

## Question 7.2

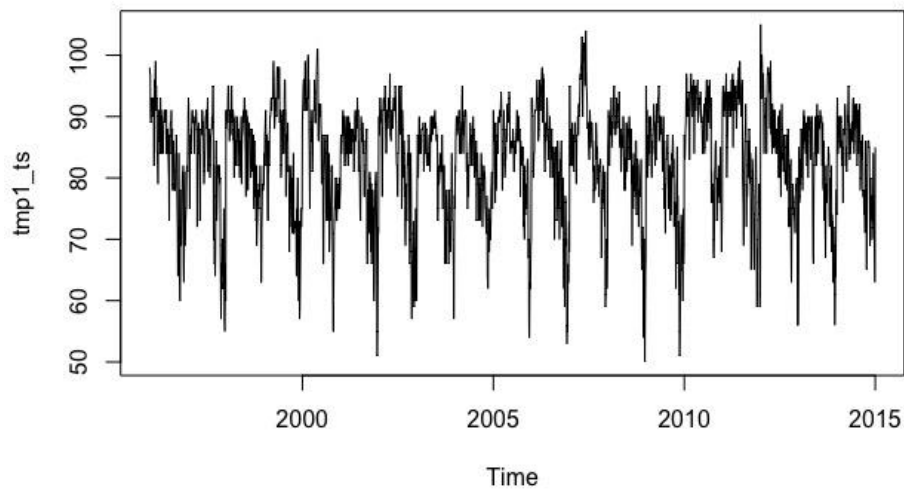
Using the 20 years of daily high temperature data for Atlanta (July through October) from Question 6.2 (file `temps.txt`), build and use an exponential smoothing model to help make a judgment of whether the unofficial end of summer has gotten later over the 20 years. (Part of the point of this assignment is for you to think about how you might use exponential smoothing to answer this question. Feel free to combine it with other models if you'd like to. There's certainly more than one reasonable approach.)

Note: in R, you can use either `HoltWinters` (simpler to use) or the `smooth` package's `es` function (harder to use, but more general). If you use `es`, the Holt-Winters model uses `model="AAM"` in the function call (the first and second constants are used "A"dditively, and the third (seasonality) is used "M"ultiplicatively; the documentation doesn't make that clear).

```
# Single Exponential Smoothness(ES) Values:
# alpha: 0.8396301
# sum of squared error(SSE):53704.15

# Double ES Values
# alpha: 0.8455303
# beta : 0.003777803
# SSE: 54071.22

# Triple ES Values:
# alpha: 0.6677614
# beta : 0
# gamma: 0.6297674
# SSE: 63025.97
```



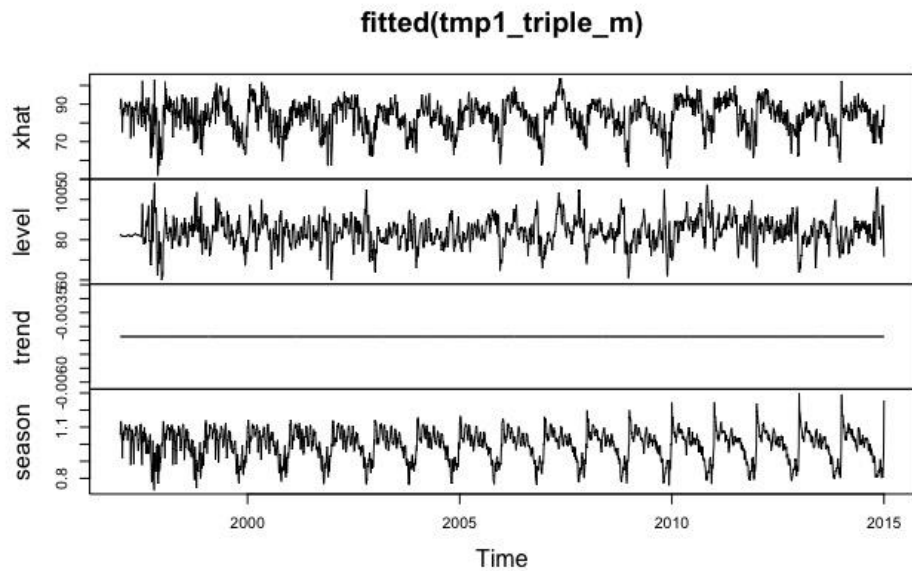
#Single ES has smallest SSE when compared to other two and it is closed to 1. It means there is less randomness in the system and recent temperature reading has more weight in predicting the current temperature.

