

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

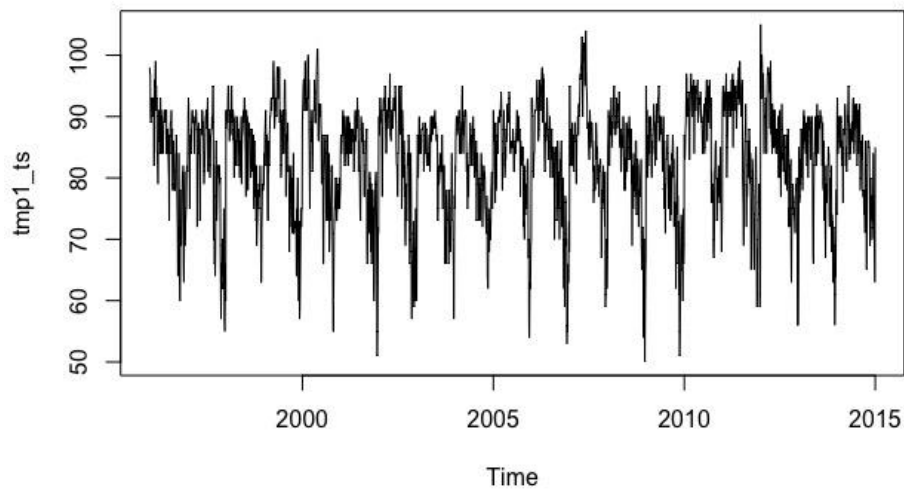
Triple ES Values:

alpha: 0.6677614

beta : 0

gamma: 0.6297674

SSE: 63025.97



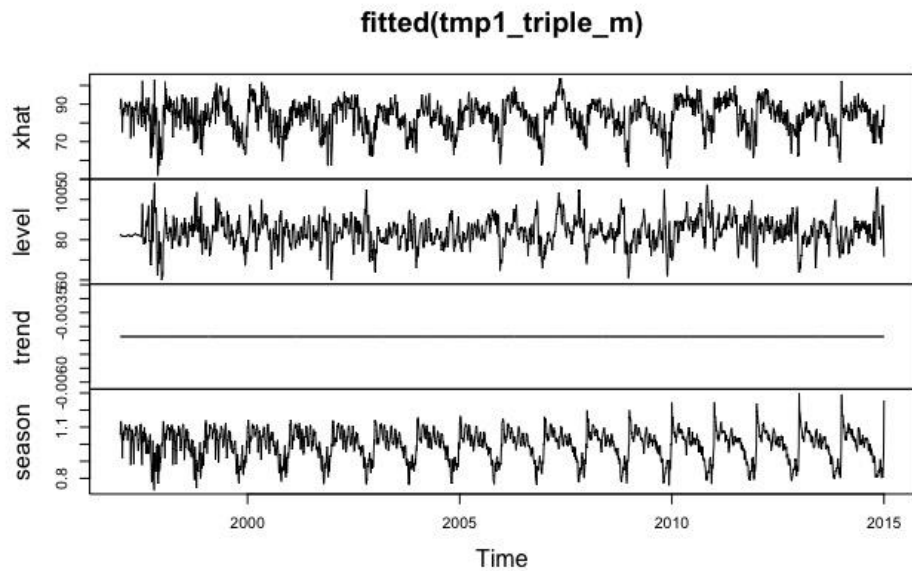
#Single ES has smallest SSE when compared to other two and it is closed to 1. It means there is less randomness in the system and recent temperature reading has more weight in predicting the current temperature.

#Seasonality uses Model AAM:

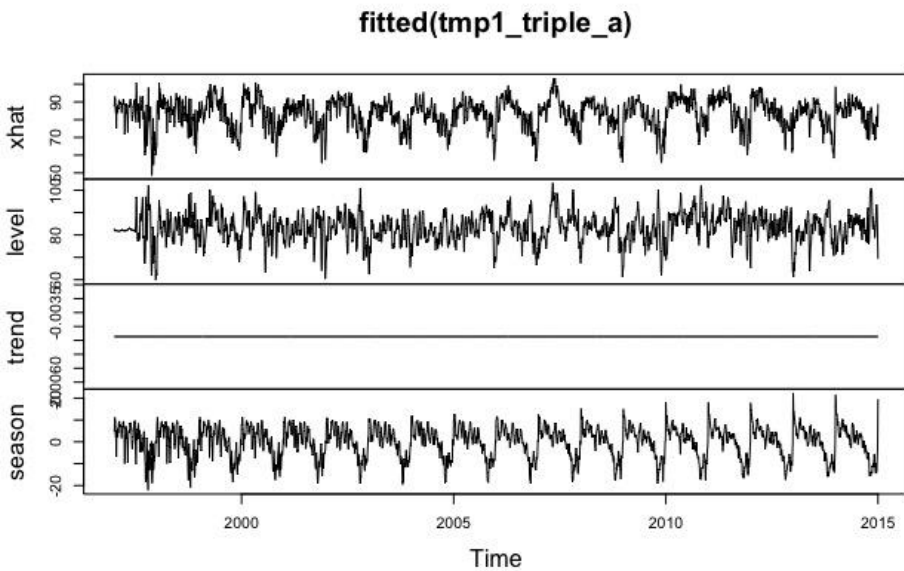
- #1. A - additive: seasonal variation is independent
- #2. M - multiplicative: seasonal variation is connected.

#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 subchart shows that the duration of each season has been constant throughout all 20 years.

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 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
```

```
[ reached getOption("max.print") -- omitted 1338 entries ]
```

```
>
```

```
> 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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