

HW4

2021-09-22

Question 7.1:

We could use smoothing model in forecasting tourism in developing countries where there isn't a clear trend to use moving average method, I would choose alpha depending on the country and historical data, most likely α will be closer to 1. Unless the country has come out of political turmoil and instability then α will likely be smaller.

Question 7.2 is in R script below:

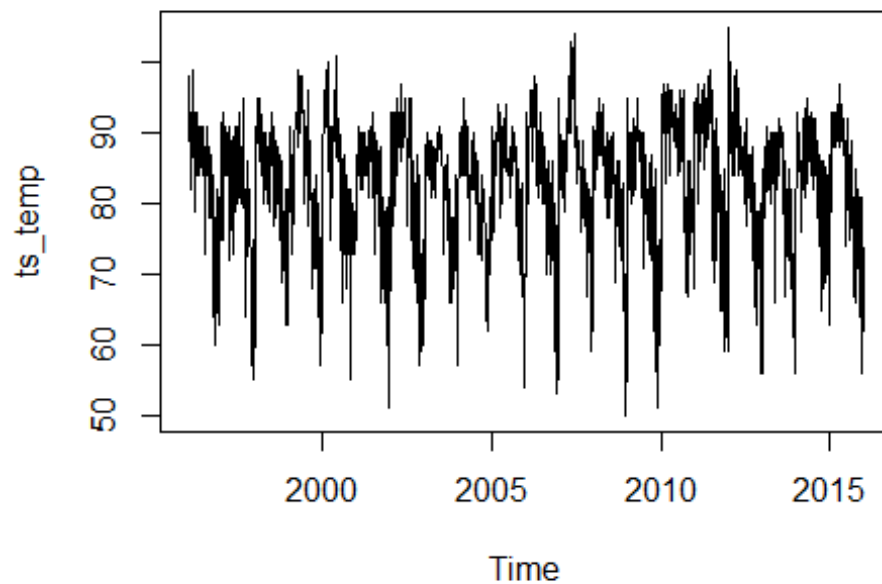
```
library(tidyverse)

#Q7.2
set.seed(1)

temp_data<- read.table("temps.txt", header = TRUE)
names(temp_data) = gsub(pattern = "X", replacement = "", x = names(temp_data)
)
temp_data <- data.frame(unlist(temp_data[,2:21]))
ts_temp= ts(temp_data, frequency = 123, start=1996)
summary(ts_temp)

## unlist.temp_data...2.21..
## Min.    : 50.00
## 1st Qu.: 79.00
## Median : 85.00
## Mean   : 83.34
## 3rd Qu.: 90.00
## Max.    :105.00

ts.plot(ts_temp)
```

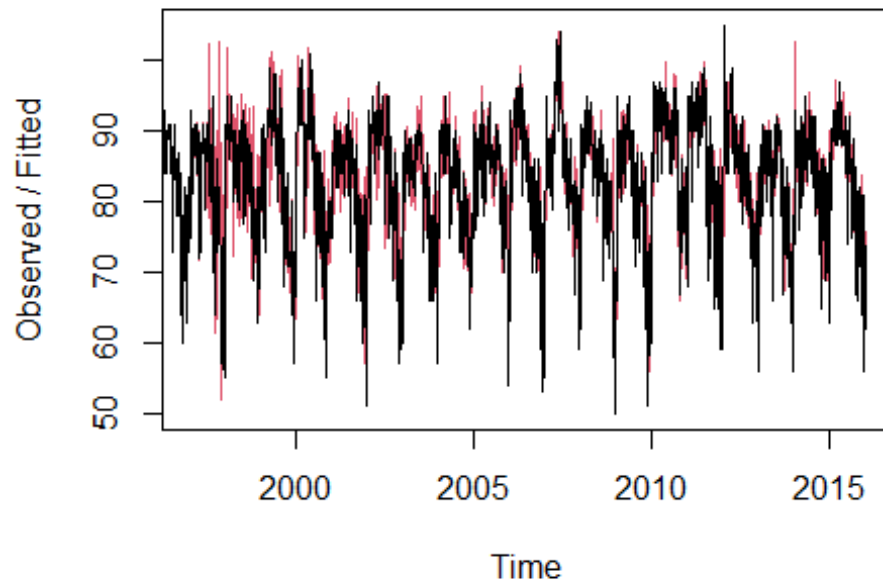


```
Model_HW_Mult<-HoltWinters(ts_temp, seasonal="multiplicative")
summary(Model_HW_Mult)

##           Length Class  Mode
## fitted      9348   mts    numeric
## x           2460    ts     numeric
## alpha         1  -none-  numeric
## beta          1  -none-  numeric
## gamma         1  -none-  numeric
## coefficients  125  -none-  numeric
## seasonal      1  -none-  character
## SSE           1  -none-  numeric
## call          3  -none-  call

plot(Model_HW_Mult)
```