#Seasonality uses Model AAM:

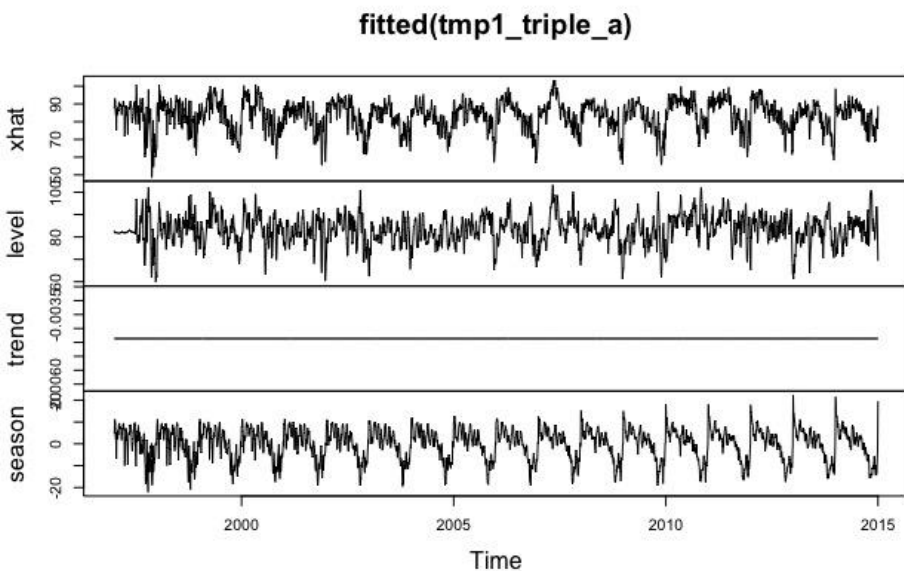
- #1. A - additive: seasonal variation is independent
- #2. M - multiplicative: seasonal variation is connected.

#Triple Exponential – using multiplicative

#SSE:65648.65



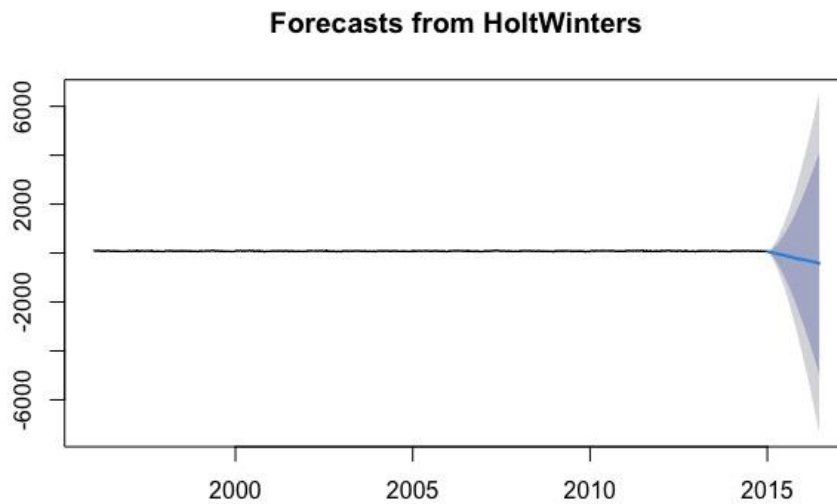
#Triple Exponential - using additive  
#SSE:63025.97



Triple ES with additive seasonal factor has better SSE.

From the above chart, Trend line is straight which means no trend.

The season sub chart shows that the duration of each season has been constant throughout all 20 years.



The above plot reflects the actual.. Use Chkum formula from previous lesson  $St = \max(0, St-1 + (X_{mean} - T - C))$  as a running equation across each daily temperature observation there wasn't much of a change.

