STAT349 based on log10

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AAL, 2015-03-23 to 2020-03-20

##Model 1: ARIMA(3,1,0)

```
## ## Call:
## arima(x = newdata(a), order = auto.selection(newdata(a), testdata(a), totalobs(a))[1:3])
##
## Coefficients:
## arl ar2 ar3
## -0.7394 -0.5403 -0.2815
## s.e. 0.0273 0.0316 0.0284
##
## sigma^2 estimated as 0.000138: log likelihood = 3787.37, aic = -7568.75
```

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

```
## [1] -0.046303430 0.009375610 0.126241528 0.053910551 -0.003781979
```

```
## Time Series:

## Start = 1254

## End = 1258

## Frequency = 1

## [1] -0.002957262 0.020636747 0.021013643 0.001243320 0.009017519
```

```
## Time Series:

## Start = 1254

## End = 1258

## Frequency = 1

## [1] -0.04334617 -0.01126114 0.10522788 0.05266723 -0.01279950
```

```
## [1] 0.003203255
```

##Model 2: ARIMA(2,1,3)

```
## ## Call:
## arima(x = newdata(c3), order = auto.selection(newdata(c3), testdata(c3), totalobs(c3))[1:3])
##
## Coefficients:
## ar1 ar2 ma1 ma2 ma3
## -1.1147 -0.9913 0.1043 -0.0975 -0.9967
## s.e. 0.0111 0.0093 0.0201 0.0201 0.0230
##
## sigma^2 estimated as 7.016e-05: log likelihood = 1381.19, aic = -2752.39
```

I split the data in to 5 subsets, which are

```
## [1] "2015-03-24" "2016-01-04"
 ## [1] "2016-01-04" "2016-07-01"
 ## [1] "2016-01-04" "2018-03-01"
 ## [1] "2018-03-01" "2019-01-02"
 ## [1] "2019-01-02" "2020-03-20"
and I use the third 2016-01-04 to 2018-03-01 to build the model.
the last 5 days actual value, forecast value, forecast errors and mean squared forecast error
 ## Time Series:
 ## Start = 414
 ## End = 418
 ## Frequency = 1
 \#\# [1] -0.0012135155 -0.0006818861 0.0002062442 -0.0013107154 -0.0005001823
 ## Time Series:
 ## Start = 414
 ## End = 418
 ## Frequency = 1
 ## [1] -0.002134496 -0.012220888 0.005676664 0.001390815 0.006708508
 ## [1] 4.661383e-05
##Model 3 ARIMA(3,1,0) I split the data in to 5 subsets, which are
 ## [1] "2015-03-24" "2015-06-09"
 ## [1] "2015-06-09" "2016-04-21"
 ## [1] "2016-04-21" "2016-07-11"
 ## [1] "2016-07-11" "2018-03-19"
 ## [1] "2018-03-19" "2020-03-20"
the transformations are obtain by Box-Cox and the last 5 days actual value, forecast value, forecast errors and
mean squared forecast error are
```

"log(x)" "x"

[1] "x²"

"x^3"

```
## call:
## arima(x = newdata(c.new), order = auto.selection(newdata(c.new), testdata(c.new),
## totalobs(c.new))[1:3])
##
## Coefficients:
## ar1 ar2 ar3
## -0.7391 -0.5405 -0.2814
## s.e. 0.0273 0.0316 0.0284
##
## sigma^2 estimated as 0.000138: log likelihood = 3787.19, aic = -7568.38
```

```
## Time Series:
## Start = 1254
## End = 1258
## Frequency = 1
## [1] -0.002995491  0.020659255  0.021005294  0.001233819  0.009004292
```

```
## Time Series:
## Start = 1254
## End = 1258
## Frequency = 1
## [1] -0.04330794 -0.01128365 0.10523623 0.05267673 -0.01278627
```

```
## [1] 0.003203178
```

BBY, 1999-06-29 to 2020-03-20

##Model 1: ARIMA(3,1,0)

```
##
## Call:
## arima(x = newdata(a), order = auto.selection(newdata(a), testdata(a), totalobs(a))[1:3])
##
## Coefficients:
## ar1 ar2 ar3
## -0.7889 -0.5597 -0.2481
## s.e. 0.0136 0.0158 0.0138
##
## sigma^2 estimated as 0.0001304: log likelihood = 15905.28, aic = -31804.56
```

```
## [1] 0.08155932 0.05931400 0.13566260 -0.03151705 -0.07918125
```

```
## Time Series:
## Start = 5211
## End = 5215
## Frequency = 1
## [1] 0.025122843 0.050570114 0.034932953 0.009474805 0.031996874
```

```
## Time Series:
## Start = 5211
## End = 5215
## Frequency = 1
## [1] 0.056436476 0.008743887 0.100729649 -0.040991856 -0.111178120
```

```
## [1] 0.00548978
```

