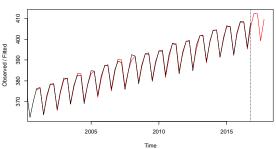
# STATS 326 Applied Time Series ASSIGNMENT TWO R & MARKING GUIDE

**Question One:** (20 marks)

# **Holt-Winters:**

```
> HW.fit = HoltWinters(red.CO2.ts)
Holt-Winters exponential smoothing with trend and additive seasonal
component.
Call:
HoltWinters(x = red.CO2.ts)
Smoothing parameters:
alpha: 0.5732177
beta : 0.03867122
gamma: 0.4140589
Coefficients:
         [,1]
  406.4502058
b
    0.5868941
   5.3662780
s1
    4.6109173
s3 -8.8837987
    0.7093982
> HW.pred = predict(HW.fit,n.ahead=4)
> HW.pred
         Qtr1
                 Qtr2
                           Qtr3
2017 412.4034 412.2349 399.3271 409.5072
> HW.RMSEP = sqrt(1/4*sum((actual-HW.pred)^2))
> HW.RMSEP
[1] 0.9645109
> plot(HW.fit, HW.pred, main="Barrow, Alaska CO2 - Holt-Winters")
```

#### Barrow, Alaska CO2 - Holt-Winters

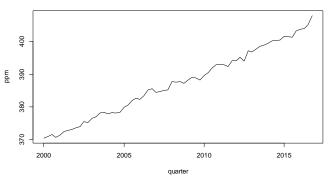


# **Question Two:** (30 marks)

# **MA Seasonally Adjusted:**

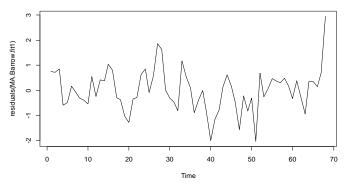
- > MA.Barrow = decompose(red.CO2.ts)
- > MA.Barrow\$figure
- [1] 4.9286328 4.3242578 -9.0356641 -0.2172266
- > MA.CO2.ts = red.CO2.ts-MA.Barrow\$seasonal
- > plot(MA.CO2.ts,main="MA Seasonally Adjusted Barrow CO2",xlab="quarter",ylab="ppm")

# MA Seasonally Adjusted Barrow CO2



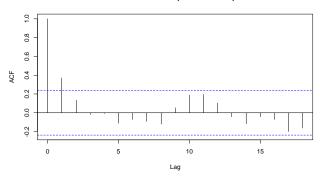
- > Time = 1:68
- > MA.Barrow.fit1 = lm(MA.CO2.ts~Time)
- > plot.ts(residuals(MA.Barrow.fit1),main="Residual Series")

#### **Residual Series**



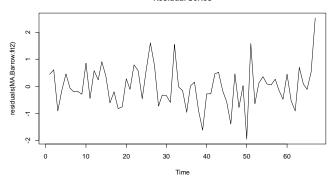
#### > acf(residuals(MA.Barrow.fit1))

#### Series residuals(MA.Barrow.fit1)



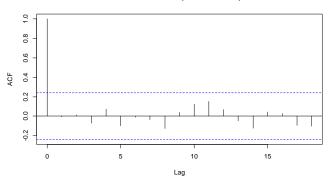
- > MA.Barrow.fit2 = lm(MA.CO2.ts[-1]~Time[-1]+MA.CO2.ts[-68])
- > plot.ts(residuals(MA.Barrow.fit2),main="Residual Series")

#### **Residual Series**



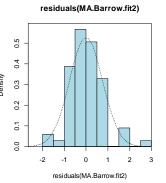
> acf(residuals(MA.Barrow.fit2))

#### Series residuals(MA.Barrow.fit2)



> normcheck(residuals(MA.Barrow.fit2),shapiro.wilk=T)

# Normal Q-Q Plot Shapiro-Wilk normality test W = 0.9732P-value = 0.157 0 0 -2 -1 Theoretical Quantiles



> summary(MA.Barrow.fit2)

```
Call:
lm(formula = MA.CO2.ts[-1] \sim Time[-1] + MA.CO2.ts[-68])
Residuals:
```

10 Median Min -1.9475 -0.4709 -0.0720 0.4565 2.5262

#### Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
              198.50565 46.50963 4.268 6.65e-05 ***
(Intercept)
Time[-1]
                0.28589
                          0.06611
                                   4.324 5.46e-05 ***
MA.CO2.ts[-68]
              0.46263
                         0.12615
                                  3.667
                                           5e-04 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 0.7696 on 64 degrees of freedom

Multiple R-squared: 0.9946, Adjusted R-squared: 0.9944 F-statistic: 5897 on 2 and 64 DF, p-value: < 2.2e-16