## I. R Script

```
## Clear Work Area ##
rm(list=ls()) #clear environment
cat("\014") #clear console

#Load data from temperature table
getwd()

##Input data
tmp1_data <- read.table("temps.txt", header = TRUE) #read txt data into a table

#Exploring the data
tail(tmp_data)[1:5,]
str(tmp1_data)
print(summary(tmp1_data))

#Marix of temperature data
tmp1_mat <- as.vector(unlist(tmp1_data[,2:21]))
str(tmp1_mat)
tmp1_mat

# Time Series Values
```

```
tmp1_ts <- ts(tmp1_mat, start=1996, end = 2015, frequency=123)
tmp1_ts

class(tmp1_ts)
plot(tmp1_ts)

# Define beta to generate trend value to forecast a future trend
tmp1_b_ts <- HoltWinters(tmp1_ts , beta=.5)
plot(tmp1_b_ts)

# Exponential Smoothing
#Simple Exponential #
tmp1_single <- HoltWinters(tmp1_ts,beta=FALSE, gamma=FALSE)

#Double Exponential - model trend #
tmp1_double <- HoltWinters(tmp1_ts,gamma=FALSE)

#Triple Exponential - model trend and seasonality#
tmp1_triple <- HoltWinters(tmp1_ts, seasonal = "additive")

#Compare 3 kinds of Exp Smoothing#
tmp1_single
tmp1_single$SSE # Sum of Squared Error

tmp1_double
tmp1_double$SSE

tmp1_triple
tmp1_triple$SSE

# Triple Exponential using multiplicative ( Seasonal variation is connected)
tmp1_triple_m <- HoltWinters(tmp1_ts, seasonal = "multiplicative")
tmp1_triple_m$SSE

# Triple Exponential using multiplicative ( Seasonal variation is independent of the level)
tmp1_triple_a <- HoltWinters(tmp1_ts, seasonal = "additive")
tmp1_triple_a$SSE

tmp1_triple_a$fitted

plot(fitted(tmp1_triple_a))

plot(fitted(tmp1_triple_m))

#Forecast
```

```
library(forecast)
tmp1_forecast=predict( tmp1_b_ts, n.ahead = 100, prediction.interval = TRUE )
plot(forecast( tmp1_b_ts, h = 180 ))
```