##Model 2: ARIMA(3,1,3)

```
##
## Call:
\#\# arima(x = newdata(c2), order = auto.selection(newdata(c2), testdata(c2), totalobs(c2))[1:3])
## Coefficients:
##
             ar1
                    ar2
                            ar3
                                     ma1
                                                ma2
                                                         ma3
         -0.2695 -0.9305 0.0527 -0.6824 0.6547 -0.9685
##
## s.e. 0.0417 0.0550 0.0324 0.0280 0.0357 0.0439
##
\#\# \text{ sigma 2 estimated as } 3.36e-05: \ \log \text{ likelihood} = 3706.22, \ \text{aic} = -7400.45
```

I split the data in to 4 subsets, which are

```
## [1] "1999-06-29" "2003-06-23"
```

```
## [1] "2003-06-23" "2007-06-13"
```

```
## [1] "2007-06-13" "2012-03-16"
```

```
## [1] "2012-03-16" "2020-03-20"
```

and I use the second 2003-06-23 to 2007-06-13 to build the model.

```
## [1] 0.010955468 -0.000620708 0.000709607 -0.006869880 -0.001570279
```

```
## Time Series:

## Start = 996

## End = 1000

## Frequency = 1

## [1] 3.390063e-04 1.047497e-05 -2.281281e-04 -3.272679e-04 -9.584453e-05
```

```
## Time Series:
## Start = 996
## End = 1000
## Frequency = 1
## [1] 0.0106164617 -0.0006311830 0.0009377351 -0.0065426121 -0.0014744345
```

```
## [1] 3.179335e-05
```

##Model 3 ARIMA(3,1,0) I split the data in to 2 subsets, which are

```
## [1] "1999-06-29" "2008-08-19"
```

```
## [1] "2008-08-19" "2020-03-20"
```

the transformations are obtain by Box-Cox and the last 5 days actual value, forecast value, forecast errors and mean squared forecast error are

```
## [1] "x" "x"
```

```
## call:
## arima(x = newdata(c.new), order = auto.selection(newdata(c.new), testdata(c.new),
## totalobs(c.new))[1:3])
##
## Coefficients:
## ar1 ar2 ar3
## -0.7874 -0.5589 -0.2475
## s.e. 0.0136 0.0158 0.0139
##
## sigma^2 estimated as 0.0001309: log likelihood = 15895.64, aic = -31785.27
```

```
## [1] 0.08155932 0.05931400 0.13566260 -0.03151705 -0.07918125
```

```
## Time Series:
## Start = 5211
## End = 5215
## Frequency = 1
## [1] 0.024939211 0.050529212 0.034874557 0.009448575 0.031882979
```

```
## Time Series:
## Start = 5211
## End = 5215
## Frequency = 1
## [1] 0.056620108 0.008784789 0.100788045 -0.040965627 -0.111064225
```

```
## [1] 0.005490937
```

BIIB, 2003-11-13 to 2020-03-20

##Model 1: ARIMA(1,1,0)

```
##
## Call:
## arima(x = newdata(a), order = auto.selection(newdata(a), testdata(a), totalobs(a))[1:3])
##
## Coefficients:
## ar1
## -0.5416
## s.e. 0.0135
##
## sigma^2 estimated as 0.0001658: log likelihood = 12049.77, aic = -24097.55
```

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

[1] 0.02118389 -0.03272020 0.09284443 -0.01311470 0.01902892

```
## Time Series:
## Start = 4110
## End = 4114
## Frequency = 1
## [1] 0.031653263 -0.023658991 0.006296675 -0.009926528 -0.001140466
```

```
## Time Series:
## Start = 4110
## End = 4114
## Frequency = 1
## [1] -0.010469376 -0.009061209 0.086547751 -0.003188169 0.020169387
```

```
## [1] 0.001619839
```

##Model 2: ARIMA(0,1,0)

```
##
## Call:
## arima(x = newdata(c1), order = auto.selection(newdata(c1), testdata(c1), totalobs(c1))[1:3])
##
## Coefficients:
## mal
## -1.0000
## s.e. 0.0028
##
## sigma^2 estimated as 0.0001002: log likelihood = 3799.84, aic = -7597.68
```

I split the data in to 5 subsets, which are

```
## [1] "2003-11-13" "2008-08-21"
```

```
## [1] "2008-08-21" "2009-06-09"
```

```
## [1] "2009-06-09" "2011-06-02"
```

```
## [1] "2011-06-02" "2011-10-24"
```

```
## [1] "2011-10-24" "2020-03-20"
```

and I use the first 2003-11-13 to 2008-08-21 to build the model.

