

PracticeMidterm

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R Markdown

```
data = read.csv("/Users/jim/Dropbox (GaTech)/Courses/ISyE6402/Midterm/HouseData.csv", header=TRUE)
data.train = data[1:(nrow(data)-4),]
data.test = data[(nrow(data)-3):nrow(data),]
data.train
```

##	Sno	Timeline	Houses_for_Sale	Employment_Ratio	GDP_Growth	Bank_Prime_Rate
## 1	1	Mar-76	314	56.7	3.1	6.75
## 2	2	Jun-76	336	56.8	2.1	7.20
## 3	3	Sep-76	341	56.9	3.0	7.00
## 4	4	Dec-76	358	57.0	4.7	6.35
## 5	5	Mar-77	353	57.4	8.1	6.25
## 6	6	Jun-77	369	57.9	7.3	6.75
## 7	7	Sep-77	392	58.1	0.0	7.13
## 8	8	Dec-77	408	58.7	1.4	7.75
## 9	9	Mar-78	398	58.8	16.5	8.00
## 10	10	Jun-78	422	59.5	4.0	8.63
## 11	11	Sep-78	419	59.5	5.5	9.41
## 12	12	Dec-78	419	59.8	0.8	11.55
## 13	13	Mar-79	416	60.0	0.5	11.75
## 14	14	Jun-79	420	59.9	2.9	11.65
## 15	15	Sep-79	414	60.0	1.0	12.90
## 16	16	Dec-79	402	60.1	1.3	15.30
## 17	17	Mar-80	369	59.7	-7.9	18.31
## 18	18	Jun-80	341	58.9	-0.6	12.63
## 19	19	Sep-80	339	58.9	7.6	12.23
## 20	20	Dec-80	342	59.0	8.5	20.35
## 21	21	Mar-81	323	59.4	-2.9	18.05
## 22	22	Jun-81	318	59.0	4.7	20.03
## 23	23	Sep-81	306	58.7	-4.6	20.08
## 24	24	Dec-81	278	58.2	-6.5	15.75
## 25	25	Mar-82	264	58.1	2.2	16.50
## 26	26	Jun-82	253	57.8	-1.4	16.50
## 27	27	Sep-82	249	57.6	0.4	13.50
## 28	28	Dec-82	255	57.2	5.3	11.50
## 29	29	Mar-83	259	57.1	9.4	10.50
## 30	30	Jun-83	283	57.8	8.1	10.50
## 31	31	Sep-83	302	58.4	8.5	11.00
## 32	32	Dec-83	304	58.8	8.2	11.00