## R-Script With Answers

```
>
> #Load data from temperature table
> getwd()
[1] "/Users/nagarajanmurugan/Documents/MS/Data"
>
> ##Input data
> tmp1_data <- read.table("temps.txt", header = TRUE) #read txt data into a table
>
>
> #Exploring the data
> str(tmp1_data)
'data.frame': 123 obs. of 21 variables:
 $ DAY : chr "1-Jul" "2-Jul" "3-Jul" "4-Jul" ...
 $ X1996: int 98 97 97 90 89 93 93 91 93 93 ...
 $ X1997: int 86 90 93 91 84 84 75 87 84 87 ...
 $ X1998: int 91 88 91 91 91 89 93 95 95 91 ...
 $ X1999: int 84 82 87 88 90 91 82 86 87 87 ...
 $ X2000: int 89 91 93 95 96 96 96 91 96 99 ...
 $ X2001: int 84 87 87 84 86 87 87 89 91 87 ...
 $ X2002: int 90 90 87 89 93 93 89 89 90 91 ...
 $ X2003: int 73 81 87 86 80 84 87 90 89 84 ...
 $ X2004: int 82 81 86 88 90 90 89 87 88 89 ...
 $ X2005: int 91 89 86 86 89 82 76 88 89 78 ...
 $ X2006: int 93 93 93 91 90 81 80 82 84 84 ...
 $ X2007: int 95 85 82 86 88 87 82 82 89 86 ...
```

```

$ X2008: int 85 87 91 90 88 82 88 90 89 87 ...
$ X2009: int 95 90 89 91 80 87 86 82 84 84 ...
$ X2010: int 87 84 83 85 88 89 94 97 96 90 ...
$ X2011: int 92 94 95 92 90 90 94 94 91 92 ...
$ X2012: int 105 93 99 98 100 98 93 95 97 95 ...
$ X2013: int 82 85 76 77 83 83 79 88 88 87 ...
$ X2014: int 90 93 87 84 86 87 89 90 90 87 ...
$ X2015: int 85 87 79 85 84 84 90 90 91 93 ...
> print(summary(tmp1_data))
  DAY      X1996      X1997      X1998      X1999
Length:123      Min. :60.00 Min. :55.00 Min. :63.00 Min. :57.00
Class :character 1st Qu.:79.00 1st Qu.:78.50 1st Qu.:79.50 1st Qu.:75.00
Mode :character  Median :84.00  Median :84.00  Median :86.00  Median :86.00
      Mean :83.72  Mean :81.67  Mean :84.26  Mean :83.36
      3rd Qu.:90.00 3rd Qu.:88.50 3rd Qu.:89.00 3rd Qu.:91.00
      Max. :99.00  Max. :95.00  Max. :95.00  Max. :99.00
  X2000      X2001      X2002      X2003      X2004      X2005
Min. :55.00 Min. :51.00 Min. :57.00 Min. :57.00 Min. :62.00 Min. :54.00
1st Qu.: 77.00 1st Qu.:78.00 1st Qu.:78.00 1st Qu.:78.00 1st Qu.:78.00 1st Qu.:81.50
Median : 86.00 Median :84.00 Median :87.00 Median :84.00 Median :82.00 Median :85.00
Mean : 84.03  Mean :81.55  Mean :83.59  Mean :81.48  Mean :81.76  Mean :83.36
3rd Qu.: 91.00 3rd Qu.:87.00 3rd Qu.:91.00 3rd Qu.:87.00 3rd Qu.:87.00 3rd Qu.:88.00
Max. :101.00 Max. :93.00  Max. :97.00  Max. :91.00  Max. :95.00  Max. :94.00
  X2006      X2007      X2008      X2009      X2010      X2011
Min. :53.00 Min. :59.0 Min. :50.00 Min. :51.00 Min. :67.00 Min. :59.00
1st Qu.:79.00 1st Qu.: 81.0 1st Qu.:79.50 1st Qu.:75.00 1st Qu.:82.00 1st Qu.:79.00
Median :85.00 Median : 86.0 Median :85.00 Median :83.00 Median :90.00 Median :89.00
Mean :83.05  Mean : 85.4  Mean :82.51  Mean :80.99  Mean :87.21  Mean :85.28
3rd Qu.:91.00 3rd Qu.: 89.5 3rd Qu.:88.50 3rd Qu.:88.00 3rd Qu.:93.00 3rd Qu.:94.00
Max. :98.00 Max. :104.0 Max. :95.00 Max. :95.00 Max. :97.00 Max. :99.00
  X2012      X2013      X2014      X2015
Min. :56.00 Min. :56.00 Min. :63.00 Min. :56.0
1st Qu.: 79.50 1st Qu.:77.00 1st Qu.:81.50 1st Qu.:77.0
Median : 85.00 Median :84.00 Median :86.00 Median :85.0
Mean : 84.65  Mean :81.67  Mean :83.94  Mean :83.3
3rd Qu.: 90.50 3rd Qu.:88.00 3rd Qu.:89.00 3rd Qu.:90.0
Max. :105.00 Max. :92.00  Max. :95.00  Max. :97.0
>
> #Marix of temperature data
> tmp1_mat <- as.vector(unlist(tmp1_data[,2:21]))
> str(tmp1_mat)
int [1:2460] 98 97 97 90 89 93 93 91 93 93 ...
> tmp1_mat
 [1] 98 97 97 90 89 93 93 91 93 93 90 91 93 93 82 91 96 95 96 99 91 95 91
[24] 93 84 84 82 79 90 91 87 86 90 84 91 93 88 91 84 90 89 88 86 84 86 89
[47] 90 91 91 90 89 90 91 91 91 84 88 84 86 88 84 82 80 73 87 84 87 89 89
[70] 89 91 84 86 88 78 79 86 82 82 78 79 79 78 81 84 84 87 84 79 75 72 64