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

```
## [1] -0.009731986 0.007056097 0.006992058 0.004176639 0.004033217
```

```
## Time Series:
## Start = 1196
## End = 1200
## Frequency = 1
## [1] -0.0002580046 -0.0002580046 -0.0002580046 -0.0002580046
```

```
## Time Series:
## Start = 1196
## End = 1200
## Frequency = 1
## [1] -0.009473981  0.007314102  0.007250063  0.004434644  0.004291222
```

```
## [1] 4.677929e-05
```

##Model 3 ARIMA(1,1,0) I split the data in to 5 subsets, which are

```
## [1] "2003-11-13" "2005-01-25"
```

```
## [1] "2005-01-25" "2007-11-05"
```

```
## [1] "2007-11-05" "2009-03-27"
```

```
## [1] "2009-03-27" "2016-03-10"
```

```
## [1] "2016-03-10" "2020-03-20"
```

```
## [1] "x" "log(x)" "x" "x" "x"
```

```
## [1] 0.02118389 -0.03272020 0.09284443 -0.01311470 0.01902892
```

```
## Time Series:
## Start = 3411
## End = 3415
## Frequency = 1
## [1] 0.048718282 -0.026622120 0.020997821 -0.009101020 0.009923368
```

```
## Time Series:
## Start = 3411
## End = 3415
## Frequency = 1
## [1] -0.027534395 -0.006098080 0.071846604 -0.004013677 0.009105553
```

```
## [1] 0.001211257
```

BSX, 1995-02-24 to 2020-03-20

##Model 1: ARIMA(1,1,0)

```
##
## Call:
## arima(x = newdata(a), order = auto.selection(newdata(a), testdata(a), totalobs(a))[1:3])
##
## Coefficients:
## ar1
## -0.4924
## s.e. 0.0110
##
## sigma^2 estimated as 0.0001797: log likelihood = 18241.48, aic = -36480.96
```

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

```
## [1] 0.06948414 -0.01912009 0.02403458 -0.01611871 0.01611871
```

```
## Time Series:

## Start = 6307

## End = 6311

## Frequency = 1

## [1] 0.0096131189 -0.0041049409 0.0026504206 -0.0006762094 0.0009619660
```

```
## Time Series:
## Start = 6307
## End = 6311
## Frequency = 1
## [1] 0.05987103 -0.01501515 0.02138416 -0.01544250 0.01515674
```

```
## [1] 0.0009470948
```

##Model 2: ARIMA(2,1,3)

```
## ## Call:
## arima(x = newdata(c4), order = auto.selection(newdata(c4), testdata(c4), totalobs(c4))[1:3])
##
## Coefficients:
## arl ar2 mal ma2 ma3
## -0.7041 -0.9791 -0.3130 0.2761 -0.9631
## s.e. 0.0248 0.0209 0.0345 0.0311 0.0265
##
## sigma^2 estimated as 0.0001726: log likelihood = 2746.6, aic = -5483.2
```

I split the data in to 5 subsets, which are

```
## [1] "1995-02-24" "1997-02-14"
```

```
## [1] "1997-02-14" "2002-09-11"
```

```
## [1] "2002-09-11" "2008-06-16"
```

```
## [1] "2008-06-16" "2012-03-22"
```

```
## [1] "2012-03-22" "2020-03-20"
```

and I use the fourth 2008-06-16 to 2012-03-22 to build the model.

