111 *
---

Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

we can not remove PG and RNUV from the model with this smallest p value.

#We did not add GNPC to the model. We should stop there because GNPC which has the best ANOVA result is not significant. So we don't have to try other ones.

```
model<-Im(Unleaded.Gasoline.Sale..UGS. ~ s2 + s3 + s4 + Quarters + RNUV+Price.of.Diesel.Gasoline..PG.,data = data2)
```

summary(model)

```
Residuals:
```

Min 1Q Median 3Q Max -50232 -21999 -259 12073 81102

Coefficients:

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 36960 on 21 degrees of freedom (4 observations deleted due to missingness)

Multiple R-squared: 0.9534, Adjusted R-squared: 0.9401 F-statistic: 71.67 on 6 and 21 DF, p-value: 6.808e-13

#All variables are significant. We have a trend and seasonality variables. R-square values are 0.9534 and 0.9401 and p value is 6.808e-13 which are so satisfying.

#We will move on by checking model assumptions.

UGS model <- model

library(car)

library(Imtest)

vif(UGS_model)

\$2 \$3 1.570896 1.619730 \$4 Quarters 1.745089 3.457702 RNUV Price.of.Diesel.Gasoline..PG. 1.158310 3.486745 #We performed Variance Inflation Factor test to see if there is multicollinearity in our model.

#since there is no value high enough to indicate there exist a multicollinearity in our model

#Then we performed Dublin-Watson test to see if there is autocorrelation

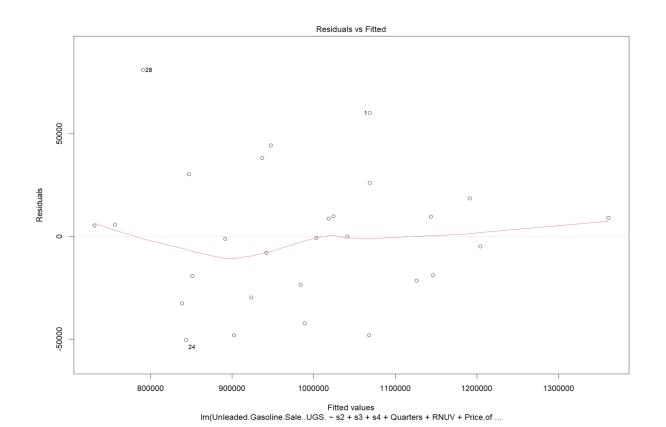
dwtest(UGS_model)

```
Durbin-Watson test

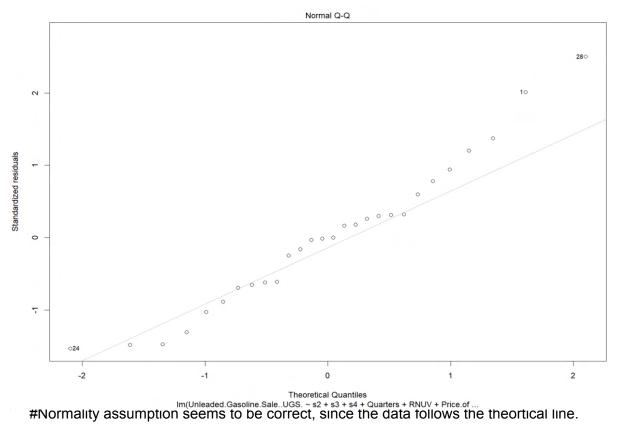
data: UGS_model
DW = 2.1835, p-value = 0.5329
alternative hypothesis: true autocorrelation is greater than 0
```

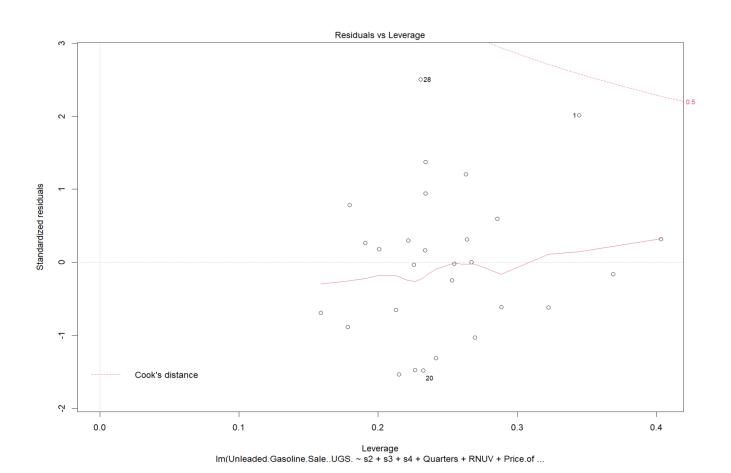
#dw value is so close to 2 and p value is 0.5329. Based on these results, we do not have sufficient evidence to reject the null hypothesis that the true autocorrelation is equal to 0.

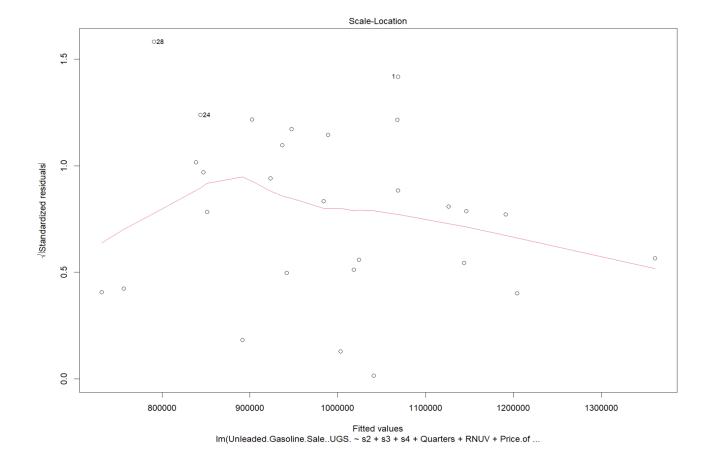
plot(UGS_model)



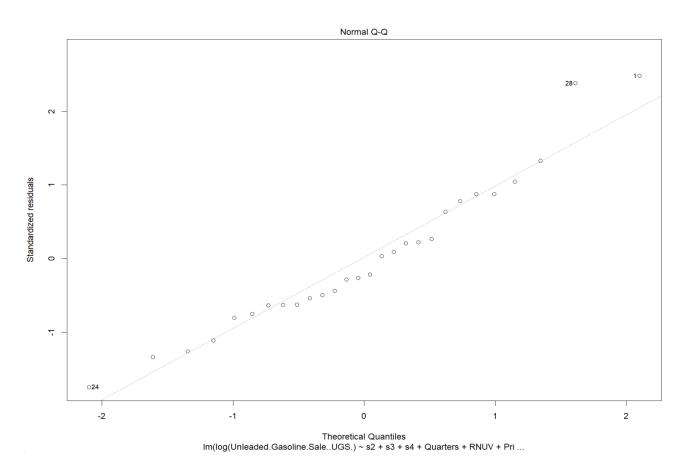
#When we look at the plot we can say that our linearity assumption holds.