```

```
[93] 66 72 84 70 66 64 60 78 70 72 69 69 73 79 81 80 82 66 63 68 79 81 69
[116] 73 73 75 75 81 82 82 81 86 90 93 91 84 84 75 87 84 87 84 88 86 90 91
[139] 91 89 89 89 90 89 84 87 88 89 89 91 91 89 88 72 80 84 88 89 88 84 84
[162] 80 73 80 86 88 88 87 88 91 91 89 89 88 82 79 81 82 84 87 90 90 91 91
[185] 88 88 91 93 81 81 82 86 88 84 80 82 86 87 87 88 88 90 88 91 95 89 70
[208] 80 82 66 70 64 68 77 86 75 73 75 78 81 82 82 82 80 82 82 79 80 68 63
[231] 57 66 64 69 70 70 62 63 62 75 71 57 55 64 66 60 91 88 91 91 91 89 93
[254] 95 95 91 91 86 88 87 91 87 90 91 95 91 91 89 91 91 86 88 80 88 89 90
[277] 86 86 82 84 86 90 89 89 86 82 87 88 84 86 80 82 86 84 87 90 79 84 87
[300] 87 88 90 91 89 90 93 93 91 87 84 77 90 91 89 90 89 79 78 81 84 89 87
[323] 87 88 87 82 80 82 82 88 84 81 82 84 87 80 75 75 86 78 77 82 82 73 82
[346] 69 72 73 78 78 78 75 79 78 77 78 82 75 73 63 63 72 75 79 79 79 78 82
[369] 79 84 82 87 88 90 91 82 86 87 87 82 77 73 81 81 86 82 87 88 90 90 91
[392] 93 93 91 93 93 93 93 97 99 96 93 88 89 91 93 93 93 91 90 96 98 97 98
[415] 93 93 96 98 98 89 91 91 90 80 82 89 88 90 91 91 84 88 91 84 93 96 96
[438] 91 91 77 87 87 87 86 87 89 81 81 82 79 68 79 72 75 78 81 82 78 80 77
[461] 71 73 75 84 71 73 71 73 73 72 72 73 70 64 75 73 77 80 71 66 60 64 73
[484] 57 59 64 69 75 73 72 75 75 89 91 93 95 96 96 96 91 96 99 96 93 91 93
[507] 93 93 91 97 100 99 93 96 87 82 75 82 88 91 89 87 86 86 81 84 88 91 91
[530] 91 91 96 95 89 89 89 89 94 97 99 101 101 97 87 86 88 92 92 90 90 92 92
[553] 88 87 79 81 82 87 81 66 66 75 80 82 84 86 87 86 80 75 73 73 84 87 77
[576] 73 81 84 82 68 71 75 73 75 77 79 82 81 82 73 66 55 55 64 71 73 75 75
[599] 77 80 80 80 73 73 75 79 75 75 78 75 78 80 75 77 78 84 87 87 84 86 87
[622] 87 89 91 87 90 90 86 82 82 84 87 88 90 87 84 87 90 84 82 88 90 84 89
[645] 89 87 84 84 84 86 88 84 86 88 87 88 86 86 81 87 84 90 91 91 87 86 88
[668] 90 88 93 90 91 91 81 86 81 82 80 75 73 81 90 88 87 86 86 89 87 84 84
[691] 86 77 77 81 81 82 84 86 87 88 69 66 72 75 78 71 71 75 80 81 80 79 70
[714] 68 79 66 73 75 78 78 75 75 62 60 64 71 75 79 80 81 79 73 64 51 55 63
[737] 72 71 90 90 87 89 93 93 89 89 90 91 84 77 82 88 91 93 93 93 93 91 95
[760] 91 89 87 84 86 89 91 91 88 90 93 91 91 91 93 97 87 87 86 88 89 91 91
[783] 89 88 90 91 93 91 93 93 91 95 93 91 88 84 82 82 78 77 84 84 89 95 93
[806] 91 88 87 91 95 95 90 75 78 91 88 86 81 80 86 84 77 82 73 69 75 75 79
[829] 73 79 82 84 84 82 87 86 80 71 66 70 78 84 79 68 57 66 64 68 71 73 71
[852] 64 59 68 60 68 69 75 75 68 60 73 81 87 86 80 84 87 90 89 84 84 86 87
[875] 84 86 88 88 88 88 88 89 86 81 82 84 87 87 89 88 84 88 84 84 84 82 84
[898] 82 84 84 86 87 84 81 87 89 90 86 89 90 90 87 88 88 90 89 88 89 90 91
[921] 89 88 89 88 86 87 87 84 73 75 81 82 79 80 81 84 82 82 81 81 81 84 87
[944] 82 75 81 80 82 82 82 73 66 71 72 68 66 77 78 75 73 73 73 73 66 78 78
[967] 78 69 72 68 70 75 78 84 78 78 73 73 68 64 57 70 77 75 82 81 86 88 90
[990] 90 89 87 88 89 90 89 91 91 84 84
[ reached getOption("max.print") -- omitted 1460 entries ]
>
> # Time Series Values
> tmp1_ts <- ts(tmp1_mat, start=1996, end = 2015, frequency=123)
> tmp1_ts
Time Series:
Start = c(1996, 1)
```