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

```
## [1] -0.008600172 0.005714484 0.000719626 0.000000000 0.002166062
```

```
## Time Series:

## Start = 946

## End = 950

## Frequency = 1

## [1] -0.0002469519 0.0007407962 0.0007330711 -0.0002286262 0.0004560532
```

```
## Time Series:

## Start = 946

## End = 950

## Frequency = 1

## [1] -0.0083532201 0.0049736878 -0.0000134451 0.0002286262 0.0017100088
```

```
## [1] 1.949809e-05
```

##Model 3 ARIMA(1,1,0) I split the data in to 5 subsets, which are

```
## [1] "1995-02-24" "1999-02-10"
```

```
## [1] "1999-02-10" "2000-11-20"
```

```
## [1] "2000-11-20" "2004-04-14"
```

```
## [1] "2004-04-14" "2012-08-14"
```

```
## [1] "2012-08-14" "2020-03-20"
```

the transformations are obtain by Box-Cox and the last 5 days actual value, forecast value, forecast errors and mean squared forecast error are

```
## [1] "x" "log(x)" "x" "x" "x"
```

```
## [1] 0.06948414 -0.01912009 0.02403458 -0.01611871 0.01611871
```

```
## Time Series:
## Start = 5858
## End = 5862
## Frequency = 1
## [1] 0.0158268527 -0.0046935595 0.0076655926 0.0002218512 0.0047051107
```

```
## Time Series:
## Start = 5858
## End = 5862
## Frequency = 1
## [1] 0.05365729 -0.01442653 0.01636898 -0.01634056 0.01141359
```

```
## [1] 0.0007504915
```

BXP, 2006-04-03 to 2020-03-20

##Model 1: ARIMA(2,1,0)

```
##
## Call:
## arima(x = newdata(a), order = auto.selection(newdata(a), testdata(a), totalobs(a))[1:3])
##
## Coefficients:
## arl ar2
## -0.7938 -0.3772
## s.e. 0.0157 0.0157
##
## sigma^2 estimated as 0.0001275: log likelihood = 10753.44, aic = -21502.87
```

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

[1] 0.08195031 -0.03819163 0.07913202 0.01319627 0.01850771

```
## Time Series:

## Start = 3511

## End = 3515

## Frequency = 1

## [1] 0.022250053 0.009631768 0.003854704 0.013200426 0.007961105
```

```
## Time Series:

## Start = 3511

## End = 3515

## Frequency = 1

## [1] 5.970025e-02 -4.782340e-02 7.527731e-02 -4.159822e-06 1.054660e-02
```

```
## [1] 0.002325821
```

##Model 2: ARIMA(2,0,3)

```
##
## Call:
\#\# arima(x = newdata(c1), order = auto.selection(newdata(c1), testdata(c1), totalobs(c1))[1:3])
##
## Coefficients:
##
           ar1
                    ar2
                             ma1
                                      ma2
                                               ma3 intercept
         1.4941 \quad -0.7542 \quad -1.4555 \quad 0.7121 \quad -0.0260
                                                       -2e-04
## s.e. 0.2508 0.2632 0.2555 0.2601 0.0639
                                                        3e-04
##
## sigma^2 estimated as 4.293e-05: log likelihood = 1425.54, aic = -2839.07
```

I split the data in to 3 subsets, which are

```
## [1] "2006-04-03" "2007-11-02"
```

```
## [1] "2007-11-02" "2011-10-21"
```

```
## [1] "2011-10-21" "2020-03-20"
```

and I use the first 2006-04-03 to 2007-11-02 to build the model.

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

```
## [1] 0.000608073 0.002196473 -0.007439159 0.012731888 0.008882582

## Time Series:
## Start = 396
## End = 400
## Frequency = 1
## [1] 0.0001769618 0.0007735013 0.0011734125 0.0011141849 0.0007240786

## Time Series:
## Start = 396
## End = 400
## Frequency = 1
## [1] 0.0004311112 0.0014229717 -0.0086125715 0.0116177031 0.0081585034
```

```
## [1] 5.558386e-05
```

##Model 3 ARIMA(0,1,0) I split the data in to 4 subsets, which are

```
## [1] "2006-04-03" "2008-03-31"

## [1] "2008-03-31" "2009-03-05"

## [1] "2009-03-05" "2011-10-21"

## [1] "2011-10-21" "2020-03-20"
```

```
## [1] "x" "x" "x" "x^2"
```

```
##
## Call:
## arima(x = newdata(c.new), order = auto.selection(newdata(c.new), testdata(c.new),
## totalobs(c.new))[1:3])
##
##
##
## sigma^2 estimated as 0.0004735: log likelihood = 3357.81, aic = -6715.63
```

```
## [1] 0.008453447 -0.009862137 0.012218379 -0.003744759 -1.988946634
```

```
## Time Series:
## Start = 1396
## End = 1400
## Frequency = 1
## [1] -0.01163015 -0.01163015 -0.01163015 -0.01163015
```

```
## Time Series:

## Start = 1396

## End = 1400

## Frequency = 1

## [1] 0.020083597 0.001768013 0.023848529 0.007885391 -1.977316484
```

```
## [1] 0.7821636
```

