

Homework01

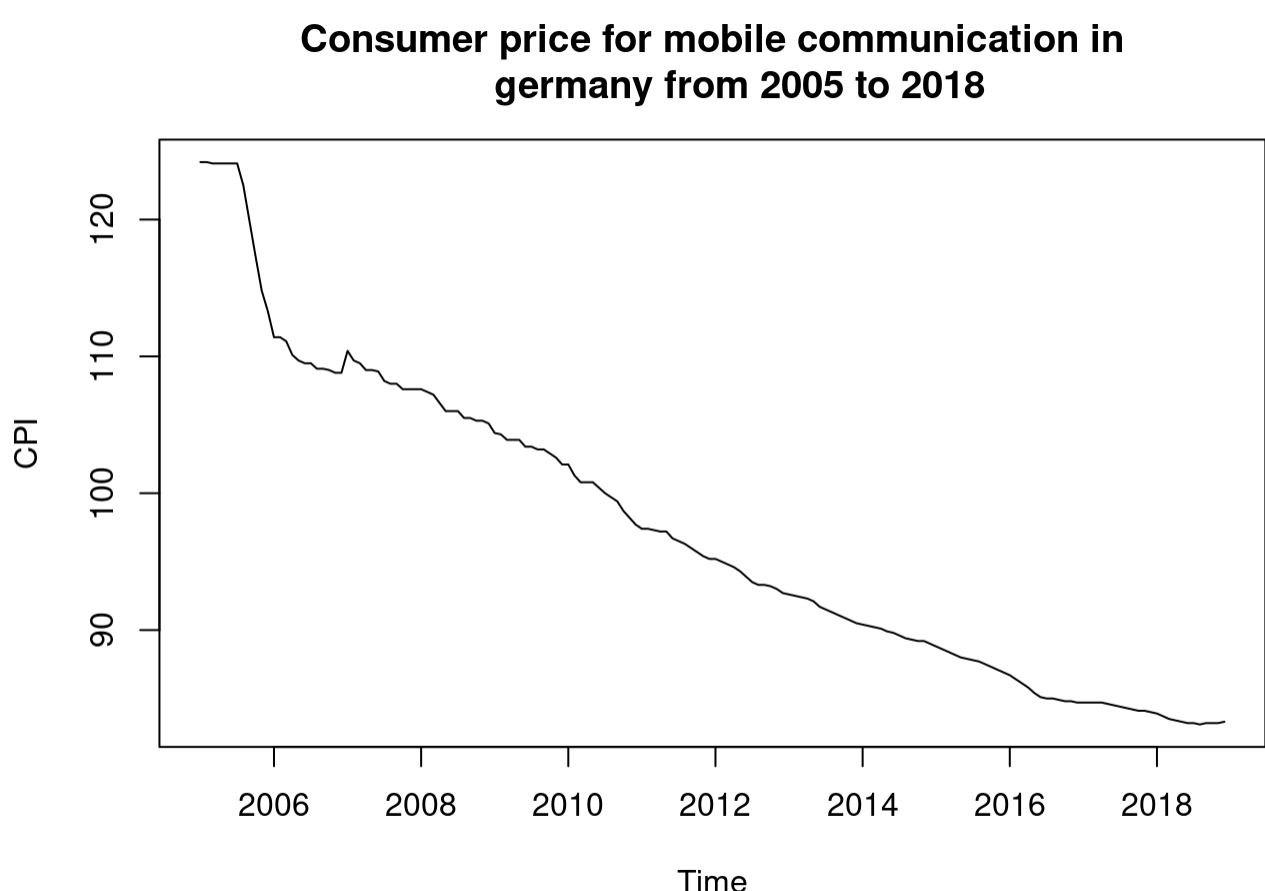
2022-05-06

R Markdown

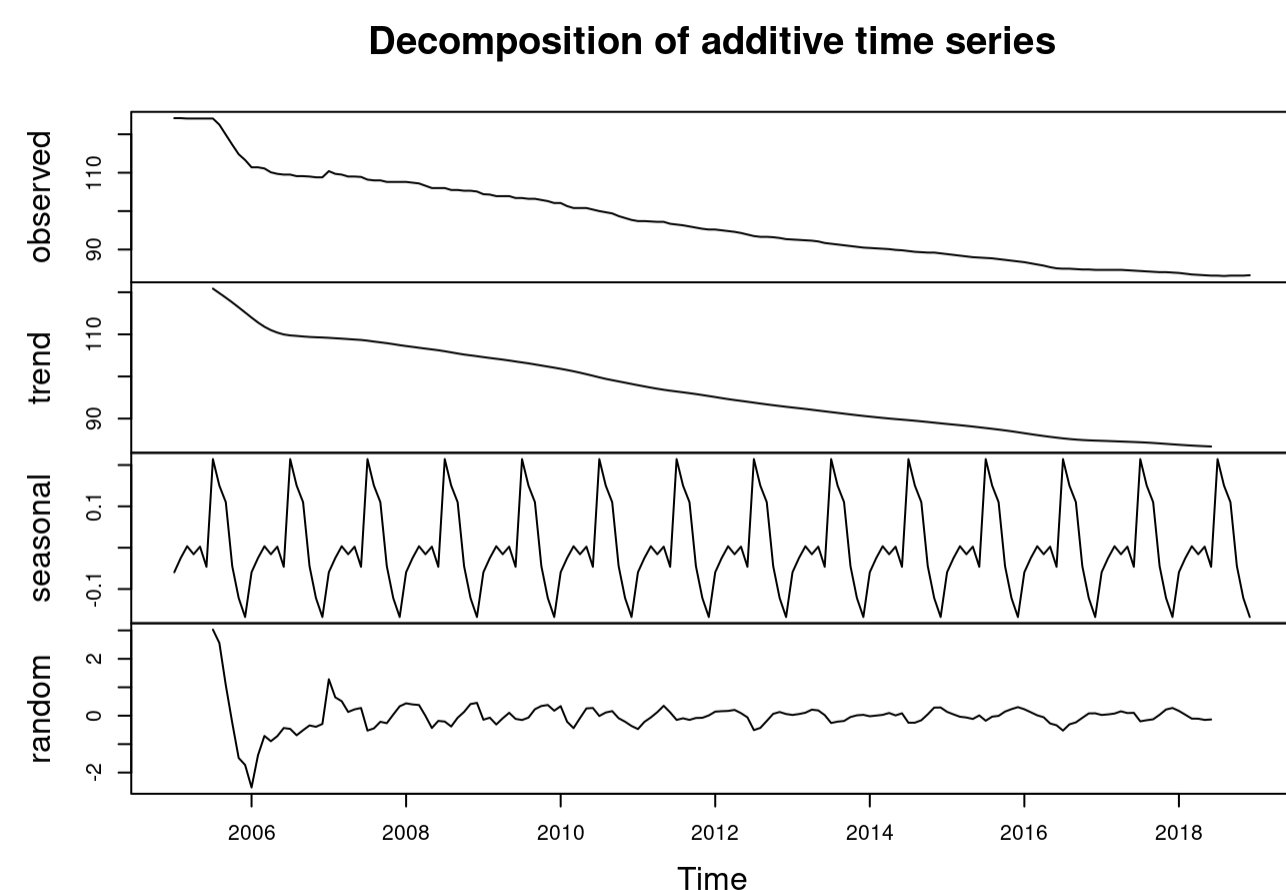
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When you click the Knit button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

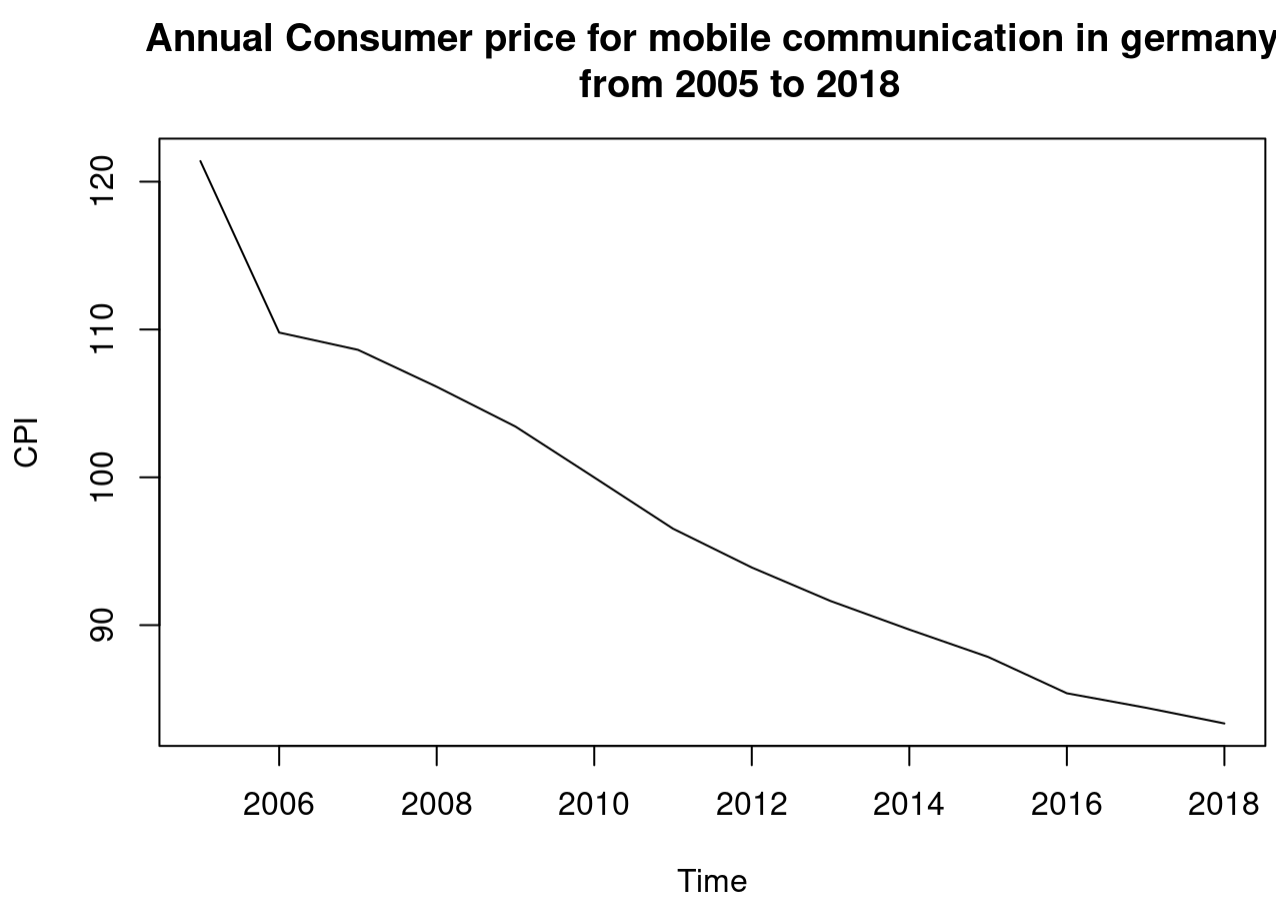
```
library(xlsx)
#(a) read the data and plot the time series
mobile.data <- read.xlsx('statistic-mobile-communications.xlsx',
  sheetName = 'Data', header=T)
attach(mobile.data)
ts.mobile.data <- ts(mobile.data$CPI, start=c(2005,1), end=c(2018,12), frequency = 12)
plot(ts.mobile.data, main="Consumer price for mobile communication in
  germany from 2005 to 2018", ylab = "CPI")