## 33	33	Mar-84	316	59.1	7.2	11.21
## 34	34	Jun-84	340	59.9	4.0	12.60
## 35	35	Sep-84	345	59.7	3.2	12.97
## 36	36	Dec-84	358	59.9	4.0	11.06
## 37	37	Mar-85	350	60.2	3.7	10.50
## 38	38	Jun-85	354	59.8	6.4	9.78
## 39	39	Sep-85	352	60.3	3.0	9.50
## 40	40	Dec-85	350	60.4	3.8	9.50
## 41	41	Mar-86	331	60.5	1.9	9.10
## 42	42	Jun-86	341	60.7	4.1	8.50
## 43	43	Sep-86	357	60.8	2.1	7.50
## 44	44	Dec-86	361	61.0	2.8	7.50
## 45	45	Mar-87	352	61.2	4.6	7.50
## 46	46	Jun-87	360	61.4	3.7	8.25
## 47	47	Sep-87	364	61.6	6.8	8.70
## 48	48	Dec-87	370	62.0	2.3	8.75
## 49	49	Mar-88	366	61.9	5.4	8.50
## 50	50	Jun-88	367	62.3	2.3	9.00
## 51	51	Sep-88	363	62.4	5.4	10.00
## 52	52	Dec-88	371	62.6	4.1	10.50
## 53	53	Mar-89	371	62.9	3.2	11.50
## 54	54	Jun-89	378	63.0	3.0	11.07
## 55	55	Sep-89	366	62.8	0.9	10.50
## 56	56	Dec-89	366	63.0	4.5	10.50
## 57	57	Mar-90	358	63.2	1.6	10.00
## 58	58	Jun-90	355	62.9	0.1	10.00
## 59	59	Sep-90	339	62.5	-3.4	10.00
## 60	60	Dec-90	321	62.2	-1.9	10.00
## 61	61	Mar-91	304	61.8	3.1	9.00
## 62	62	Jun-91	296	61.7	1.9	8.50
## 63	63	Sep-91	292	61.6	1.8	8.20
## 64	64	Dec-91	284	61.2	4.8	7.21
## 65	65	Mar-92	275	61.5	4.5	6.50
## 66	66	Jun-92	275	61.5	3.9	6.50
## 67	67	Sep-92	268	61.4	4.1	6.00
## 68	68	Dec-92	267	61.4	0.8	6.00
## 69	69	Mar-93	268	61.5	2.4	6.00
## 70	70	Jun-93	276	61.8	2.0	6.00
## 71	71	Sep-93	290	61.7	5.4	6.00
## 72	72	Dec-93	295	62.0	4.0	6.00
## 73	73	Mar-94	296	62.1	5.6	6.06
## 74	74	Jun-94	316	62.3	2.4	7.25
## 75	75	Sep-94	332	62.7	4.6	7.75
## 76	76	Dec-94	340	63.1	1.4	8.50
## 77	77	Mar-95	343	63.1	1.4	9.00
## 78	78	Jun-95	349	62.7	3.5	9.00
## 79	79	Sep-95	354	62.9	2.9	8.75
## 80	80	Dec-95	374	62.7	2.7	8.65
## 81	81	Mar-96	362	63.0	7.2	8.25
## 82	82	Jun-96	355	63.2	3.7	8.25
## 83	83	Sep-96	332	63.4	4.3	8.25
## 84	84	Dec-96	326	63.4	3.1	8.25
## 85	85	Mar-97	284	63.6	6.2	8.30
## 86	86	Jun-97	288	63.7	5.2	8.50