Holt-Winters filtering

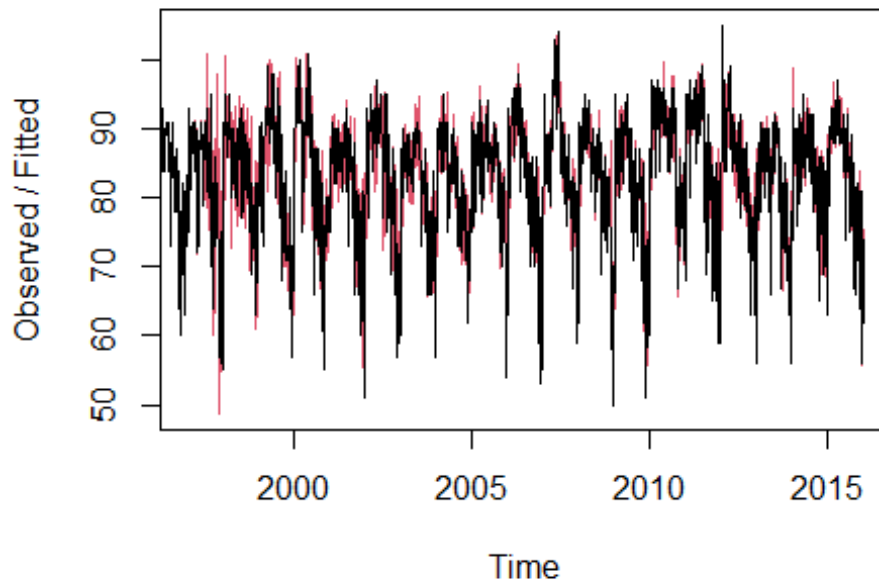


```
Model_HW_ADD<-HoltWinters(ts_temp, seasonal="additive")
summary(Model_HW_ADD)

##           Length Class  Mode
## fitted      9348   mts    numeric
## x           2460    ts     numeric
## alpha         1   -none-  numeric
## beta          1   -none-  numeric
## gamma         1   -none-  numeric
## coefficients  125   -none-  numeric
## seasonal      1   -none-  character
## SSE           1   -none-  numeric
## call          3   -none-  call

plot(Model_HW_ADD)
```

Holt-Winters filtering



```
cat("Sum of squared Errors in Multiplicative model",Model_HW_Mult$SSE,"\n")
```

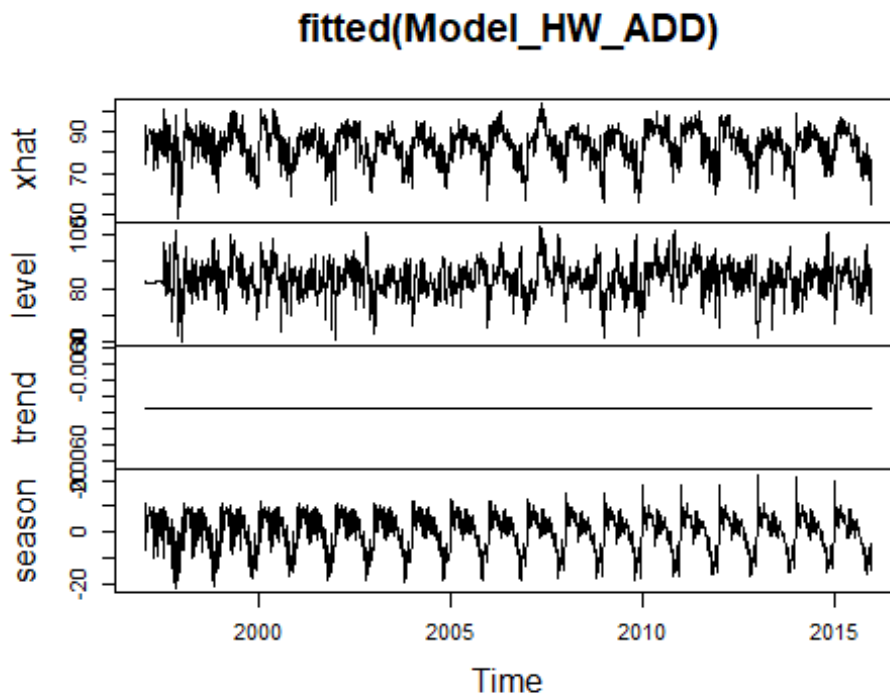
```
## Sum of squared Errors in Multiplicative model 68904.57
```

```
cat("Sum of squared Errors in Additive model",Model_HW_ADD$SSE,"\n")
```

```
## Sum of squared Errors in Additive model 66244.25
```