COG, 2008-06-23 to 2020-03-20

##Model 1: ARIMA(3,1,0)

```
##
## Call:
## arima(x = newdata(a), order = auto.selection(newdata(a), testdata(a), totalobs(a))[1:3])
##
## Coefficients:
## arl ar2 ar3
## -0.8067 -0.5914 -0.2691
## s. e. 0.0181 0.0213 0.0188
##
## sigma^2 estimated as 0.000125: log likelihood = 9069.83, aic = -18133.66
```

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

```
## [1] 0.08155932 0.05931400 0.13566260 -0.03151705 -0.07918125
```

```
## Time Series:
## Start = 2952
## End = 2956
## Frequency = 1
## [1] 0.023673884 0.053437146 0.036597433 0.007422808 0.032906763
```

```
## Time Series:
## Start = 2952
## End = 2956
## Frequency = 1
## [1] 0.057885435 0.005876855 0.099065169 -0.038939859 -0.112088009
```

```
## [1] 0.005455841
```

##Model 2: ARIMA(1,1,0)

```
\#\# arima(x = newdata(c2), order = auto.selection(newdata(c2), testdata(c2), totalobs(c2))[1:3])
##
## Coefficients:
##
            ar1
                     ar2
                             ar3
                                      ma1
                                             ma2 intercept
         0.9481 \quad -0.0796 \quad 0.0211 \quad -1.0318 \quad 0.1135
##
                                                    -1e-04
## s.e. 0.7675
                 0.6104 0.0509 0.7673 0.6729
                                                        3e-04
## sigma^2 estimated as 9.298e-05: log likelihood = 2449.17, aic = -4886.35
```

I split the data in to 3 subsets, which are

```
## [1] "2008-06-23" "2009-05-29"
```

```
## [1] "2009-05-29" "2012-06-11"
```

```
## [1] "2012-06-11" "2020-03-20"
```

and I use the second 2009-05-29 to 2012-06-110 to build the model.

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

```
## [1] -0.000839203 -0.010219110 -0.004480894 -0.011727468 0.004095164
```

```
## Time Series:
## Start = 761
## End = 765
## Frequency = 1
## [1] -0.0008974133 -0.0004100617 -0.0003161052 -0.0003023782 -0.0002865842
```

```
## Time Series:
## Start = 761
## End = 765
## Frequency = 1
## [1] 5.821031e-05 -9.809048e-03 -4.164789e-03 -1.142509e-02 4.381748e-03
```

```
## [1] 5.265974e-05
```

##Model 3 ARIMA(1,1,0) I split the data in to 4 subsets, which are

```
## [1] "2008-06-23" "2010-09-27"
```

```
## [1] "2010-09-27" "2016-06-02"
```

```
## [1] "2016-06-02" "2017-10-20"
```

```
## [1] "2017-10-20" "2020-03-20"
```

```
## [1] "x" "log(x)" "x" "x"
```

```
##
## Call:
## arima(x = newdata(c.new), order = auto.selection(newdata(c.new), testdata(c.new),
##
    totalobs(c.new))[1:3])
##
## Coefficients:
##    arl
##    -0.6617
## s.e.    0.0193
##
## sigma^2 estimated as 0.002732: log likelihood = 2330.4, aic = -4658.8
```

```
## [1] 0.08155932 0.05931400 0.13566260 -0.03151705 -0.07918125
```

```
## Time Series:
## Start = 1523
## End = 1527
## Frequency = 1
## [1] 0.083749421 -0.017858408 0.049379539 0.004885511 0.034328982
```

```
## Time Series:
## Start = 1523
## End = 1527
## Frequency = 1
## [1] -0.002190102  0.077172409  0.086283063 -0.036402563 -0.113510228
```

```
## [1] 0.005522973
```