## 87	87	Sep-97	285	63.9	3.1	8.50
## 88	88	Dec-97	287	64.0	4.0	8.50
## 89	89	Mar-98	281	64.0	3.9	8.50
## 90	90	Jun-98	287	64.0	5.3	8.50
## 91	91	Sep-98	291	64.2	6.7	8.49
## 92	92	Dec-98	300	64.3	3.2	7.75
## 93	93	Mar-99	286	64.2	3.3	7.75
## 94	94	Jun-99	296	64.2	5.1	7.75
## 95	95	Sep-99	303	64.2	7.1	8.25
## 96	96	Dec-99	315	64.4	1.2	8.50
## 97	97	Mar-00	310	64.6	7.8	8.83
## 98	98	Jun-00	308	64.5	0.5	9.50
## 99	99	Sep-00	301	64.2	2.3	9.50
## 100	100	Dec-00	301	64.4	-1.1	9.50
## 101	101	Mar-01	286	64.3	2.1	8.32
## 102	102	Jun-01	302	63.7	-1.3	6.98
## 103	103	Sep-01	310	63.5	1.1	6.28
## 104	104	Dec-01	310	62.9	3.7	4.84
## 105	105	Mar-02	307	62.8	2.2	4.75
## 106	106	Jun-02	328	62.7	2.0	4.75
## 107	107	Sep-02	332	63.0	0.3	4.75
## 108	108	Dec-02	344	62.4	2.1	4.25
## 109	109	Mar-03	330	62.4	3.8	4.25
## 110	110	Jun-03	342	62.3	6.9	4.22
## 111	111	Sep-03	350	62.0	4.8	4.00
## 112	112	Dec-03	377	62.2	2.3	4.00
## 113	113	Mar-04	375	62.2	3.0	4.00
## 114	114	Jun-04	385	62.4	3.7	4.01
## 115	115	Sep-04	413	62.3	3.5	4.58
## 116	116	Dec-04	431	62.4	4.3	5.15
## 117	117	Mar-05	441	62.4	2.1	5.58
## 118	118	Jun-05	458	62.7	3.4	6.01
## 119	119	Sep-05	491	62.8	2.3	6.59
## 120	120	Dec-05	515	62.8	4.9	7.15
## 121	121	Mar-06	550	63.1	1.2	7.53
## 122	122	Jun-06	570	63.1	0.4	8.02
## 123	123	Sep-06	561	63.1	3.2	8.25
## 124	124	Dec-06	537	63.4	0.2	8.25
## 125	125	Mar-07	542	63.3	3.1	8.25
## 126	126	Jun-07	543	63.0	2.7	8.25
## 127	127	Sep-07	527	62.9	1.4	8.03
## 128	128	Dec-07	496	62.7	-2.7	7.33
## 129	129	Mar-08	465	62.7	2.0	5.66
## 130	130	Jun-08	435	62.4	-1.9	5.00
## 131	131	Sep-08	398	61.9	-8.2	5.00
## 132	132	Dec-08	352	61.0	-5.4	3.61
## 133	133	Mar-09	311	59.9	-0.5	3.25
## 134	134	Jun-09	282	59.4	1.3	3.25
## 135	135	Sep-09	254	58.7	3.9	3.25
## 136	136	Dec-09	232	58.3	1.7	3.25
## 137	137	Mar-10	227	58.5	3.9	3.25
## 138	138	Jun-10	213	58.5	2.7	3.25
## 139	139	Sep-10	204	58.5	2.5	3.25
## 140	140	Dec-10	188	58.3	-1.5	3.25

##	141	141	Mar-11	179	58.4	2.9	3.25
##	142	142	Jun-11	167	58.2	0.8	3.25
##	143	143	Sep-11	163	58.4	4.6	3.25
##	144	144	Dec-11	150	58.6	2.7	3.25
##	145	145	Mar-12	144	58.5	1.9	3.25
##	146	146	Jun-12	145	58.6	0.5	3.25
##	147	147	Sep-12	146	58.7	0.1	3.25
##	148	148	Dec-12	148	58.7	2.8	3.25
##	149	149	Mar-13	152	58.5	0.8	3.25
##	150	150	Jun-13	161	58.6	3.1	3.25
##	151	151	Sep-13	185	58.7	4.0	3.25
##	152	152	Dec-13	186	58.7	-0.9	3.25
##	153	153	Mar-14	187	58.9	4.6	3.25
##	154	154	Jun-14	197	59.0	5.2	3.25
##	155	155	Sep-14	213	59.1	2.0	3.25
##	156	156	Dec-14	212	59.3	3.2	3.25
##	157	157	Mar-15	202	59.3	2.7	3.25
##	158	158	Jun-15	216	59.4	1.6	3.25
##	159	159	Sep-15	227	59.2	0.5	3.25
##	160	160	Dec-15	235	59.6	0.6	3.37
##	161	161	Mar-16	240	59.8	2.2	3.50
##	162	162	Jun-16	243	59.7	2.8	3.50
##	163	163	Sep-16	248	59.7	1.8	3.50
##	164	164	Dec-16	257	59.8	1.2	3.64
##			SP500				
##	1		102.77				
##	2		104.28				
##	3		105.24				
##	4		107.46				
##	5		98.42				
##	6		100.48				
##	7		96.53				
##	8		95.10				
##	9		89.21				
##	10		95.53				
##	11		102.54				
##	12		96.11				
##	13		101.59				
##	14		102.91				
##	15		109.32				
##	16		107.94				
##	17		102.09				
##	18		114.24				
##	19		125.46				
##	20		135.76				
##	21		136.00				
##	22		131.21				
##	23		116.18				
##	24		122.55				
##	25		111.96				
##	26		109.61				
##	27		120.42				
##	28		140.64				
##	29		152.96				