> t.69.ma.pred = MA.Barrow.fit2\$coef[1]+MA.Barrow.fit2\$coef[2]\*69+ MA.Barrow.fit2\$coef[3]\*MA.CO2.ts[68] > t.69.ma.pred

```
(Intercept)
  406.9338
> t.69.pred = t.69.ma.pred+MA.Barrow$figure[1]
> t.69.pred
```

(Intercept) 411.8624

> t.70.ma.pred = MA.Barrow.fit2\$coef[1]+MA.Barrow.fit2\$coef[2]\*70+ MA.Barrow.fit2\$coef[3]\*t.69.ma.pred

> t.70.ma.pred

(Intercept) 406.7786

> t.70.pred = t.70.ma.pred+MA.Barrow\$figure[2]

> t.70.pred (Intercept)

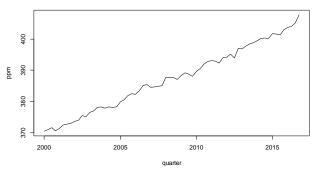
411.1028

```
> t.71.ma.pred = MA.Barrow.fit2$coef[1]+MA.Barrow.fit2$coef[2]*71+
  MA.Barrow.fit2$coef[3]*t.70.ma.pred
> t.71.ma.pred
(Intercept)
  406.9927
> t.71.pred = t.71.ma.pred+MA.Barrow$figure[3]
> t.71.pred
(Intercept)
   397.957
> t.72.ma.pred = MA.Barrow.fit2$coef[1]+MA.Barrow.fit2$coef[2]*72+
  MA.Barrow.fit2$coef[3]*t.71.ma.pred
> t.72.ma.pred
(Intercept)
  407.3776
> t.72.pred = t.72.ma.pred+MA.Barrow$figure[4]
> t.72.pred
(Intercept)
  407.1604
> MA.pred = c(t.69.pred,t.70.pred,t.71.pred,t.72.pred)
> names(MA.pred) = c("2017.1","2017.2","2017.3","2017.4")
> MA.pred
 2017.1 2017.2 2017.3 2017.4
411.8624 411.1028 397.9570 407.1604
> RMSEP.MA.Barrow = sqrt(1/4*sum((actual-MA.pred)^2))
> RMSEP.MA.Barrow
[1] 1.341502
```

# **STL Seasonally Adjusted:**

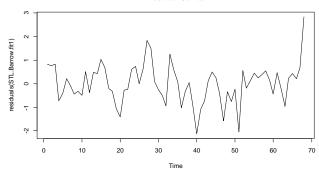
- > STL.Barrow = stl(red.CO2.ts,s.window="periodic")
- > STL.Barrow\$time.series[1:4,1]
- [1] 4.85060732 4.26390951 -9.02024922 -0.09426758
- > STL.CO2.ts = red.CO2.ts-STL.Barrow\$time.series[,1]
- > plot(STL.CO2.ts,main="STL Seasonally Adjusted Barrow CO2",xlab="quarter",ylab="ppm")

# STL Seasonally Adjusted Barrow CO2



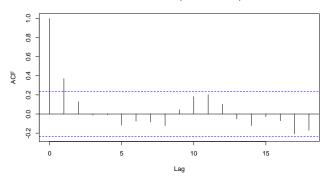
- > STL.Barrow.fit1 = lm(STL.CO2.ts~Time)
- > plot.ts(residuals(STL.Barrow.fit1),main="Residual Series")

#### Residual Series



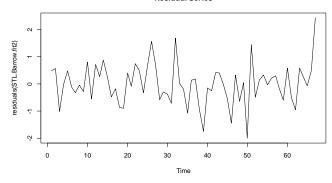
> acf(residuals(STL.Barrow.fit1))

#### Series residuals(STL.Barrow.fit1)



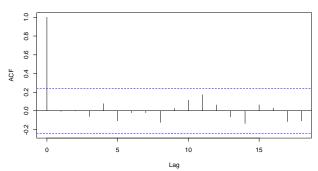
- > STL.Barrow.fit2 = lm(STL.CO2.ts[-1]~Time[-1]+STL.CO2.ts[-68])
- > plot.ts(residuals(STL.Barrow.fit2),main="Residual Series")

# Residual Series



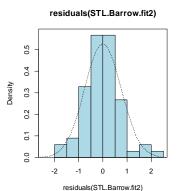
> acf(residuals(STL.Barrow.fit2))

## Series residuals(STL.Barrow.fit2)



> normcheck(residuals(STL.Barrow.fit2),shapiro.wilk=T)

# 



> summary(STL.Barrow.fit2)