GS, 2002-07-22 to 2020-03-20

##Model 1: ARIMA(3,1,0)

```
## call:
## arima(x = newdata(a), order = auto.selection(newdata(a), testdata(a), totalobs(a))[1:3])
##
## Coefficients:
## arl ar2 ar3
## -0.7980 -0.5663 -0.2543
## s.e. 0.0148 0.0173 0.0152
##
## sigma^2 estimated as 0.0001072: log likelihood = 13995.21, aic = -27984.42
```

```
## [1] 0.08155932 0.05931400 0.13566260 -0.03151705 -0.07918125
```

```
## Time Series:
## Start = 4443
## End = 4447
## Frequency = 1
## [1] 0.025918073 0.050660504 0.035708130 0.009290783 0.032547511
```

```
## Time Series:
## Start = 4443
## End = 4447
## Frequency = 1
## [1] 0.055641246 0.008653497 0.099954472 -0.040807834 -0.111728757
```

```
## [1] 0.005462064
```

##Model 2: ARIMA(1,0,2)

```
##
## Call:
\#\# arima(x = newdata(c2), order = auto.selection(newdata(c2), testdata(c2), totalobs(c2))[1:3])
## Coefficients:
                   ar2
##
           ar1
                           ma1
                                   ma2 intercept
        1. 1268 -0. 4686 -1. 0725 0. 4088
                                          -2e-04
##
## s.e. 0.3630 0.3558 0.3740 0.3657
                                             2e-04
##
## sigma^2 estimated as 3.8e-05: log likelihood = 4385.68, aic = -8761.35
```

I split the data in to 4 subsets, which are

```
## [1] "2002-07-22" "2003-05-07"
```

```
## [1] "2003-05-07" "2008-02-12"
```

```
## [1] "2008-02-12" "2011-12-22"
```

```
## [1] "2011-12-22" "2020-03-20"
```

and I use the second 2003-05-07 to 2008-02-12 to build the model.

```
## [1] 0.001281855 -0.002772433 0.000531811 -0.005919091 -0.010270333
```

```
## Time Series:

## Start = 1196

## End = 1200

## Frequency = 1

## [1] 0.0011501573 -0.0002972733 -0.0009276863 -0.0009598562 -0.0007007227
```

```
## Time Series:
## Start = 1196
## End = 1200
## Frequency = 1
## [1] 0.0001316977 -0.0024751597 0.0014594973 -0.0049592348 -0.0095696103
```

```
## [1] 2.488907e-05
```

##Model 3 ARIMA(3,1,0) I split the data in to 4 subsets, which are

```
## [1] "2002-07-22" "2004-12-07"

## [1] "2004-12-07" "2008-11-25"

## [1] "2008-11-25" "2017-11-02"

## [1] "2017-11-02" "2020-03-20"
```

the transformations are obtain by Box-Cox and the last 5 days actual value, forecast value, forecast errors and mean squared forecast error are

```
## [1] "x" "x" "x" "x"
##
## Call:
## arima(x = newdata(c.new), order = auto.selection(newdata(c.new), testdata(c.new),
      totalobs(c.new))[1:3])
##
## Coefficients:
##
            ar1
                    ar2
                             ar3
##
        -0.7963 \quad -0.5650 \quad -0.2536
## s.e. 0.0148 0.0173 0.0152
##
## sigma^2 estimated as 0.0001077: log likelihood = 13984.42, aic = -27962.84
```

```
## [1] 0.08155932 0.05931400 0.13566260 -0.03151705 -0.07918125
```

```
## Time Series:
## Start = 4443
## End = 4447
## Frequency = 1
## [1] 0.02573819 0.05056980 0.03565115 0.00927605 0.03241051
```

```
## Time Series:
## Start = 4443
## End = 4447
## Frequency = 1
## [1] 0.055821131 0.008744198 0.100011452 -0.040793101 -0.111591758
```

```
## [1] 0.00546231
```

INTC, 1986-08-28 to 2020-03-20

##Model 1: ARIMA(1,1,0)

```
##
## Call:
## arima(x = newdata(a), order = auto.selection(newdata(a), testdata(a), totalobs(a))[1:3])
##
## Coefficients:
## ar1
## -0.4756
## s.e. 0.0096
##
## sigma^2 estimated as 0.0002223: log likelihood = 23554.11, aic = -47106.21
```

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

[1] 0.013578679 -0.075429400 0.040299072 -0.035278274 0.002122872

[1] 0.02069740 -0.05904133 0.05227842 -0.02120203 0.01520179

```
## Time Series:
## Start = 8454
## End = 8458
## Frequency = 1
## [1] -0.007118724 -0.016388074 -0.011979353 -0.014076244 -0.013078913

## Time Series:
## Start = 8454
## End = 8458
```

```
## [1] 0.001465583
```