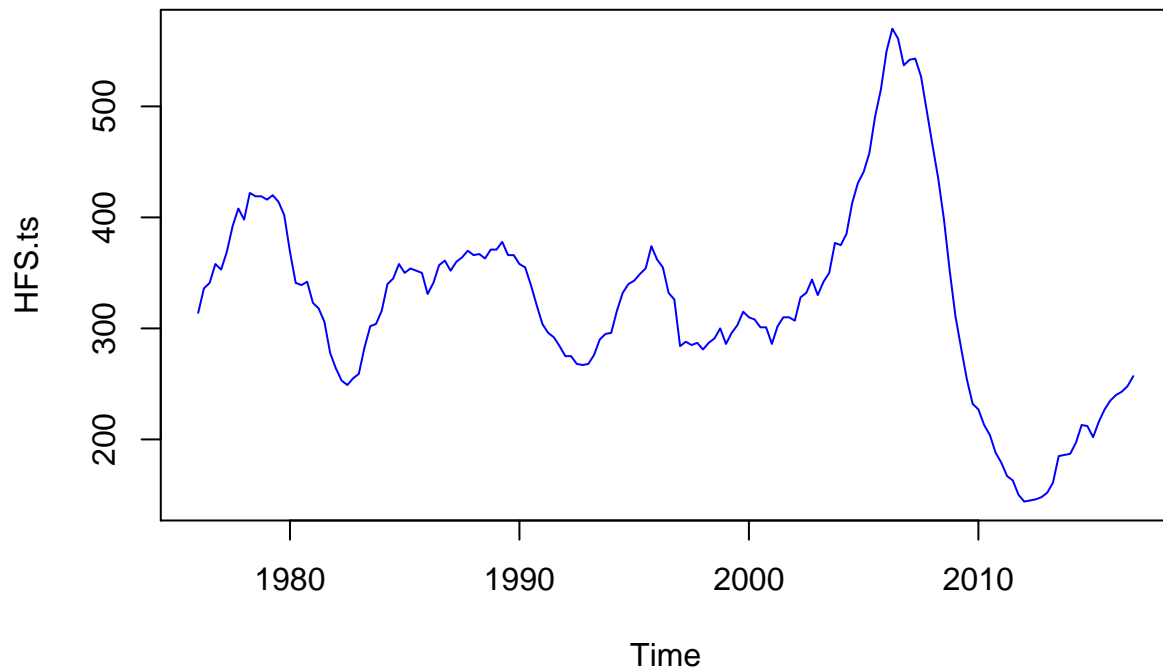
##	30	167.64
##	31	166.07
##	32	164.93
##	33	159.18
##	34	153.18
##	35	166.10
##	36	167.24
##	37	180.66
##	38	191.85
##	39	182.08
##	40	211.28
##	41	238.90
##	42	250.84
##	43	231.32
##	44	242.17
##	45	291.70
##	46	304.00
##	47	321.83
##	48	247.08
##	49	258.89
##	50	273.50
##	51	271.91
##	52	277.72
##	53	294.87
##	54	317.98
##	55	349.15
##	56	353.40
##	57	339.94
##	58	358.02
##	59	306.05
##	60	330.22
##	61	375.22
##	62	371.16
##	63	387.86
##	64	417.09
##	65	403.69
##	66	408.14
##	67	417.80
##	68	435.71
##	69	451.67
##	70	450.53
##	71	458.93
##	72	466.45
##	73	445.77
##	74	444.27
##	75	462.71
##	76	459.27
##	77	500.71
##	78	544.75
##	79	584.41
##	80	615.93
##	81	645.50
##	82	670.63
##	83	687.33