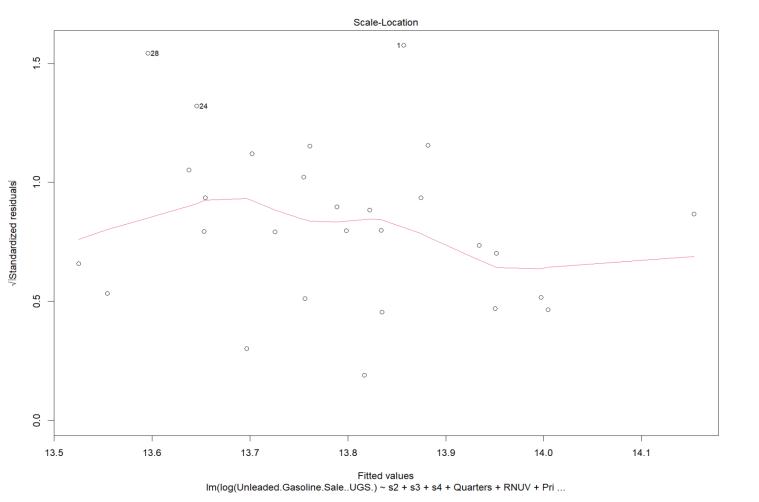


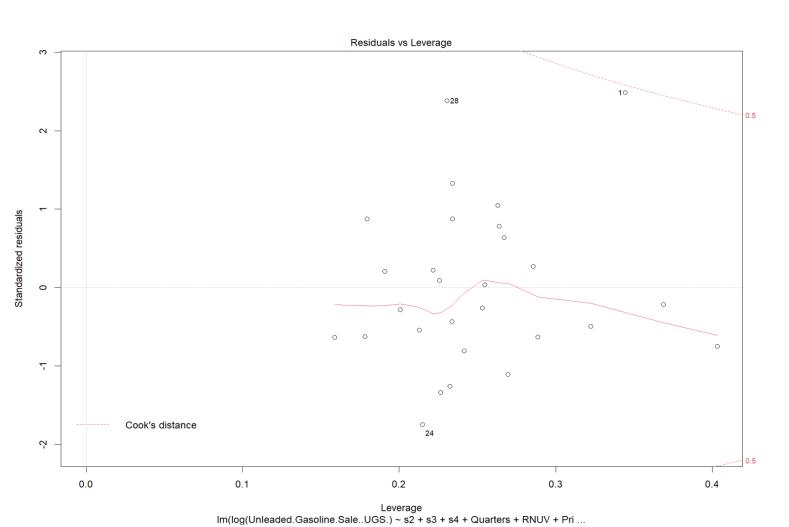


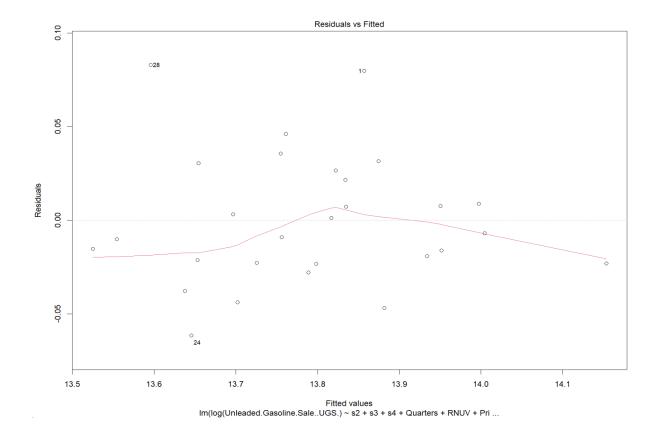


model<-lm(log(Unleaded.Gasoline.Sale..UGS.) ~ s2 + s3 + s4 + Quarters + RNUV+Price.of.Diesel.Gasoline..PG.,data=data2) plot(model)

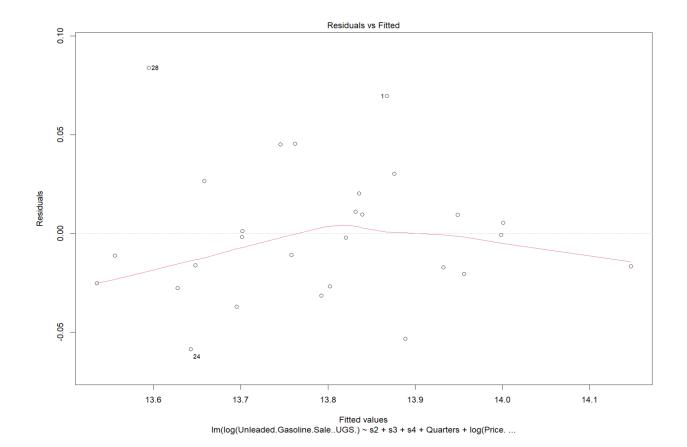




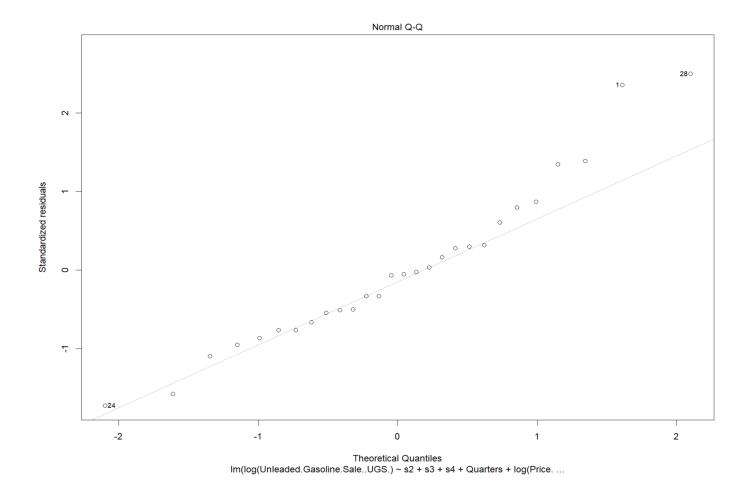




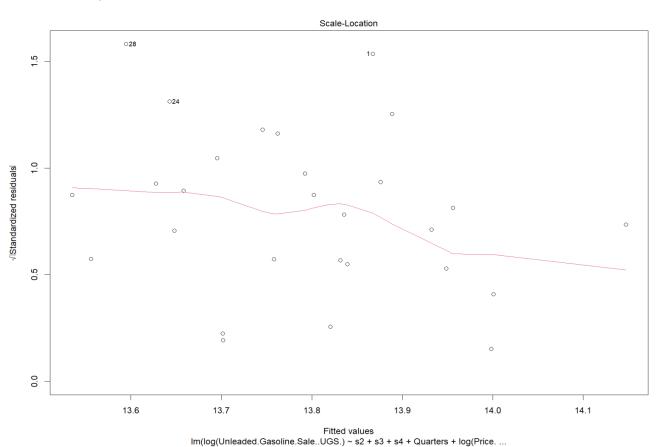
#The average value of standardized residuals appears to remain consistent across the range of fitted values, indicating that the assumption of constant variance is fulfilled. Linearity assumption does not maintain we should take the log of the explaining variables.



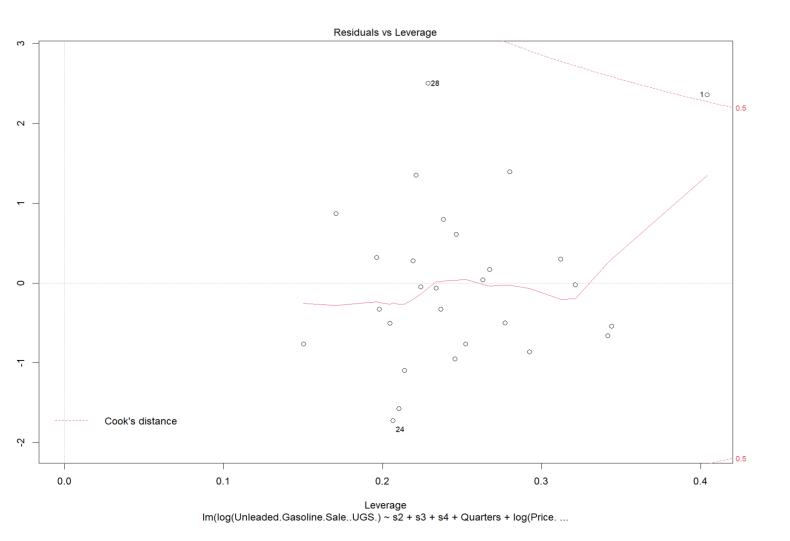
#Linearity assumption is maintained.