End = c(2015, 1)

Frequency = 123

```
[1] 98 97 97 90 89 93 93 91 93 93 90 91 93 93 82 91 96 95 96 99 91 95 91
[24] 93 84 84 82 79 90 91 87 86 90 84 91 93 88 91 84 90 89 88 86 84 86 89
[47] 90 91 91 90 89 90 91 91 91 84 88 84 86 88 84 82 80 73 87 84 87 89 89
[70] 89 91 84 86 88 78 79 86 82 82 78 79 79 78 81 84 84 87 84 79 75 72 64
[93] 66 72 84 70 66 64 60 78 70 72 69 69 73 79 81 80 82 66 63 68 79 81 69
[116] 73 73 75 75 81 82 82 81 86 90 93 91 84 84 75 87 84 87 84 88 86 90 91
[139] 91 89 89 89 90 89 84 87 88 89 89 91 91 89 88 72 80 84 88 89 88 84 84
[162] 80 73 80 86 88 88 87 88 91 91 89 89 88 82 79 81 82 84 87 90 90 91 91
[185] 88 88 91 93 81 81 82 86 88 84 80 82 86 87 87 88 88 90 88 91 95 89 70
[208] 80 82 66 70 64 68 77 86 75 73 75 78 81 82 82 82 80 82 82 79 80 68 63
[231] 57 66 64 69 70 70 62 63 62 75 71 57 55 64 66 60 91 88 91 91 91 89 93
[254] 95 95 91 91 86 88 87 91 87 90 91 95 91 91 89 91 91 86 88 80 88 89 90
[277] 86 86 82 84 86 90 89 89 86 82 87 88 84 86 80 82 86 84 87 90 79 84 87
[300] 87 88 90 91 89 90 93 93 91 87 84 77 90 91 89 90 89 79 78 81 84 89 87
[323] 87 88 87 82 80 82 82 88 84 81 82 84 87 80 75 75 86 78 77 82 82 73 82
[346] 69 72 73 78 78 78 75 79 78 77 78 82 75 73 63 63 72 75 79 79 79 78 82
[369] 79 84 82 87 88 90 91 82 86 87 87 82 77 73 81 81 86 82 87 88 90 90 91
[392] 93 93 91 93 93 93 93 97 99 96 93 88 89 91 93 93 93 91 90 96 98 97 98
[415] 93 93 96 98 98 89 91 91 90 80 82 89 88 90 91 91 84 88 91 84 93 96 96
[438] 91 91 77 87 87 87 86 87 89 81 81 82 79 68 79 72 75 78 81 82 78 80 77
[461] 71 73 75 84 71 73 71 73 73 72 72 73 70 64 75 73 77 80 71 66 60 64 73
[484] 57 59 64 69 75 73 72 75 75 89 91 93 95 96 96 96 91 96 99 96 93 91 93
[507] 93 93 91 97 100 99 93 96 87 82 75 82 88 91 89 87 86 86 81 84 88 91 91
[530] 91 91 96 95 89 89 89 89 94 97 99 101 101 97 87 86 88 92 92 90 90 92 92
[553] 88 87 79 81 82 87 81 66 66 75 80 82 84 86 87 86 80 75 73 73 84 87 77
[576] 73 81 84 82 68 71 75 73 75 77 79 82 81 82 73 66 55 55 64 71 73 75 75
[599] 77 80 80 80 73 73 75 79 75 75 78 75 78 80 75 77 78 84 87 87 84 86 87
[622] 87 89 91 87 90 90 86 82 82 84 87 88 90 87 84 87 90 84 82 88 90 84 89
[645] 89 87 84 84 84 86 88 84 86 88 87 88 86 86 81 87 84 90 91 91 87 86 88
[668] 90 88 93 90 91 91 81 86 81 82 80 75 73 81 90 88 87 86 86 89 87 84 84
[691] 86 77 77 81 81 82 84 86 87 88 69 66 72 75 78 71 71 75 80 81 80 79 70
[714] 68 79 66 73 75 78 78 75 75 62 60 64 71 75 79 80 81 79 73 64 51 55 63
[737] 72 71 90 90 87 89 93 93 89 89 90 91 84 77 82 88 91 93 93 93 93 91 95
[760] 91 89 87 84 86 89 91 91 88 90 93 91 91 91 93 97 87 87 86 88 89 91 91
[783] 89 88 90 91 93 91 93 93 91 95 93 91 88 84 82 82 78 77 84 84 89 95 93
[806] 91 88 87 91 95 95 90 75 78 91 88 86 81 80 86 84 77 82 73 69 75 75 79
[829] 73 79 82 84 84 82 87 86 80 71 66 70 78 84 79 68 57 66 64 68 71 73 71
[852] 64 59 68 60 68 69 75 75 68 60 73 81 87 86 80 84 87 90 89 84 84 86 87
[875] 84 86 88 88 88 88 88 89 86 81 82 84 87 87 89 88 84 88 84 84 84 82 84
[898] 82 84 84 86 87 84 81 87 89 90 86 89 90 90 87 88 88 90 89 88 89 90 91
[921] 89 88 89 88 86 87 87 84 73 75 81 82 79 80 81 84 82 82 81 81 81 84 87
[944] 82 75 81 80 82 82 82 73 66 71 72 68 66 77 78 75 73 73 73 73 66 78 78
[967] 78 69 72 68 70 75 78 84 78 78 73 73 68 64 57 70 77 75 82 81 86 88 90
[990] 90 89 87 88 89 90 89 91 91 84 84
```