##	84	740.74
##	85	757.12
##	86	885.14
##	87	947.28
##	88	970.43
##	89	1101.75
##	90	1133.84
##	91	1017.01
##	92	1229.23
##	93	1286.37
##	94	1372.71
##	95	1282.71
##	96	1469.25
##	97	1498.58
##	98	1454.60
##	99	1436.51
##	100	1320.28
##	101	1160.33
##	102	1224.38
##	103	1040.94
##	104	1148.08
##	105	1147.39
##	106	989.82
##	107	815.28
##	108	879.82
##	109	848.18
##	110	974.50
##	111	995.97
##	112	1111.92
##	113	1126.21
##	114	1140.84
##	115	1114.58
##	116	1211.92
##	117	1180.59
##	118	1191.33
##	119	1228.81
##	120	1248.29
##	121	1294.87
##	122	1270.20
##	123	1335.85
##	124	1418.30
##	125	1420.86
##	126	1503.35
##	127	1526.75
##	128	1468.36
##	129	1322.70
##	130	1280.00
##	131	1166.36
##	132	903.25
##	133	797.87
##	134	919.32
##	135	1057.08
##	136	1115.10
##	137	1169.43

```
## 138 1030.71
## 139 1141.20
## 140 1257.64
## 141 1325.83
## 142 1320.64
## 143 1131.42
## 144 1257.60
## 145 1408.47
## 146 1362.16
## 147 1440.67
## 148 1426.19
## 149 1569.19
## 150 1606.28
## 151 1681.55
## 152 1848.36
## 153 1872.34
## 154 1960.23
## 155 1972.29
## 156 2058.90
## 157 2067.89
## 158 2063.11
## 159 1920.03
## 160 2043.94
## 161 2059.74
## 162 2098.86
## 163 2168.27
## 164 2238.83
```

Question 1. Provide the time series plot of the ‘Houses for Sale’ data. Which assumptions of stationarity, if any, seem to be violated. Provide the time series plot of the 1st order difference of the Houses for Sale data.

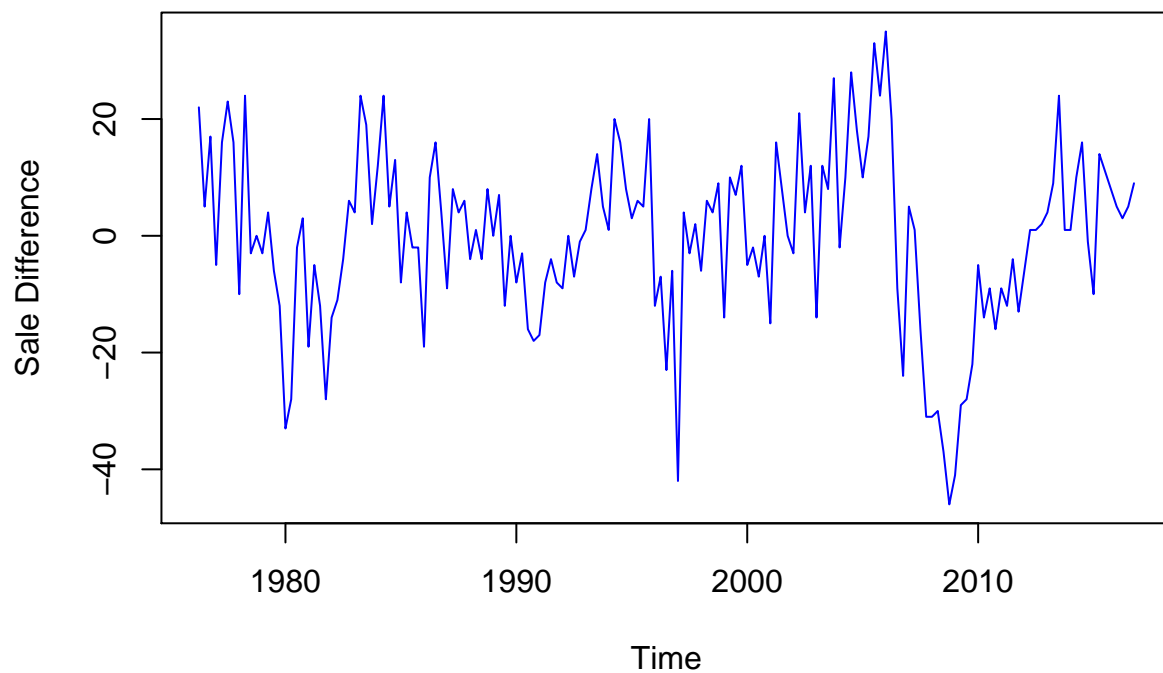
```
HFS = data.train$Houses_for_Sale
HFS.ts = ts(data=HFS, start=1976, frequency = 4)
ts.plot(HFS.ts, col='blue')
```