#Normality assumption is maintained.



#constant variance is maintained



#After that we should check the correlation between the explanatory variable and the residuals.

```
y_predict<-predict(model,data2)
y_act<-log(data$'Unleaded Gasoline Sale (UGS)')
residuals<- y_act - y_predict
cor(data$'Unleaded Gasoline Sale (UGS)',residuals, use = "complete.obs")</pre>
```

[1] 0.1966006

#Correlation between explanatory variables and residuals is 0.1966 which is insignificant. So our model satisfied the all model assumptions.

#We will make predicts for 2007 based on our model

```
predict2007Q1 <-
```

data.frame(s2=0,s3=0,s4=0,Quarters=29,Price.of.Diesel.Gasoline..PG.=449.1909,RNUV=0.0 07386855)

predict(model,predict2007Q1,level = 0.9,interval = "prediction")

```
fit lwr upr
1 13.49022 13.41497 13.56547
```

exp(13.49022)

[1] 722317.4

predict2007Q2 <-

data.frame(s2=1,s3=0,s4=0,Quarters=30,Price.of.Diesel.Gasoline..PG.=449.1909,RNUV=0.0 10591663)

predict(model,predict2007Q2,level = 0.9,interval = "prediction")

```
fit lwr upr
1 13.6137 13.53811 13.68929
```

exp(13.6137)

[1] 817249.7

predict2007Q3 <-

data.frame(s2=0,s3=1,s4=0,Quarters=31,Price.of.Diesel.Gasoline..PG.=449.1909,RNUV=0.0 10077553)

predict(model,predict2007Q3,level = 0.9,interval = "prediction")

```
fit lwr upr
1 13.74766 13.67124 13.82409
```

exp(13.74766)

[1] 934400.1

predict2007Q4 <data.frame(s2=0,s3=0,s4=1,Quarters=32,Price.of.Diesel.Gasoline..PG.=449.1909,RNUV=0.0 12371491) predict(model,predict2007Q4,level = 0.9,interval = "prediction")

fit lwr upr 1 13.55011 13.47487 13.62535

exp(13.55011)

[1] 766898.7

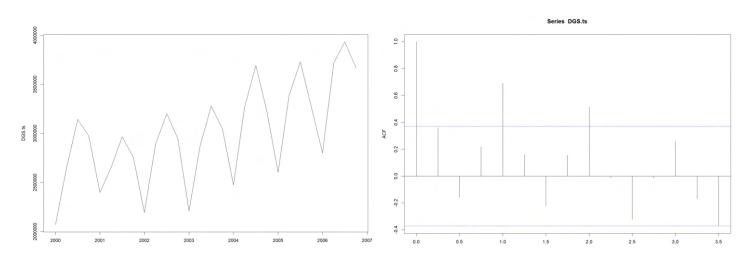
DGS

data<-read_excel("IE360-ProjectData.xlsx",col_names = TRUE)

DGS.ts<- ts(data\$`Diesel Gasoline Sale (DGS)`,start=c(2000,1),end=c(2006,4),frequency=4)

plot(DGS.ts)

acf(DGS.ts)



#data is not stationary

#data seems seasonal

#We should add the trend and seasonality variable to the model

cor(data[sapply(data, is.numeric)],use = "complete.obs")

Unleaded	Gasoline	Sale	(IIGS)
UIITEAUEU	Gasottile	Sale	(000)

Unleaded Gasoline Sale (UGS)	1.0000000
Diesel Gasoline Sale (DGS)	0.04670539
RNUV	0.26498942
# LPG Vehicles (NLPG)	-0.62890683
Price of Unleaded Gasoline (PU)	-0.48993252
Price of Diesel Gasoline (PG)	-0.54812079
# Unleaded Gasoline Vehicles (NUGV)	-0.64862885
# of Diesel Gasoline Vehicles (NDGV)	-0.52526929
GNP Agriculture	0.58815962
GNP Commerce	0.20843693
GNP Total	0.19348598
	Diesel Gasoline Sale (DGS) RNUV
Unleaded Gasoline Sale (UGS)	0.04670539 0.26498942
Diesel Gasoline Sale (DGS)	1.00000000 0.30364955
RNUV	0.30364955 1.00000000
# LPG Vehicles (NLPG)	0.63895652 0.18522216
Price of Unleaded Gasoline (PU)	0.49630315 0.04177338
Price of Diesel Gasoline (PG)	0.54026568 0.07378405
# Unleaded Gasoline Vehicles (NUGV)	0.67288637 0.07413199

GNP Agriculture 0.60365821 0.16530707

GNP Commerce 0.91741674 0.29698204

LPG Vehicles (NLPG)

0.65656807 0.30576056

0.87962217 0.30260566

Unleaded Gasoline Sale (UGS)	-0.62890683
Diesel Gasoline Sale (DGS)	0.63895652
RNUV	0.18522216
# LPG Vehicles (NLPG)	1.00000000
Price of Unleaded Gasoline (PU)	0.67602376
Price of Diesel Gasoline (PG)	0.76400431
# Unleaded Gasoline Vehicles (NUGV)	0.92714875
# of Diesel Gasoline Vehicles (NDGV)	0.90937746

of Diesel Gasoline Vehicles (NDGV)

GNP Total

GNP Agriculture	0.03664584
GNP Commerce	0.49970806
GNP Total	0.49188935
	Price of Unleaded Gasoline (PU)
Unleaded Gasoline Sale (UGS)	-0.48993252
Diesel Gasoline Sale (DGS)	0.49630315
RNUV	0.04177338
# LPG Vehicles (NLPG)	0.67602376
Price of Unleaded Gasoline (PU)	1.0000000
Price of Diesel Gasoline (PG)	0.98211135
# Unleaded Gasoline Vehicles (NUGV)	0.73923142
# of Diesel Gasoline Vehicles (NDGV)	0.68439875
GNP Agriculture	0.11288812
GNP Commerce	0.48075691
GNP Total	0.45775513
	Price of Diesel Gasoline (PG)
Unleaded Gasoline Sale (UGS)	-0.54812079
Diesel Gasoline Sale (DGS)	0.54026568
RNUV	0.07378405
# LPG Vehicles (NLPG)	0.76400431
Price of Unleaded Gasoline (PU)	0.98211135
Price of Diesel Gasoline (PG)	1.0000000
# Unleaded Gasoline Vehicles (NUGV)	0.81214044
# of Diesel Gasoline Vehicles (NDGV)	0.75830059
GNP Agriculture	0.10869816
GNP Commerce	0.51351013
GNP Total	0.49819657
	# Unleaded Gasoline Vehicles (NUGV)
Unleaded Gasoline Sale (UGS)	-0.64862885
Diesel Gasoline Sale (DGS)	0.67288637
RNUV	0.07413199
# LPG Vehicles (NLPG)	0.92714875
Price of Unleaded Gasoline (PU)	0.73923142
Price of Unleaded Gasoline (PU) Price of Diesel Gasoline (PG)	0.73923142 0.81214044