##Model 2: ARIMA(3,1,3)

Frequency = 1

```
##
## Call:
\#\# arima(x = newdata(c4), order = auto.selection(newdata(c4), testdata(c4), totalobs(c4))[1:3])
##
## Coefficients:
##
           ar1
                  ar2
                          ar3
                                   ma1
                                           ma2
                                                   ma3
        0.9099 - 0.665 - 0.0029 - 1.9154 1.6164 - 0.701
                  NaN 0.0290
## s.e.
          NaN
                                   NaN
                                           NaN
                                                   NaN
##
## sigma^2 estimated as 9.313e-05: log likelihood = 3843.33, aic = -7674.66
```

I split the data in to 6 subsets, which are

```
## [1] "1986-08-28" "1989-01-12"
```

```
## [1] "1989-01-12" "1997-09-24"
```

```
## [1] "1997-09-24" "2002-07-05"
```

```
## [1] "2002-07-05" "2007-04-12"
```

```
## [1] "2007-04-12" "2011-11-01"
```

```
## [1] "2011-11-01" "2020-03-20"
```

and I use the fourth 2002-07-05 to 2007-04-12 to build the model.

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

```
## [1] 0.000947108 0.002282025 0.001623542 0.003072473 -0.006503511
```

```
## Time Series:
## Start = 1196
## End = 1200
## Frequency = 1
## [1] -9.326684e-05 -1.499534e-04 -4.794399e-04 -7.423593e-04 -7.623136e-04
```

```
## Time Series:
## Start = 1196
## End = 1200
## Frequency = 1
## [1] 0.001040375 0.002431978 0.002102982 0.003814832 -0.005741197
```

```
## [1] 1.178674e-05
```

##Model 3 ARIMA(1,1,0) I split the data in to 4 subsets, which are

```
## [1] "1986-08-28" "2002-07-05"
```

```
## [1] "2002-07-05" "2010-06-15"
```

```
## [1] "2010-06-15" "2016-10-20"
```

```
## [1] "2016-10-20" "2020-03-20"
```

```
## [1] "log(x)" "log(x)" "x" "x"
```

```
## [1] 0.013578683 -0.075429410 0.040299052 -0.035278267 0.002122879
```

```
## Time Series:

## Start = 2456

## End = 2460

## Frequency = 1

## [1] -0.004422225 -0.016434032 -0.009930468 -0.013451699 -0.011545195
```

```
## Time Series:
## Start = 2456
## End = 2460
## Frequency = 1
## [1] 0.01800091 -0.05899538 0.05022952 -0.02182657 0.01366807
```

```
## [1] 0.001398141
```

NEE, 1980-01-02 to 2020-03-20

##Model 1: ARIMA(1,1,0)

```
## ## Call:
## arima(x = newdata(a), order = auto.selection(newdata(a), testdata(a), totalobs(a))[1:3])
##
## Coefficients:
## arl
## -0.4972
## s.e. 0.0087
##
## sigma^2 estimated as 4.591e-05: log likelihood = 36237.8, aic = -72473.61
```

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

```
## [1] 0.041496593 -0.041888444 0.022342359 0.007360733 0.033444739
```

```
## Time Series:
## Start = 10137
## End = 10141
## Frequency = 1
## [1] 0.018249000 -0.003539060 0.007293670 0.001907783 0.004585574
```

```
## Time Series:

## Start = 10137

## End = 10141

## Frequency = 1

## [1] 0.02324759 -0.03834938 0.01504869 0.00545295 0.02885917
```

```
## [1] 0.000620035
```

##Model 2: ARIMA(0,1,2)

```
##
## Call:
## arima(x = newdata(c1), order = auto.selection(newdata(c1), testdata(c1), totalobs(c1))[1:3])
##
## Coefficients:
## mal ma2
## -0.9810 -0.0190
## s.e. 0.0142 0.0142
##
## sigma^2 estimated as 2.161e-05: log likelihood = 18548.01, aic = -37092.01
```

I split the data in to 3 subsets, which are

```
## [1] "1980-01-02" "1998-08-06"

## [1] "1998-08-06" "2011-09-15"

## [1] "2011-09-15" "2020-03-20"
```

and I use the first 1980-01-02 to 1998-08-06 to build the model.