Constant mean/constant variance

```
HFS.dif = diff(HFS.ts)
ts.plot(HFS.dif,main='House of Sale - Differenced',ylab='Sale Difference',col='blue')
```

House of Sale – Differenced



Question 2.

Identify the best ARIMA model to fit the Houses for Sale data according to the AIC criterion. Use regular 1st order differencing and consider up to order 5 for AR and MA terms. Report the AIC and estimated

coefficients of your selected model. Which coefficients are significant at the 5% significance level?

Provide plots of the ACF and PACF of the residuals from your ARIMA model and perform the Ljung-Box test. Do the residuals appear uncorrelated?

```
n = length(HFS.ts)
test_modelA <- function(p,d,q){
  mod = arima(HFS.ts, order=c(p,d,q), method="ML")
  current.aic = AIC(mod)
  df = data.frame(p,d,q,current.aic)
  names(df) <- c("p","d","q","AIC")
  print(paste(p,d,q,current.aic,sep=" "))
  return(df)
}

orders = data.frame(Inf,Inf,Inf,Inf)
names(orders) <- c("p","d","q","AIC")

for (p in 0:5){
  for (d in 0:1){
    for (q in 0:5) {
      possibleError <- tryCatch(
        orders<-rbind(orders,test_modelA(p,d,q)),
        error=function(e) e
      )
      if(inherits(possibleError, "error")) next
    }
  }
}
```

```
## [1] "0 0 0 1936.28822900367"
## [1] "0 0 1 1750.12086560537"
## [1] "0 0 2 1564.52151886075"
## [1] "0 0 3 1502.77987432438"
## [1] "0 0 4 1414.04420380128"
## [1] "0 0 5 1398.82179493849"
## [1] "0 1 0 1345.99676335138"
## [1] "0 1 1 1313.44753993353"
## [1] "0 1 2 1289.66768637719"
## [1] "0 1 3 1291.61382559805"
## [1] "0 1 4 1264.51999484175"
## [1] "0 1 5 1265.6773305181"
## [1] "1 0 0 1359.72019871953"
```

```
## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1
```

```
## [1] "1 0 2 1301.89757172154"
## [1] "1 0 3 1309.08454336027"
```

```
## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1
```

```

## [1] "1 0 4 1306.04260365583"
## [1] "1 0 5 1291.06419082058"
## [1] "1 1 0 1285.35526692511"
## [1] "1 1 1 1272.37486246077"

## Warning in log(s2): NaNs produced

## Warning in log(s2): NaNs produced

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "1 1 2 1274.01272070035"
## [1] "1 1 3 1275.69626249276"
## [1] "1 1 4 1262.83351521415"
## [1] "1 1 5 1261.72500934647"
## [1] "2 0 0 1295.91390695195"
## [1] "2 0 1 1361.05370748814"
## [1] "2 0 2 1299.25825693534"

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "2 0 3 1287.4978176598"
## [1] "2 0 4 1287.07562048983"
## [1] "2 0 5 1270.60099115942"
## [1] "2 1 0 1273.99529402935"
## [1] "2 1 1 1261.35419760246"
## [1] "2 1 2 1260.45811666923"
## [1] "2 1 3 1257.18925731033"
## [1] "2 1 4 1251.43150579343"

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "2 1 5 1250.73144451251"
## [1] "3 0 0 1280.75552380391"
## [1] "3 0 1 1270.4341180022"
## [1] "3 0 2 1262.11058887685"

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "3 0 3 1293.92063721938"

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "3 0 4 1260.25426290657"