# of Diesel Gasoline Vehicles (NDGV)	0.94565611	
GNP Agriculture	0.06599231	
GNP Commerce	0.51971136	
GNP Total	0.51050156	
	# of Diesel Gasoline Vehicles (NDGV)	
Unleaded Gasoline Sale (UGS)	-0.52526929	
Diesel Gasoline Sale (DGS)	0.65656807	
RNUV	0.30576056	
# LPG Vehicles (NLPG)	0.90937746	
Price of Unleaded Gasoline (PU)	0.68439875	
Price of Diesel Gasoline (PG)	0.75830059	
# Unleaded Gasoline Vehicles (NUGV)	0.94565611	
# of Diesel Gasoline Vehicles (NDGV)	1.00000000	
GNP Agriculture	0.06716321	
GNP Commerce	0.50529894	
GNP Total	0.49132614	
	GNP Agriculture GNP Commerce GNP Total	
Unleaded Gasoline Sale (UGS)	0.58815962 0.2084369 0.1934860	
Diesel Gasoline Sale (DGS)	0.60365821 0.9174167 0.8796222	
RNUV	0.16530707 0.2969820 0.3026057	
# LPG Vehicles (NLPG)	0.03664584 0.4997081 0.4918894	
Price of Unleaded Gasoline (PU)	0.11288812 0.4807569 0.4577551	
Price of Diesel Gasoline (PG)	0.10869816 0.5135101 0.4981966	
# Unleaded Gasoline Vehicles (NUGV)	0.06599231 0.5197114 0.5105016	
# of Diesel Gasoline Vehicles (NDGV)	0.06716321 0.5052989 0.4913261	
GNP Agriculture	1.00000000 0.8170954 0.8343358	
GNP Commerce	0.81709538 1.0000000 0.9866848	
GNP Total	0.83433576 0.9866848 1.0000000	

#We added dummy quarter variables

```
x3<- 1:32 %% 4==3
x4<- 1:32 %% 4==0
data1<-data.frame(data, s2=1*x2,s3=1*x3,s4=1*x4)
data2<-data.frame(data,quarters=1:32,s2=1*x2,s3=1*x3,s4=1*x4)
lm1<- lm(Diesel.Gasoline.Sale..DGS.~s2+s3+s4,data=data1)
summary(lm1)
Call:
lm(formula = Diesel.Gasoline.Sale..DGS. ~ s2 + s3 + s4, data = data1)
Residuals:
   Min 1Q Median 3Q Max
-454639 -203224 -110736 225573 653772
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 2393141 126376 18.937 6.16e-16 ***
            670435
                       178722 3.751 0.000985 ***
s2
           1027346
                       178722 5.748 6.35e-06 ***
s3
            732856 178722 4.101 0.000409 ***
s4
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 334400 on 24 degrees of freedom
  (4 observations deleted due to missingness)
Multiple R-squared: 0.5958, Adjusted R-squared: 0.5453
F-statistic: 11.79 on 3 and 24 DF, p-value: 6.065e-05
```

Im2<- Im(Diesel.Gasoline.Sale..DGS.~s2+s3+s4+quarters,data=data2) summary(Im2)

```
Call:
```

```
lm(formula = Diesel.Gasoline.Sale..DGS. ~ s2 + s3 + s4 + quarters,
    data = data2)
```

Residuals:

```
Min 1Q Median 3Q Max
-193300 -81875 -35990 93354 285163
```

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)

(Intercept) 1937690 70613 27.441 < 2e-16 ***

s2 635400 77580 8.190 2.86e-08 ***

s3 957277 77806 12.303 1.34e-11 ***

s4 627752 78182 8.029 4.03e-08 ***

quarters 35035 3425 10.228 4.98e-10 ***
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1

Residual standard error: 145000 on 23 degrees of freedom (4 observations deleted due to missingness)

Multiple R-squared: 0.9272, Adjusted R-squared: 0.9145 F-statistic: 73.18 on 4 and 23 DF, p-value: 9.646e-13

#Check the largest absolute correlation value with DGS from the table of correlation. GNPC is the largest one.First, add the GNPC to the model

lm3<- lm(Diesel.Gasoline.Sale..DGS.~s2+s3+s4+quarters+GNP.Commerce,data=data2) summary(lm3)

Call:

```
Residuals:
```

```
Min 1Q Median 3Q Max
-192864 -89343 17474 51635 298413
```

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)

(Intercept) 8.735e+05 3.842e+05 2.274 0.0331 *

s2 3.471e+05 1.232e+05 2.817 0.0100 *

s3 1.502e+05 2.956e+05 0.508 0.6163

s4 2.040e+05 1.659e+05 1.230 0.2318

quarters 1.658e+04 7.230e+03 2.293 0.0318 *

GNP.Commerce 3.344e-01 1.192e-01 2.806 0.0103 *

---

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 127200 on 22 degrees of freedom
(4 observations deleted due to missingness)
Multiple R-squared: 0.9464, Adjusted R-squared: 0.9342

F-statistic: 77.63 on 5 and 22 DF, p-value: 3.162e-13

#s3 and s4 have large p value so remove them.

Im3<-Im(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total,data=data2) summary(Im3)

```
Call:
```

```
lm(formula = Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total,
    data = data2)
```

Residuals:

```
Min 1Q Median 3Q Max
-258030 -91264 -37451 58326 395684
```

```
Coefficients:
```

```
Estimate Std. Error t value Pr(>|t|)

(Intercept) 6.090e+05 1.606e+05 3.792 0.000891 ***

s2 3.528e+05 6.814e+04 5.177 2.66e-05 ***

quarters 1.397e+04 4.081e+03 3.424 0.002221 **

GNP.Total 8.359e-02 6.938e-03 12.049 1.15e-11 ***

---

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1

Residual standard error: 149100 on 24 degrees of freedom

(4 observations deleted due to missingness)

Multiple R-squared: 0.9197, Adjusted R-squared: 0.9096

F-statistic: 91.59 on 3 and 24 DF, p-value: 2.801e-13
```

model<-lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total,data=data2)

#Now we do anova to see if we can add any more variables to the model anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+RNUV,data=data2))

```
Analysis of Variance Table
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+X..LPG.Vehicles..NLP G.,data=data2))

```
Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total
Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total +
X..LPG.Vehicles..NLPG.
Res.Df RSS Df Sum of Sq F Pr(>F)

24 5.3326e+11
2 23 5.1363e+11 1 1.9629e+10 0.879 0.3582
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Unleaded.Gasoline..PU.,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Unleaded.Gasoline..PU.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 24 5.3326e+11

2 23 4.2717e+11 1 1.0609e+11 5.7125 0.02543 *

---

Signif. codes: 0 \***' 0.001 \**' 0.01 \*' 0.05 \'.' 0.1 \' 1
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Diesel.Gasoline..PG.,data=data2))

```
Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total +
Price.of.Diesel.Gasoline..PG.

Res.Df RSS Df Sum of Sq F Pr(>F)
```