```



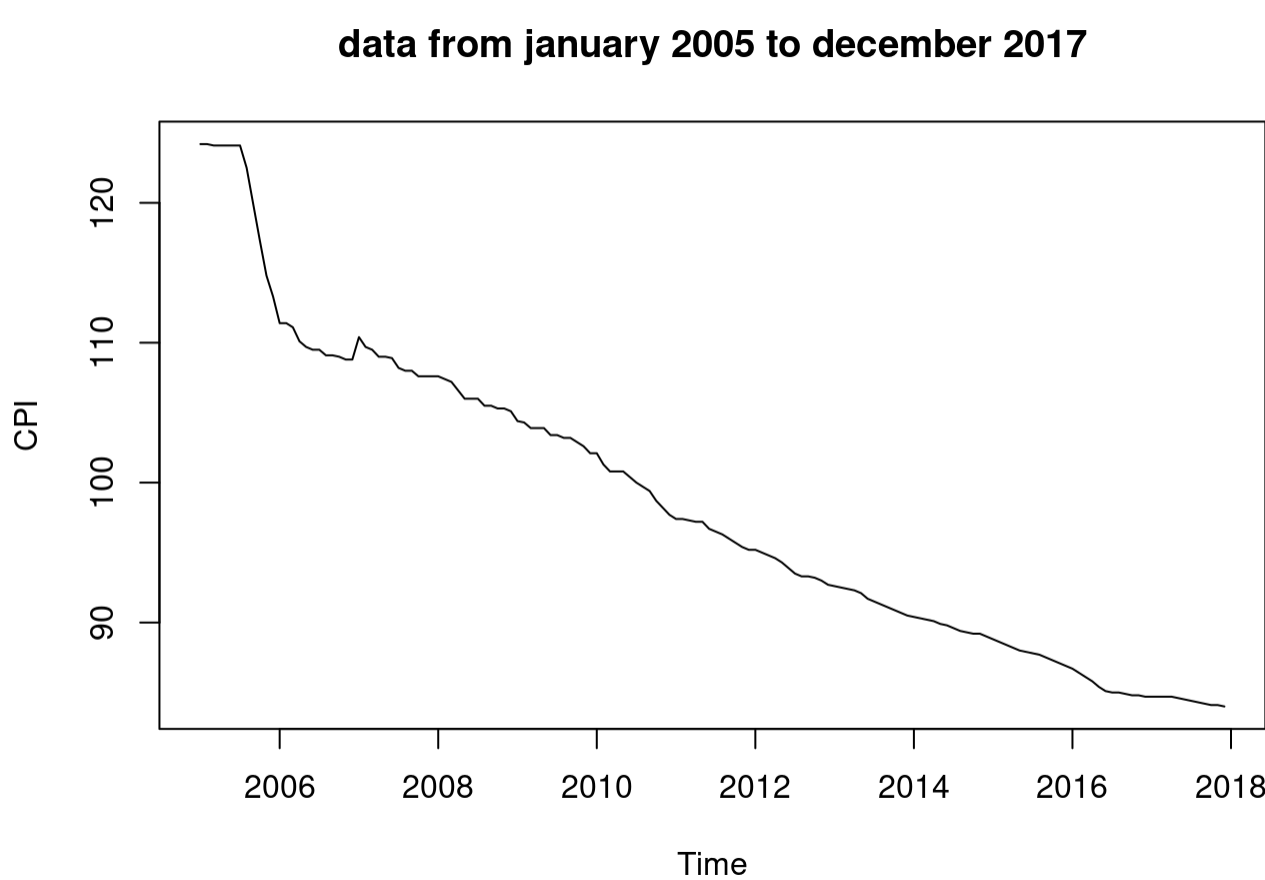
```
#(b) decompose the time series into three parts: estimating trends, seasonal
#effects, and random series.
ts.mobile.data.decompose <- decompose(ts.mobile.data)
plot(ts.mobile.data.decompose)
```



```
#(c) Use the aggregate function to remove any seasonal effects within each
#year and produce an annual series of mean CPI for the period 2005-2018
ts.mobile.annual <- aggregate(ts.mobile.data) / 12
plot(ts.mobile.annual, main="Annual Consumer price for mobile communication in germany
  from 2005 to 2018", ylab="CPI")
```



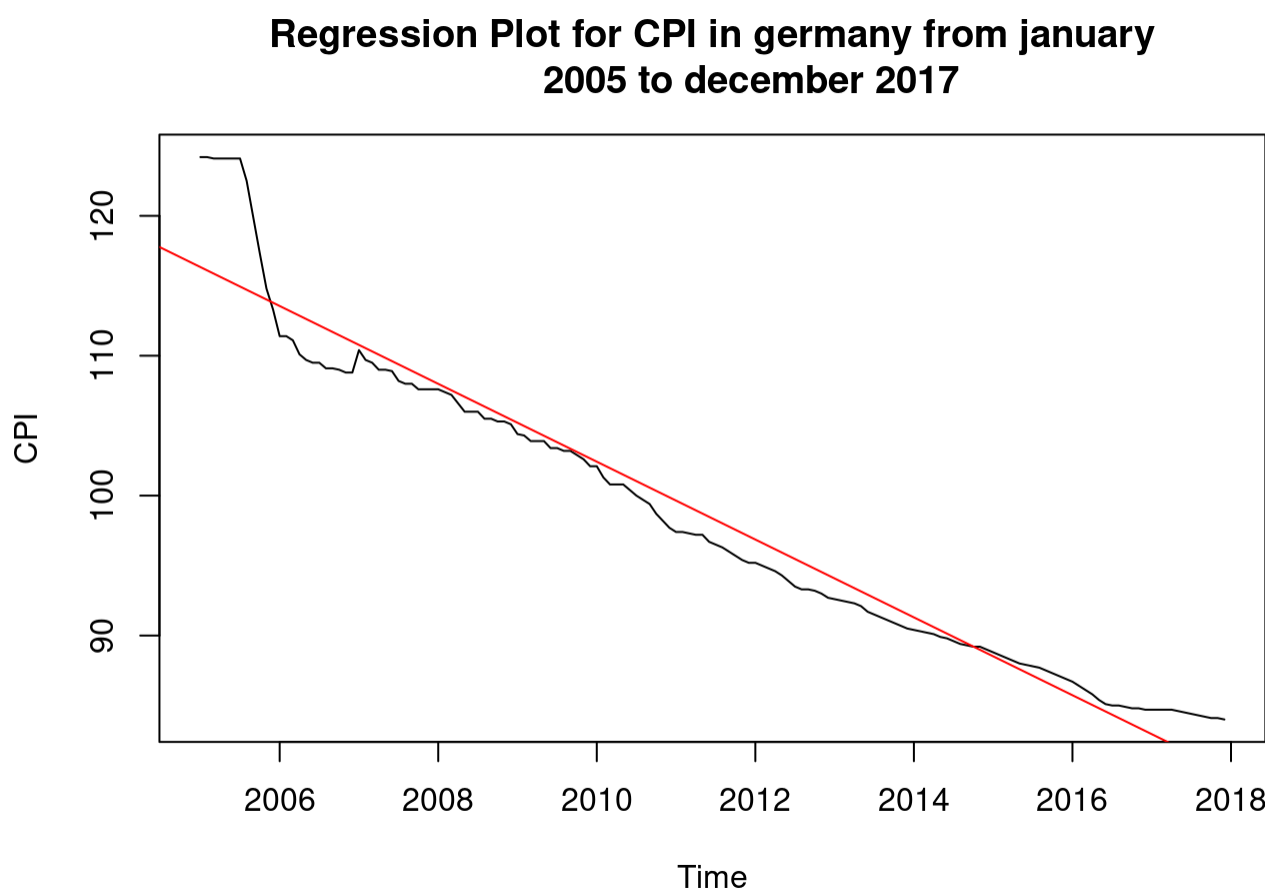
```
#(d) Use the window function to plot the data from January 2005 to December 2017