```

```

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "3 0 5 1253.6789671528"
## [1] "3 1 0 1275.71385853575"
## [1] "3 1 1 1262.70117520159"
## [1] "3 1 2 1260.82130041566"
## [1] "3 1 3 1256.90242292559"
## [1] "3 1 4 1250.14731447623"
## [1] "3 1 5 1245.12277920984"
## [1] "4 0 0 1281.25151121983"
## [1] "4 0 1 1270.71855442803"

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "4 0 2 1274.26947922504"

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "4 0 3 1261.8485532801"
## [1] "4 0 4 1260.28567658561"

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "4 0 5 1254.59416640471"
## [1] "4 1 0 1262.01701610238"
## [1] "4 1 1 1249.71107590406"
## [1] "4 1 2 1251.64857530809"
## [1] "4 1 3 1251.71045923905"
## [1] "4 1 4 1252.98899063494"
## [1] "4 1 5 1254.95126964376"
## [1] "5 0 0 1257.51023698568"
## [1] "5 0 1 1249.99803340576"

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "5 0 2 1252.3924480544"

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "5 0 3 1253.5880376859"

## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1

## [1] "5 0 4 1255.22816394422"

```

```
## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1
```

```
## [1] "5 0 5 1257.05512819809"
## [1] "5 1 0 1252.86213433597"
## [1] "5 1 1 1251.60669459247"
## [1] "5 1 2 1242.08470194206"
## [1] "5 1 3 1243.51282706088"
## [1] "5 1 4 1245.22727489838"
```