Analysis of Variance Table

```
24 5.3326e+11
   23 4.1300e+11 1 1.2026e+11 6.6971 0.01645 *
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+X..Unleaded.Gasoline.
Vehicles..NUGV.,data=data2))
Analysis of Variance Table
Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total
Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total +
X..Unleaded.Gasoline.Vehicles..NUGV.
              RSS Df Sum of Sq F Pr(>F)
 Res.Df
1
     24 5.3326e+11
      23 4.2182e+11 1 1.1144e+11 6.076 0.0216 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+X..of.Diesel.Gasoline.
Vehicles..NDGV.,data=data2))
Analysis of Variance Table
Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total
Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total +
X..of.Diesel.Gasoline.Vehicles..NDGV.
               RSS Df Sum of Sq F Pr(>F)
 Res.Df
     24 5.3326e+11
      23 4.5828e+11 1 7.4978e+10 3.763 0.06476 .
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+GNP.Agriculture,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + GNP.Agriculture

Res.Df RSS Df Sum of Sq F Pr(>F)

1 24 5.3326e+11

2 23 5.3039e+11 1 2871620039 0.1245 0.7274
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+GNP.Commerce,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + GNP.Commerce

Res.Df RSS Df Sum of Sq F Pr(>F)

1 24 5.3326e+11

2 23 4.2975e+11 1 1.0351e+11 5.5401 0.0275 *

---

Signif. codes: 0 \***' 0.001 \**' 0.01 \*' 0.05 \'.' 0.1 \' 1
```

#We selected to PG which has the smallest p-value.

model<-lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Diesel.Gasoline..PG.,data=data2)

summary(model)

```
Call:
lm(formula = Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total +
   Price.of.Diesel.Gasoline..PG., data = data2)
Residuals:
   Min
           1Q Median
                           3Q
                                    Max
-217662 -76646 -10177 88420 314403
Coefficients:
                               Estimate Std. Error t value Pr(>|t|)
                              1.358e+06 3.234e+05 4.199 0.000343 ***
(Intercept)
                              3.691e+05 6.158e+04 5.994 4.12e-06 ***
s2
                               2.538e+04 5.733e+03 4.426 0.000195 ***
quarters
                               8.658e-02 6.343e-03 13.650 1.62e-12 ***
GNP.Total
Price.of.Diesel.Gasoline..PG. -2.524e+03 9.755e+02 -2.588 0.016450 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 134000 on 23 degrees of freedom
  (4 observations deleted due to missingness)
Multiple R-squared: 0.9378, Adjusted R-squared: 0.927
F-statistic: 86.67 on 4 and 23 DF, p-value: 1.589e-13
#PG is indeed significant. But we still want to check if removing GNPT increases the
significance of PG and therefore improves the model.
Analysis of Variance Table
Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total +
Price.of.Diesel.Gasoline..PG.
Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG.
```

RSS Df Sum of Sq F Pr(>F)

24 3.7585e+12 -1 -3.3455e+12 186.31 1.623e-12 ***

Res.Df

23 4.1300e+11

1

```
---
```

```
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
```

#We can not remove the GNPT from the model.

#So check anova results to determine variable which will be add

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Diesel.Gasoline..PG.+RNUV,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG. +

RNUV

Res.Df RSS Df Sum of Sq F Pr(>F)

1 23 4.1300e+11

2 22 3.9326e+11 1 1.9745e+10 1.1046 0.3047
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Diesel.Gasoline..PG.+X..LPG.Vehicles..NLPG.,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG. +

X..LPG.Vehicles..NLPG.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 23 4.1300e+11

2 22 4.0666e+11 1 6340148042 0.343 0.5641
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Diesel.Gasoline..PG.+Price.of.Unleaded.Gasoline..PU.,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG. +

Price.of.Unleaded.Gasoline..PU.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 23 4.1300e+11

2 22 4.0922e+11 1 3783654094 0.2034 0.6564
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Diesel.Gasoline.PG.+X..Unleaded.Gasoline.Vehicles..NUGV.

,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG. +

X..Unleaded.Gasoline.Vehicles..NUGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 23 4.1300e+11

2 22 2.8833e+11 1 1.2467e+11 9.5124 0.005421 **

---

Signif. codes: 0 `***' 0.001 `**' 0.01 `*' 0.05 `.' 0.1 `' 1
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Diesel.Gasoline.PG.+X..of.Diesel.Gasoline.Vehicles..NDGV.,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline.PG.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline.PG. +

X..of.Diesel.Gasoline.Vehicles..NDGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 23 4.1300e+11

2 22 3.1641e+11 1 9.6596e+10 6.7164 0.01665 *

---

Signif. codes: 0 `***' 0.001 `**' 0.01 `*' 0.05 `.' 0.1 `' 1
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Diesel.Gasoline..PG.+GNP.Agriculture,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG. +

GNP.Agriculture

Res.Df RSS Df Sum of Sq F Pr(>F)

1 23 4.1300e+11

2 22 4.0633e+11 1 6668217786 0.361 0.5541
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Diesel.Gasoline..PG.+GNP.Commerce,data=data2))

```
Analysis of Variance Table
Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total +
Price.of.Diesel.Gasoline..PG.
Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total +
Price.of.Diesel.Gasoline..PG. +
   GNP.Commerce
 Res.Df RSS Df Sum of Sq F Pr(>F)
1
    23 4.130e+11
     22 2.794e+11 1 1.3361e+11 10.52 0.003729 **
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
#GNPC is selected because the smallest p-value is in GNPC
model<-lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+Price.of.Diesel.Gasoline..P
G.+GNP.Commerce,data=data2)
summary(model)
Call:
lm(formula = Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total +
   Price.of.Diesel.Gasoline..PG. + GNP.Commerce, data = data2)
Residuals:
   Min 1Q Median 3Q
                                 Max
-204125 -85488 -14629 92439 178383
Coefficients:
                              Estimate Std. Error t value Pr(>|t|)
                             1.611e+06 2.830e+05 5.694 9.99e-06 ***
(Intercept)
```

```
2.307e+05 6.711e+04 3.438 0.00235 **

quarters 2.690e+04 4.845e+03 5.553 1.40e-05 ***

GNP.Total -2.817e-02 3.578e-02 -0.787 0.43946

Price.of.Diesel.Gasoline..PG. -2.843e+03 8.262e+02 -3.441 0.00233 **

GNP.Commerce 5.409e-01 1.668e-01 3.243 0.00373 **

---

Signif. codes: 0 \***' 0.001 \**' 0.01 \*' 0.05 \'.' 0.1 \' 1

Residual standard error: 112700 on 22 degrees of freedom

(4 observations deleted due to missingness)

Multiple R-squared: 0.9579, Adjusted R-squared: 0.9483

F-statistic: 100.1 on 5 and 22 DF, p-value: 2.232e-14
```

#GNPT seems insignificant we should do ANOVA to determine remove or not #We still want to check for removing both GNPT or PG

#For PG:

Analysis of Variance Table

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Total+GNP.Commerce,data=data2))

Since p value is quite small, we can not remove PG

#For GNPT:

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Total + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 2.7940e+11

2 23 2.8727e+11 -1 -7873681465 0.62 0.4395
```

Due to the high p value, we can remove GNPT

model<-lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GNP.Commerce,data=data2)

summary(model)

All variables are significant, so check anova results to determine variable which will be add.

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce+RNUV,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + RNUV

Res.Df RSS Df Sum of Sq F Pr(>F)

1 23 2.8727e+11

2 22 2.5981e+11 1 2.746e+10 2.3252 0.1415
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce+X..LPG.Vehicles..NLPG.,data=data2))

```
Analysis of Variance Table
Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +
    GNP.Commerce
Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +
    GNP.Commerce + X..LPG.Vehicles..NLPG.
 Res.Df
              RSS Df Sum of Sq F Pr(>F)
     23 2.8727e+11
1
      22 2.8655e+11 1 717964400 0.0551 0.8166
anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+guarters+Price.of.Diesel.Gasoline..PG.+GN
P.Commerce+Price.of.Unleaded.Gasoline..PU.,data=data2))
Analysis of Variance Table
Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +
    GNP.Commerce
Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +
    GNP.Commerce + Price.of.Unleaded.Gasoline..PU.
 Res.Df
               RSS Df Sum of Sq
                                     F Pr(>F)
     23 2.8727e+11
      22 2.8529e+11 1 1979616563 0.1527 0.6998
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce+X..Unleaded.Gasoline.Vehicles..NUGV.,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 23 2.8727e+11