#### Call:

 $lm(formula = STL.CO2.ts[-1] \sim Time[-1] + STL.CO2.ts[-68])$ 

#### Residuals:

Min 1Q Median 3Q Max -1.98853 -0.48582 -0.01539 0.44476 2.42309

Theoretical Quantiles

# Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 201.76350 46.10705 4.376 4.56e-05 \*\*\*
Time[-1] 0.29041 0.06553 4.432 3.74e-05 \*\*\*
STL.CO2.ts[-68] 0.45380 0.12505 3.629 0.000566 \*\*\*
--Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.7692 on 64 degrees of freedom Multiple R-squared: 0.9946, Adjusted R-squared: 0.9944 F-statistic: 5901 on 2 and 64 DF, p-value: < 2.2e-16

```
> t.69.stl.pred = STL.Barrow.fit2$coef[1]+STL.Barrow.fit2$coef[2]*69+
  STL.Barrow.fit2$coef[3]*STL.CO2.ts[68]
> t.69.stl.pred
(Intercept)
   406.8451
> t.69.pred = t.69.stl.pred+STL.Barrow$time.series[1,1]
> t.69.pred
(Intercept)
   411.6957
> t.70.stl.pred = STL.Barrow.fit2$coef[1]+STL.Barrow.fit2$coef[2]*70+
  STL.Barrow.fit2$coef[3]*t.69.stl.pred
> t.70.stl.pred
(Intercept)
   406.7183
> t.70.pred = t.70.stl.pred+STL.BarrowStime.series[2,1]
> t.70.pred
(Intercept)
   410.9822
> t.71.stl.pred = STL.Barrow.fit2$coef[1]+STL.Barrow.fit2$coef[2]*71+
  STL.Barrow.fit2$coef[3]*t.70.stl.pred
> t.71.stl.pred
(Intercept)
   406.9512
> t.71.pred = t.71.stl.pred+STL.Barrow$time.series[3,1]
> t.71.pred
(Intercept)
   397.931
> t.72.stl.pred = STL.Barrow.fit2$coef[1]+STL.Barrow.fit2$coef[2]*72+
  STL.Barrow.fit2$coef[3]*t.71.stl.pred
> t.72.stl.pred
(Intercept)
   407.3473
> t.72.pred = t.72.stl.pred+STL.Barrow$time.series[4,1]
> t.72.pred
(Intercept)
   407.2531
> STL.pred = c(t.69.pred,t.70.pred,t.71.pred,t.72.pred)
> names(STL.pred) = c("2017.1","2017.2","2017.3","2017.4")
> STL.pred
 2017.1 2017.2 2017.3 2017.4
411.6957 410.9822 397.9310 407.2531
> RMSEP.STL.Barrow = sqrt(1/4*sum((actual-STL.pred)^2))
> RMSEP.STL.Barrow
[1] 1.413046
Question Three: (30 marks)
```

Tech Notes for MA Seasonally Adjusted model.

**Question Four:** (20 marks)

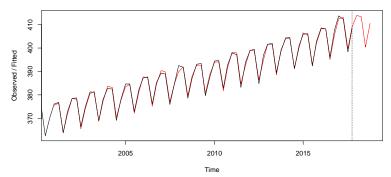
# **Holt-Winters (Full):**

```
> HW.Full.fit = HoltWinters(full.CO2.ts)
```

```
> HW.Full.fit
Holt-Winters exponential smoothing with trend and additive seasonal
component.
Call:
HoltWinters(x = full.CO2.ts)
Smoothing parameters:
alpha: 0.6655127
beta: 0.02583596
gamma: 0.5504329
Coefficients:
         [,1]
a 407.8598399
  0.5251393
sl 5.6488674
s2 4.5145080
s3 -9.0761435
s4 0.6730279
> HW.Full.pred = predict(HW.Full.fit,n.ahead=4)
> HW.Full.pred
        Otr1
                 Qtr2
                          Qtr3
2018 414.0338 413.4246 400.3591 410.6334
```

## Barrow, Alaska CO2 - Holt-Winters

> plot(HW.Full.fit,HW.Full.pred,main="Barrow, Alaska CO2 - Holt-Winters")



```
> HW.full.pred.int = predict(HW.Full.fit,n.ahead=4,prediction.interval=T)
> HW.full.pred.int
             fit
                      upr
2018 01 414.0338 415.8081 412.2596
2018 02 413.4246 415.5729 411.2763
2018 03 400.3591 402.8405 397.8778
2018 04 410.6334 413.4219 407.8449
> 415.8081-412.2596
[1] 3.5485
> 415.5729-411.2763
[1] 4.2966
> 402.8405-397.8778
[1] 4.9627
> 413.4219-407.8449
[1] 5.577
```