the last 5 days actual value, forecast value, forecast errors and mean squared forecast error

```
## [1] -0.000873110  0.010146653 -0.007522544  0.007076486  0.000892575

## Time Series:
## Start = 4696
## End = 4700
## Frequency = 1
## [1] -0.0003302053 -0.0003106816 -0.0003106816 -0.0003106816 -0.0003106816
```

```
## Time Series:

## Start = 4696

## End = 4700

## Frequency = 1

## [1] -0.0005429047 0.0104573346 -0.0072118624 0.0073871676 0.0012032566
```

```
## [1] 4.353592e-05
```

##Model 3 ARIMA(1,1,0) I split the data in to 3 subsets, which are

```
## [1] "1980-01-02" "2006-02-24"

## [1] "2006-02-24" "2008-12-04"

## [1] "2008-12-04" "2020-03-20"
```

```
## [1] "x" "x" "x"
```

```
##
## Call:
## arima(x = newdata(c.new), order = auto.selection(newdata(c.new), testdata(c.new),
## totalobs(c.new))[1:3])
##
## Coefficients:
## ar1
## -0.4972
## s.e. 0.0087
##
## sigma^2 estimated as 4.639e-05: log likelihood = 36184.37, aic = -72366.73
```

```
## [1] 0.041496593 -0.041888444 0.022342359 0.007360733 0.033444739
```

```
## Time Series:
## Start = 10137
## End = 10141
## Frequency = 1
## [1] 0.018249417 -0.003539057 0.007293982 0.001907889 0.004585808
```

```
## Time Series:
## Start = 10137
## End = 10141
## Frequency = 1
## [1] 0.023247177 -0.038349387 0.015048377 0.005452843 0.028858931
```

```
## [1] 0.0006200263
```

TXN, 1980-01-02 to 2020-03-20

##Model 1: ARIMA(1,1,0)

```
##
## Call:
## arima(x = newdata(a), order = auto.selection(newdata(a), testdata(a), totalobs(a))[1:3])
##
## Coefficients:
## ar1
## -0.4657
## s.e. 0.0088
##
## sigma^2 estimated as 0.0001769: log likelihood = 29401.69, aic = -58801.37
```

```
## [1] 0.054658112 -0.055108396 0.025920007 -0.006765452 0.017315644
```

```
## Time Series:
## Start = 10137
## End = 10141
## Frequency = 1
## [1] -0.001619286 -0.017783644 -0.010256424 -0.013761608 -0.012129357
```

```
## Time Series:

## Start = 10137

## End = 10141

## Frequency = 1

## [1] 0.056277398 -0.037324752 0.036176431 0.006996156 0.029445001
```

```
## [1] 0.001356994
```

##Model 2: ARIMA(1,1,0)

```
## ## Call:
## arima(x = newdata(c1), order = auto.selection(newdata(c1), testdata(c1), totalobs(c1))[1:3])
##
## sigma^2 estimated as 0.00018: log likelihood = 11552.23, aic = -23104.47
```

I split the data in to 3 subsets, which are

```
## [1] "1980-01-02" "1995-10-26"
```

```
## [1] "1995-10-26" "2008-12-04"
```

```
## [1] "2008-12-04" "2020-03-20"
```

and I use the first 1980-01-02 to 1995-10-26 to build the model.

```
## [1] 0.021263205 -0.009240774 0.006912277 0.017373996 0.004047767
```

```
## Time Series:
## Start = 3996
## End = 4000
## Frequency = 1
## [1] 0.009529315 0.009529315 0.009529315 0.009529315
```

```
## Time Series:

## Start = 3996

## End = 4000

## Frequency = 1

## [1] 0.011733890 -0.018770089 -0.002617038 0.007844681 -0.005481548
```

```
## [1] 0.0001176871
```

##Model 3 ARIMA(0,0,0) I split the data in to 4 subsets, which are

```
## [1] "1980-01-02" "1995-10-26"

## [1] "1995-10-26" "2000-05-17"

## [1] "2000-05-17" "2003-03-05"

## [1] "2003-03-05" "2020-03-20"
```

```
## [1] "x" "log(x)" "x"

##
```

```
## [1] 0.054658112 -0.055108396 0.025920007 -0.006765452 0.017315644
```

```
## Time Series:
## Start = 8289
## End = 8293
## Frequency = 1
## [1] -0.0002150617 -0.0002150617 -0.0002150617 -0.0002150617
```

```
## Time Series:
## Start = 8289
## End = 8293
## Frequency = 1
## [1] 0.054873174 -0.054893334 0.026135068 -0.006550391 0.017530705
```

```
## [1] 0.001411524
```