#Additive model has smaller SSE hence I am choosing additive seasonality, better way to do it would be using cross-validation but for now I am choosing seasonality as additive

#Lets visualize the trend to understand if the temperature is increasing
`plot(fitted(Model_HW_ADD))`



We can observe that there is no visible change in trend over time in the temperature

I think one of the reason why it's hard to see the change over time in temperature is because the temperature is increasing very slowly

and the changes would be very minute, additionally our data is too small to capture that since there are random fluctuations in yearly temperature as well such as heatwaves etc.

Let us further use cusum to determine if there is change in seasonality over the years

```
Seasonality_data<- matrix(Model_HW_ADD$fitted[,4],nrow = 123)
```

```
temp_data1<- read.table("temps.txt", header = TRUE)
names(temp_data1) = gsub(pattern = "X", replacement = "", x = names(temp_data1))
```

```
dim(Seasonality_data)
```

```
## [1] 123 19
```

```
dim(temp_data1)
```

```
## [1] 123 21
```

```
colnames(Seasonality_data)<-c(1997:2015)
```

```
days <- temp_data1[,1]
```

```

rownames(Seasonality_data)<-days

cumsum_data <- data.frame(matrix(0, nrow=123, ncol=19))
rownames(cumsum_data) <-days
colnames(cumsum_data)<-c(1997:2015)

mean_year1= mean(Seasonality_data[,1])
var=sd(Seasonality_data[,1])
day_counter=1
threshold= var*3  #(Using threshold for change as 3 times SD, as I want to give some room for variation)

for(i in 1:ncol(cumsum_data)){
  date_change=NULL
  for (j in 1:nrow(cumsum_data)){
    val=Seasonality_data[j,i]
    cumsum_data[j,i] <- max(0, cumsum_data[j-1,i] + (mean_year1 - val- var*0.5))

    if (cumsum_data[j,i]>=threshold){

      cat("Year",colnames(cumsum_data[i]),"Day winter started is:",
          rownames(cumsum_data)[j],"Day","\n")
      break
    }

  }
}

## Year 1997 Day winter started is: 30-Sep Day
## Year 1998 Day winter started is: 1-Oct Day
## Year 1999 Day winter started is: 1-Oct Day
## Year 2000 Day winter started is: 1-Oct Day
## Year 2001 Day winter started is: 2-Oct Day
## Year 2002 Day winter started is: 2-Oct Day
## Year 2003 Day winter started is: 4-Oct Day
## Year 2004 Day winter started is: 3-Oct Day
## Year 2005 Day winter started is: 4-Oct Day
## Year 2006 Day winter started is: 5-Oct Day
## Year 2007 Day winter started is: 6-Oct Day
## Year 2008 Day winter started is: 6-Oct Day
## Year 2009 Day winter started is: 6-Oct Day
## Year 2010 Day winter started is: 5-Oct Day
## Year 2011 Day winter started is: 5-Oct Day
## Year 2012 Day winter started is: 5-Oct Day
## Year 2013 Day winter started is: 5-Oct Day

```

```
## Year 2014 Day winter started is: 6-Oct Day  
## Year 2015 Day winter started is: 6-Oct Day
```

#According to my CUSUM model with C of 0.5 SD and threshold of 3 times SD, on seasonality measure from our Holt Winters model, it appears that summer end is delayed slowly over the years, it ends with 30th September in 1997

#but for 2015 it ends in 6th October

#Note that we can't say if summer is longer or shorter because we don't have data to know if summer started earlier or delayed

#but according to our model it is gradually delayed