2 22 2.0453e+11 1 8.2744e+10 8.9004 0.006854 **

---

Signif. codes: 0 \***' 0.001 \**' 0.01 \*' 0.05 \'.' 0.1 \' 1
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce+X..of.Diesel.Gasoline.Vehicles..NDGV.,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + X..of.Diesel.Gasoline.Vehicles..NDGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 23 2.8727e+11

2 22 2.1980e+11 1 6.7466e+10 6.7526 0.0164 *

---

Signif. codes: 0 \***' 0.001 \**' 0.01 \*' 0.05 \'' 0.1 \'' 1
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce+GNP.Agriculture,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + GNP.Agriculture

Res.Df RSS Df Sum of Sq F Pr(>F)

1 23 2.8727e+11

2 22 2.6199e+11 1 2.5278e+10 2.1226 0.1593
```

#select the variable which has a smallest p-value. NUGV is the smallest one.

model<-lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG. + GNP.Commerce+X..Unleaded.Gasoline.Vehicles..NUGV. ,data=data2)

summary(model)

```
Call:
```

```
lm(formula = Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV., data = data2)
```

Residuals:

```
Min 1Q Median 3Q Max
-144167 -71433 -17206 76505 121482
```

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)

(Intercept) -1.844e+06 1.158e+06 -1.592 0.125676

s2 2.730e+05 4.299e+04 6.351 2.16e-06

quarters 1.118e+03 9.429e+03 0.119 0.906685

Price.of.Diesel.Gasoline..PG. -2.881e+03 7.048e+02 -4.088 0.000486
```

```
4.052e-01 2.136e-02 18.970 4.01e-15
GNP.Commerce
X..Unleaded.Gasoline.Vehicles..NUGV. 7.344e-01 2.462e-01 2.983 0.006854
(Intercept)
                                    * * *
quarters
Price.of.Diesel.Gasoline..PG.
GNP.Commerce
X..Unleaded.Gasoline.Vehicles..NUGV. **
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Residual standard error: 96420 on 22 degrees of freedom
  (4 observations deleted due to missingness)
Multiple R-squared: 0.9692, Adjusted R-squared: 0.9622
F-statistic: 138.4 on 5 and 22 DF, p-value: 7.336e-16
#All variables seem significant but we do ANOVA to remove any variable.
anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+X..
Unleaded.Gasoline.Vehicles..NUGV.,data=data2))
Analysis of Variance Table
Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +
   GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV.
Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +
   X..Unleaded.Gasoline.Vehicles..NUGV.
 Res.Df
        RSS Df Sum of Sq F Pr(>F)
     22 2.0453e+11
2
     23 3.5502e+12 -1 -3.3456e+12 359.87 4.01e-15 ***
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

#We can not remove GNPC.

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Commerce+X..Unleaded.Gasoline.Vehicles..NUGV.,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline.PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 2.0453e+11

2 23 3.5993e+11 -1 -1.554e+11 16.715 0.0004861 ***
---

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

#We can not remove PG.

#Now determine the next variable to add the model.

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce+X..Unleaded.Gasoline.Vehicles..NUGV.

```
+RNUV,data=data2))
```

Analysis of Variance Table

```
Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +

    GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +

    GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV. + RNUV
    Res.Df    RSS Df Sum of Sq    F Pr(>F)

1     22 2.0453e+11
2     21 1.9579e+11    1 8734607531 0.9368 0.3441
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce+X..Unleaded.Gasoline.Vehicles..NUGV.

+X..LPG.Vehicles..NLPG.,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV. + X..LPG.Vehicles..NLPG.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 2.0453e+11

2 21 2.0426e+11 1 266700287 0.0274 0.8701
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce+X..Unleaded.Gasoline.Vehicles..NUGV. + Price.of.Unleaded.Gasoline..PU.,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV. + Price.of.Unleaded.Gasoline..PU.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 2.0453e+11

2 21 2.0453e+11 1 986634 1e-04 0.9921
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce+X..Unleaded.Gasoline.Vehicles..NUGV. + X..of.Diesel.Gasoline.Vehicles..NDGV.,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline..PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV. + X..of.Diesel.Gasoline.Vehicles..NDGV.

Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 2.0453e+11

2 21 2.0243e+11 1 2.093e+09 0.2171 0.646
```

anova(model,lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce+X..Unleaded.Gasoline.Vehicles..NUGV. +GNP.Agriculture,data=data2))

```
Analysis of Variance Table

Model 1: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline.PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV.

Model 2: Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + Price.of.Diesel.Gasoline.PG. +

GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV. + GNP.Agriculture Res.Df RSS Df Sum of Sq F Pr(>F)

1 22 2.0453e+11

2 21 1.9172e+11 1 1.2809e+10 1.4031 0.2494
```

#We checked all the variables and they cannot be added to the model because of their ANOVA results, no variable seems significant due to high p values.

model<-lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GNP.Commerce+X..Unleaded.Gasoline.Vehicles..NUGV.

```
,data=data2)
```

summary(model)

```
Call:
lm(formula = Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +
   GNP.Commerce + X..Unleaded.Gasoline.Vehicles..NUGV., data = data2)
Residuals:
   Min 1Q Median 3Q
                                  Max
-144167 -71433 -17206 76505 121482
Coefficients:
                                     Estimate Std. Error t value Pr(>|t|)
(Intercept)
                                   -1.844e+06 1.158e+06 -1.592 0.125676
                                    2.730e+05 4.299e+04 6.351 2.16e-06
s2
quarters
                                    1.118e+03 9.429e+03 0.119 0.906685
Price.of.Diesel.Gasoline..PG.
                                   -2.881e+03 7.048e+02 -4.088 0.000486
                                   4.052e-01 2.136e-02 18.970 4.01e-15
GNP.Commerce
X..Unleaded.Gasoline.Vehicles..NUGV. 7.344e-01 2.462e-01 2.983 0.006854
(Intercept)
s2
quarters
Price.of.Diesel.Gasoline..PG.
                                   ***
GNP.Commerce
X..Unleaded.Gasoline.Vehicles..NUGV. **
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 '' 1
```

Residual standard error: 96420 on 22 degrees of freedom

```
(4 observations deleted due to missingness)
Multiple R-squared: 0.9692, Adjusted R-squared: 0.9622
F-statistic: 138.4 on 5 and 22 DF, p-value: 7.336e-16
#All variables are significant but NUGV and PG are correlated. So, we should remove one of
them.
model1<-Im(Diesel.Gasoline.Sale..DGS.~s2+quarters+Price.of.Diesel.Gasoline..PG.+GNP.C
ommerce,data=data2)
summary(model1)
Call:
lm(formula = Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +
   GNP.Commerce, data = data2)
Residuals:
   Min 1Q Median 3Q
                                   Max
-204583 -71473 -20736 89535 180997
Coefficients:
                              Estimate Std. Error t value Pr(>|t|)
                              1.543e+06 2.670e+05 5.778 6.92e-06 ***
(Intercept)
                              2.658e+05 4.975e+04 5.343 2.00e-05 ***
s2
                              2.643e+04 4.768e+03 5.544 1.22e-05 ***
quarters
Price.of.Diesel.Gasoline..PG. -2.784e+03 8.160e+02 -3.412 0.00239 **
                              4.111e-01 2.466e-02 16.671 2.45e-14 ***
GNP.Commerce
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 111800 on 23 degrees of freedom
  (4 observations deleted due to missingness)
Multiple R-squared: 0.9567, Adjusted R-squared: 0.9492
```

F-statistic: 127.1 on 4 and 23 DF, p-value: 2.489e-15

model2<-lm(Diesel.Gasoline.Sale..DGS.~s2+quarters+GNP.Commerce+X..Unleaded.Gasoline.Vehicles..NUGV. ,data=data2)

summary(model2)