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```
>
> class(tmp1_ts)
[1] "ts"
> plot(tmp1_ts)
>
> # Define beta to generate trend value to forecast a future trend
> tmp1_b_ts <- HoltWinters(tmp1_ts , beta=.5)
> plot(tmp1_b_ts)
>
> # Exponential Smoothing
> #Simple Exponential #
> tmp1_single <- HoltWinters(tmp1_ts,beta=FALSE, gamma=FALSE)
>
> #Double Exponential - model trend #
> tmp1_double <- HoltWinters(tmp1_ts,gamma=FALSE)
>
> #Triple Exponential - model trend and seasonality#
> tmp1_triple <- HoltWinters(tmp1_ts, seasonal = "additive")
>
> #Compare 3 kinds of Exp Smoothing#
> tmp1_single
Holt-Winters exponential smoothing without trend and without seasonal component.
```

Call:

```
HoltWinters(x = tmp1_ts, beta = FALSE, gamma = FALSE)
```

Smoothing parameters:

alpha: 0.8396301

beta : FALSE

gamma: FALSE

Coefficients:

[,1]

a 81.62444

```
> tmp1_single$SSE # Sum of Squared Error
```

[1] 53704.15

```
>
```

```
> tmp1_double
```

Holt-Winters exponential smoothing with trend and without seasonal component.

Call:

```
HoltWinters(x = tmp1_ts, gamma = FALSE)
```

Smoothing parameters:

alpha: 0.8455303

beta : 0.003777803

gamma: FALSE

Coefficients:

[,1]

a 81.729657393

b -0.004838906

> tmp1\_double\$SSE

[1] 54071.22

>

> tmp1\_triple

Holt-Winters exponential smoothing with trend and additive seasonal component.

Call:

HoltWinters(x = tmp1\_ts, seasonal = "additive")

Smoothing parameters:

alpha: 0.6677614

beta : 0

gamma: 0.6297674

Coefficients:

[,1]

a 66.739214602

b -0.004362918

s1 17.167113056

s2 12.692593452

s3 11.926233267

s4 12.862822489

s5 11.026083880

s6 8.860499089

s7 9.547553333

s8 7.755384526

s9 4.419013466

s10 2.272689626

s11 4.628251667

s12 2.396834852

s13 3.512957136

s14 1.734948091

s15 3.035023890

s16 6.257944053

s17 5.086362292

s18 8.599153274

s19 5.507486014

s20 10.404819396

s21 10.115801978

s22 9.628840064

s23 7.658623118

s24 7.150473636

s25 6.306599371  
s26 5.850691115  
s27 5.770487458  
s28 4.280481134  
s29 7.229771199  
s30 4.632381095  
s31 6.006248308  
s32 6.443645890  
s33 5.701166527  
s34 3.546887269  
s35 3.879569716  
s36 3.517339384  
s37 2.828550977  
s38 2.122971410  
s39 2.627923984  
s40 1.658896597  
s41 0.165866282  
s42 -0.001574460  
s43 -1.557500303  
s44 -2.159601227  
s45 -2.260609558  
s46 0.474052766  
s47 2.501631056  
s48 6.552191593  
s49 7.240238719  
s50 8.395899120  
s51 8.633263084  
s52 7.504540260  
s53 4.804135812  
s54 0.449902809  
s55 -1.045831475  
s56 1.562077049  
s57 1.632745190  
s58 0.857309158  
s59 2.909614779  
s60 0.626594899  
s61 4.491805650  
s62 4.567058619  
s63 3.065433531  
s64 3.787652805  
s65 -2.147135463  
s66 1.759895146  
s67 1.541155061  
s68 1.278521842  
s69 0.895959617  
s70 2.009912430  
s71 3.695537344

s72 4.675235988  
s73 4.535880359  
s74 1.710420810  
s75 0.822675780  
s76 2.363162195  
s77 1.925012161  
s78 -1.656914701  
s79 -1.809929506  
s80 -0.427021203  
s81 0.056812125  
s82 -1.137248149  
s83 -1.037423821  
s84 -2.817503990  
s85 -4.578240308  
s86 -3.080091372  
s87 -2.710719111  
s88 -2.255335538  
s89 -4.518502545  
s90 -5.159556421  
s91 -4.440834373  
s92 -5.790113744  
s93 -7.461163074  
s94 -8.882612687  
s95 -8.619859733  
s96 -6.200719796  
s97 -6.055889182  
s98 -11.167287691  
s99 -13.489975101  
s100 -13.615536188  
s101 -14.373453486  
s102 -15.142110213  
s103 -14.419874185  
s104 -14.023613348  
s105 -16.187082843  
s106 -15.999259045  
s107 -12.074075053  
s108 -9.199729415  
s109 -10.403127076  
s110 -12.075113349  
s111 -9.722863134  
s112 -5.846856763  
s113 -8.047801338  
s114 -9.636669876  
s115 -10.510269852  
s116 -12.876648138  
s117 -8.657362442  
s118 -9.828539578

```

s119 -14.522204766
s120 -11.852457644
s121 -8.714763993
s122 -4.711332904
s123 18.737998957
> tmp1_triple$SSE
[1] 63025.97
>
> # Triple Exponential using multiplicative ( Seasonal variation is connected)
> tmp1_triple_m <- HoltWinters(tmp1_ts, seasonal = "multiplicative")
> tmp1_triple_m$SSE
[1] 65648.65
>
>
> # Triple Exponential using multiplicative ( Seasonal variation is independent of the level)
> tmp1_triple_a <- HoltWinters(tmp1_ts, seasonal = "additive")
> tmp1_triple_a$SSE
[1] 63025.97
>
> tmp1_triple_a$fitted
Time Series:
Start = c(1997, 1)
End = c(2015, 1)
Frequency = 123
      xhat  level  trend  season
1997.000 87.17619 82.87739 -0.004362918 4.303159495
1997.008 90.32137 82.08762 -0.004362918 8.238118845
1997.016 92.95607 81.86865 -0.004362918 11.091777381
1997.024 90.93226 81.89363 -0.004362918 9.042996893
1997.033 83.99752 81.93450 -0.004362918 2.067387137
1997.041 84.04359 81.93179 -0.004362918 2.116167625
1997.049 75.06703 81.89832 -0.004362918 -6.826921806
1997.057 87.04230 81.84919 -0.004362918 5.197468438
1997.065 84.01782 81.81658 -0.004362918 2.205598519
1997.073 87.05847 81.80032 -0.004362918 5.262509089
1997.081 84.04758 81.75692 -0.004362918 2.295029414
1997.089 88.04397 81.72078 -0.004362918 6.327549739
1997.098 86.02650 81.68706 -0.004362918 4.343809902
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> plot(fitted(tmp1_triple_a))
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> plot(fitted(tmp1_triple_m))
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> tmp1_forecast=predict( tmp1_b_ts, n.ahead = 100, prediction.interval = TRUE )  
> plot(forecast( tmp1_b_ts, h = 180 ))
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