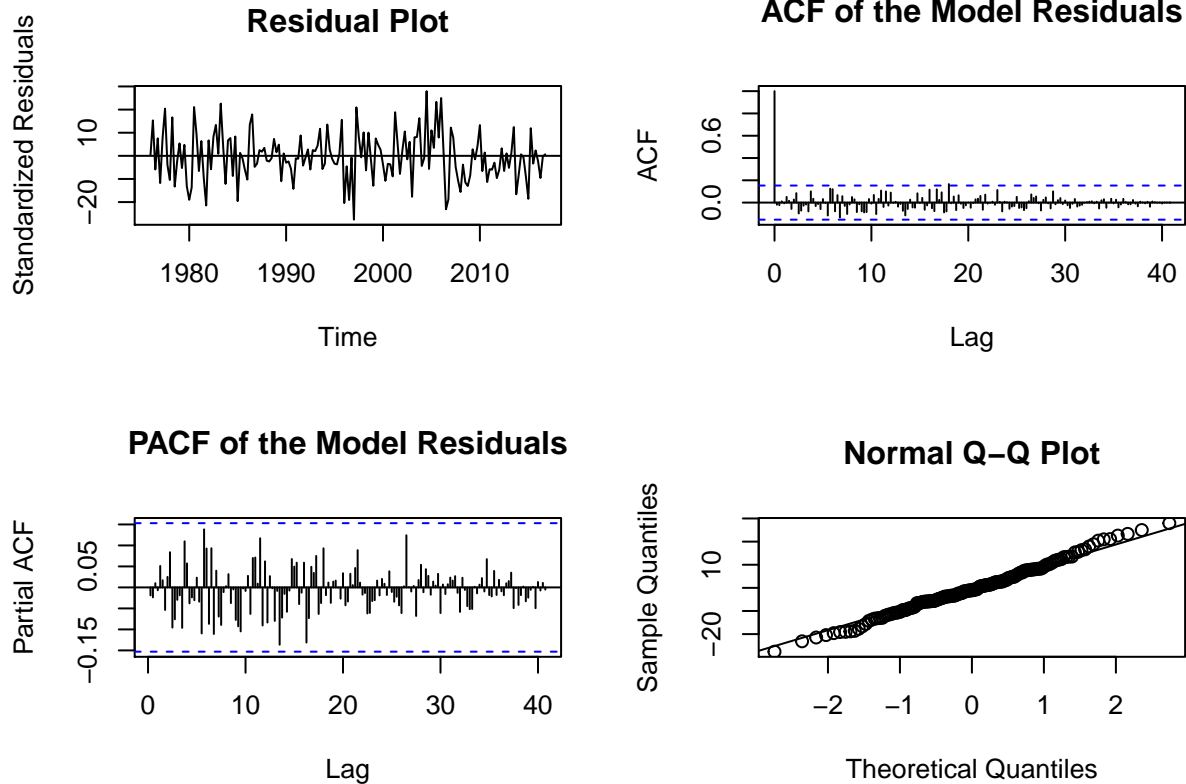
```
## Warning in arima(HFS.ts, order = c(p, d, q), method = "ML"): possible
## convergence problem: optim gave code = 1
```

```
## [1] "5 1 5 1247.15459748106"
```

```
orders <- orders[order(-orders$AIC),]
tail(orders)
```

```
##      p d q      AIC
## 56 4 1 1 1249.711
## 72 5 1 5 1247.155
## 71 5 1 4 1245.227
## 48 3 1 5 1245.123
## 70 5 1 3 1243.513
## 69 5 1 2 1242.085
```

```
final_HFS = arima(HFS.ts, order = c(5,1,2), method = "ML")
resids.HFS = resid(final_HFS)
## Residual Analysis
par(mfrow=c(2,2))
plot(resids.HFS, ylab='Standardized Residuals', main="Residual Plot")
abline(h=0)
acf(resids.HFS, main='ACF of the Model Residuals', lag.max = 52*4)
pacf(resids.HFS, main='PACF of the Model Residuals', lag.max = 52*4)
qqnorm(resids.HFS)
qqline(resids.HFS)
```



```
## Test for Independence for final model
Box.test(resids.HFS, lag = (5+1+2), type = "Box-Pierce", fitdf = (5+2))
```

```
##
## Box-Pierce test
##
## data: resids.HFS
## X-squared = 1.2875, df = 1, p-value = 0.2565
```

```
Box.test(resids.HFS, lag = (5+1+2), type = "Ljung-Box", fitdf = (5+2))
```

```
##
## Box-Ljung test
##
## data: resids.HFS
## X-squared = 1.3504, df = 1, p-value = 0.2452
```

```
library(lmtest)
```

```
## Loading required package: zoo
```

```
##
## Attaching package: 'zoo'
```

```
## The following objects are masked from 'package:base':
##
## as.Date, as.Date.numeric
```

```
coeftest(final_HFS)
```

```
##
## z test of coefficients:
##
##      Estimate Std. Error z value Pr(>|z|)
## ar1  0.781603   0.117265  6.6653 2.642e-11 ***
## ar2  0.547297   0.163586  3.3456 0.000821 ***
## ar3 -0.277578   0.098355 -2.8222 0.004769 **
## ar4  0.292068   0.112128  2.6048 0.009194 **
## ar5 -0.403426   0.075578 -5.3379 9.403e-08 ***
## ma1 -0.348029   0.124228 -2.8015 0.005086 **
## ma2 -0.651964   0.122592 -5.3182 1.048e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Question 3.

Plot the forecasts for the 2017 data using the ARIMA model fit in Question 2 alongside the actual observations. Using the precision measure, assess the prediction accuracy of the model.

EU Forecasting