```
Call:
lm(formula = Diesel.Gasoline.Sale..DGS. ~ s2 + quarters + GNP.Commerce +
   X..Unleaded.Gasoline.Vehicles..NUGV., data = data2)
Residuals:
   Min
           1Q Median
                          3Q
                                   Max
-222507 -96935 9087 82912 199487
Coefficients:
                                      Estimate Std. Error t value Pr(>|t|)
                                    -2.481e+06 1.489e+06 -1.666 0.109203
(Intercept)
                                     2.577e+05 5.556e+04 4.638 0.000115
s2
                                    -1.019e+04 1.170e+04 -0.871 0.392684
quarters
                                    3.885e-01 2.720e-02 14.282 6.36e-13
GNP.Commerce
X..Unleaded.Gasoline.Vehicles..NUGV. 6.877e-01 3.190e-01 2.156 0.041809
(Intercept)
s2
quarters
GNP.Commerce
X..Unleaded.Gasoline.Vehicles..NUGV. *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 125100 on 23 degrees of freedom
  (4 observations deleted due to missingness)
Multiple R-squared: 0.9458, Adjusted R-squared: 0.9363
F-statistic: 100.3 on 4 and 23 DF, p-value: 3.293e-14
```

#We should remove the NUGV

model<-model1

summary(model)

```
Call:
```

```
lm(formula = Diesel.Gasoline.Sale..DGS. ~ s2 + quarters +
Price.of.Diesel.Gasoline..PG. +

GNP.Commerce, data = data2)
```

Residuals:

```
Min 1Q Median 3Q Max
-204583 -71473 -20736 89535 180997
```

Coefficients:

```
Residual standard error: 111800 on 23 degrees of freedom
(4 observations deleted due to missingness)

Multiple R-squared: 0.9567, Adjusted R-squared: 0.9492

F-statistic: 127.1 on 4 and 23 DF, p-value: 2.489e-15
```

library(Imtest)

library(car)

Now we do Model Analysis with Durbin-Watson test.

dwtest(model)

Durbin-Watson test

data: model

DW = 1.8296, p-value = 0.2319

alternative hypothesis: true autocorrelation is greater than 0

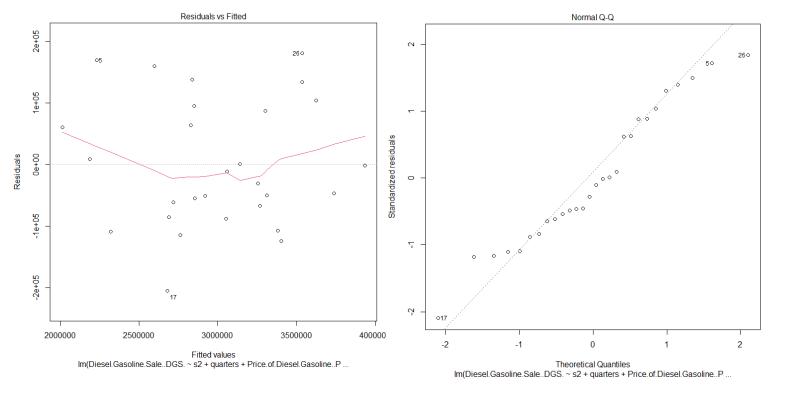
Residuals are independent and there is no autocorrelation #Now we perform vif test to check collinearity

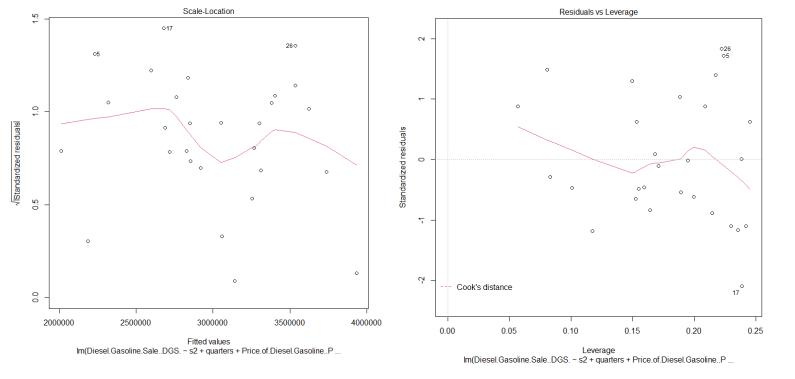
vif(model)

s2 quarters
1.040329 3.324759
Price.of.Diesel.Gasoline..PG. GNP.Commerce
3.354455 1.454041

#No collinearity problem

#Now we should check model assumptions.

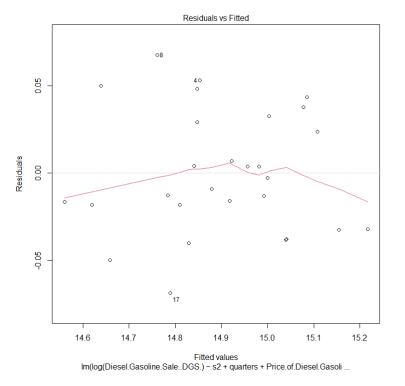


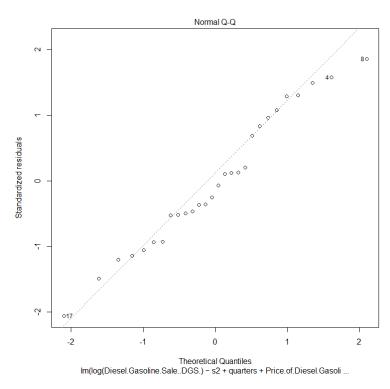


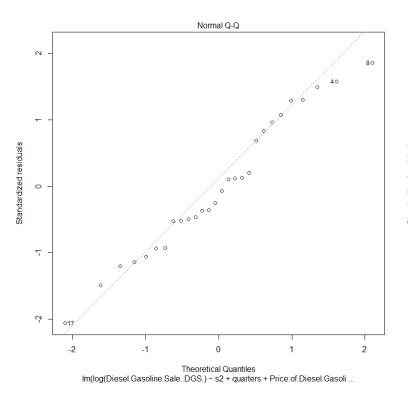
model<-lm(log(Diesel.Gasoline.Sale..DGS.)~s2+quarters+Price.of.Diesel.Gasoline..PG.+GN P.Commerce,data=data2)

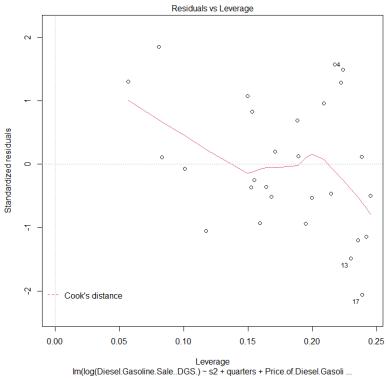
summary(model)

```
Call:
lm(formula = log(Diesel.Gasoline.Sale..DGS.) ~ s2 + quarters +
   Price.of.Diesel.Gasoline..PG. + GNP.Commerce, data = data2)
Residuals:
    Min 1Q Median 3Q
                                      Max
-0.06873 -0.02161 -0.00580 0.03005 0.06788
Coefficients:
                             Estimate Std. Error t value Pr(>|t|)
                             1.443e+01 9.130e-02 158.047 < 2e-16 ***
(Intercept)
                             9.932e-02 1.701e-02 5.838 5.99e-06 ***
s2
quarters
                             9.015e-03 1.630e-03 5.530 1.27e-05 ***
Price.of.Diesel.Gasoline..PG. -1.062e-03 2.791e-04 -3.806 0.000909 ***
                             1.433e-07 8.432e-09 16.996 1.62e-14 ***
GNP.Commerce
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 '' 1
Residual standard error: 0.03822 on 23 degrees of freedom
  (4 observations deleted due to missingness)
Multiple R-squared: 0.9567, Adjusted R-squared: 0.9492
F-statistic: 127.2 on 4 and 23 DF, p-value: 2.482e-15
```





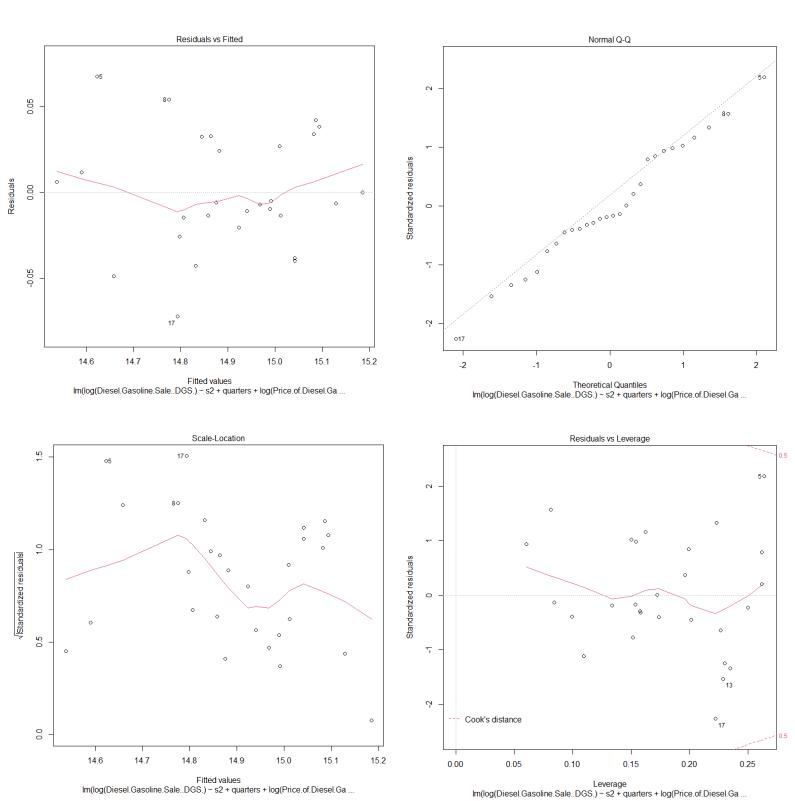




#We should do log transformations to other variables because assumptions are not satisfied.

```
Call:
lm(formula = log(Diesel.Gasoline.Sale..DGS.) ~ s2 + quarters +
   log(Price.of.Diesel.Gasoline..PG.) + log(GNP.Commerce), data = data2)
Residuals:
     Min 1Q Median 3Q
                                      Max
-0.072111 -0.016057 -0.006014 0.028559 0.067776
Coefficients:
                              Estimate Std. Error t value Pr(>|t|)
(Intercept)
                              6.208686 0.750343 8.274 2.39e-08 ***
                              s2
                              0.008971 0.001509 5.945 4.63e-06 ***
quarters
log(Price.of.Diesel.Gasoline..PG.) -0.413934 0.099120 -4.176 0.000363 ***
                              log(GNP.Commerce)
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1
Residual standard error: 0.03613 on 23 degrees of freedom
 (4 observations deleted due to missingness)
Multiple R-squared: 0.9613, Adjusted R-squared: 0.9546
F-statistic: 143 on 4 and 23 DF, p-value: 6.83e-16
```

plot(model)



#Now constant variance and normality assumptions seem better in this model.

```
model < -lm(Diesel. Gasoline. Sale.. DGS. \sim s2 + quarters + Price. of. Diesel. Gasoline.. PG. + GNP. Commerce, data = data2)
```

modelDGS<-model

#Correlation between residuals and explanatory variable

```
y_predict<-predict(model,data2)
y_act<-data$`Diesel Gasoline Sale (DGS)`
residuals<- y_act - y_predict
cor(data$`Diesel Gasoline Sale (DGS)`,residuals, use = "complete.obs")
[1] 0.2080289</pre>
```

#There is no correlation between residuals and explanatory variables.

model<-lm(log(Diesel.Gasoline.Sale..DGS.)~s2+quarters+log(Price.of.Diesel.Gasoline..PG.) +log(GNP.Commerce),data=data2)

```
predict2007Q1 <-
```

data.frame(s2=0,quarters=29,Price.of.Diesel.Gasoline..PG.=449.1909,GNP.Commerce=4857 305,12539562)

predict(model,predict2007Q1,level = 0.9,interval = "prediction")

```
fit lwr upr
1 14.9292 14.86063 14.99777
```

exp(14.9292)

[1] 3045574

```
predict2007Q2 <-
```

data.frame(s2=1,quarters=30,Price.of.Diesel.Gasoline..PG.=449.1909,GNP.Commerce=5852 403,78328151)

predict(model,predict2007Q2,level = 0.9,interval = "prediction")

```
fit lwr upr
1 15.15688 15.08596 15.22779

exp(15.15688)
```

[1] 3824277

predict2007Q3 <-

data.frame(s2=0,quarters=31,Price.of.Diesel.Gasoline..PG.=449.1909,GNP.Commerce=7480 414,37162258)

predict(model,predict2007Q3,level = 0.9,interval = "prediction")

fit lwr upr 1 15.25533 15.18549 15.32516

exp(15.25533)

[1] 4219934

predict2007Q4 <-

data.frame(s2=0,quarters=32,Price.of.Diesel.Gasoline..PG.=449.1909,GNP.Commerce=6397 744,76104164)

predict(model,predict2007Q4,level = 0.9,interval = "prediction")

fit lwr upr
1 15.15272 15.08332 15.22211

exp(15.15272)

[1] 3808401

FORECASTS

UGS	Method A	Method B
2007_Q1	678288	722317
2007_Q2	896475	817249
2007_Q3	948056	934400
2007_Q4	826669	766898

DGS	Method A	Method B
2007_Q1	3105950	3045574
2007_Q2	3968121	3824277
2007_Q3	4087278	4219934
2007_Q4	3808176	3808401

COMPARISON & CONCLUSION

SARIMA is preferred when the focus is on modeling and capturing the inherent temporal patterns and autocorrelation in the time series. Time series regression is more suitable when the goal is to understand the influence of external factors and covariates on the time series behavior.

UGS has a decreasing trend, that's why Method B is overestimating UGS. Similarly, DGS has an increasing trend and Method B is underestimating DGS. It is observed that both UGS and DGS have seasonality. Method A includes seasonal and regular differencing factors, this is the reason that Method A gives better results than Method B for this data. That's why Method A is used for forecasting this data.

In summary, SARIMA is commonly used for forecasting time series with seasonality and autocorrelation, while time series regression is useful when there are external factors that can explain the variation in the time series. The choice depends on the specific characteristics and objectives of the analysis.