help("window")
ts.mobile.window <- window(ts.mobile.data, start=c(2005,1), end=c(2017,12))
plot(ts.mobile.window, main="data from january 2005 to december 2017", ylab="CPI")
```



```
#(e) Use the command lm to estimate the parameters 'a' and 'b' in the simple
#linear regression model.
#(f) Use commands summary and abline to add lines to existing plots in
#step (e)
linearRegressionModel <- function(data) {
  t <- time(data)
  reg <- lm(data~t)
  print("====coefficients for Linear Regression Model:")
  print(reg)
  print("====Summary of Linear Regression Model:")
  print(summary(reg))

  plot(data, ylab="CPI", main="Regression Plot for CPI in germany from january
    2005 to december 2017")
  abline(reg, col="red")
}
#apply regression model
linearRegressionModel(ts.mobile.window)
```

```
## [1] "====coefficients for Linear Regression Model:"
##
## Call:
## lm(formula = data ~ t)
##
## Coefficients:
## (Intercept)          t
## 5694.604         -2.782
##
## [1] "====Summary of Linear Regression Model:"
##
## Call:
## lm(formula = data ~ t)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.9260 -1.4874 -0.6613  0.6117  9.1555
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 5694.60420    105.18543   54.18  <2e-16 ***
## t           -2.78218     0.05225   -53.24  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.449 on 154 degrees of freedom
## Multiple R-squared:  0.9485, Adjusted R-squared:  0.9481
## F-statistic: 2835 on 1 and 154 DF, p-value: < 2.2e-16
```



```
#####
#PREDICTION BY HAND (g)
#####

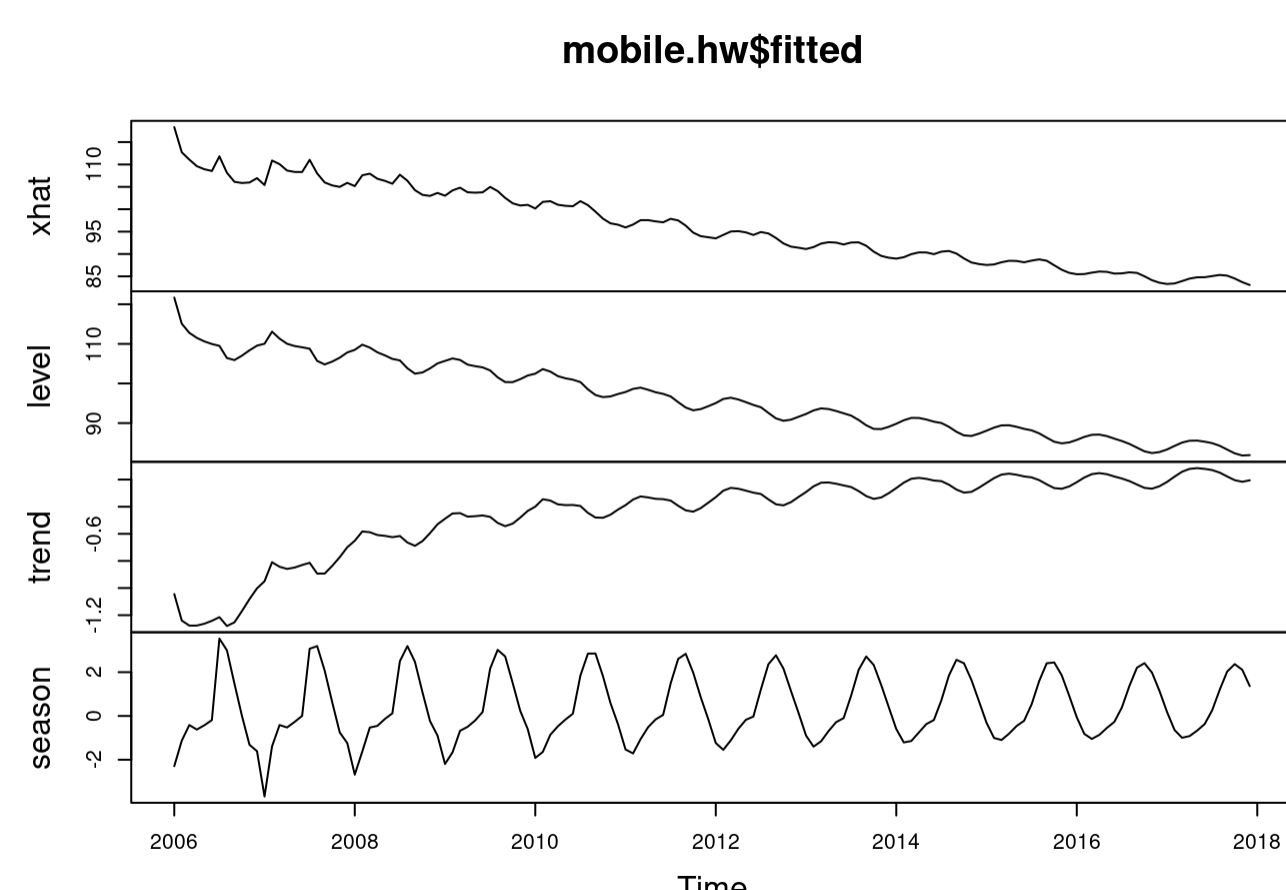
#####JUNE 2018#####
#<_hat(2018) = a + bt = 5694.604 -2.782 * (2018+5/12) = 34,369

#####JUNE 2019#####
#y_hat(2019) = a + bt = 5694.604 -2.782 * (2019+5/12) = 31,587

#####JUNE 2020#####
#y_hat(2020) = a + bt = 5694.604 -2.782 * (2020+5/12) = 28,805

#(h)
```

```
#Holt-Winters
mobile.hw <- HoltWinters(ts.mobile.window)
plot(mobile.hw$fitted)
```



```
summary(mobile.hw)
```

```
##          Length Class  Mode
## fitted      576 mts    numeric
## x           156 ts     numeric
## alpha        1 -none-  numeric
## beta         1 -none-  numeric
## gamma        1 -none-  numeric
## coefficients 14 -none-  numeric
## seasonal     1 -none-  character
## SSE          1 -none-  numeric
## call         2 -none-  call
```

```
#(i) make predictions using holt-winters
predict(mobile.hw, n.ahead = 36)
```

```
##          Jan      Feb      Mar      Apr      May      Jun      Jul      Aug
## 2018 82.73340 81.78623 81.06730 80.85304 80.84591 80.93922 81.32052 82.00581
## 2019 80.58376 79.55659 78.91766 78.70339 78.69626 78.78958 79.17088 79.85617
## 2020 78.43411 77.48695 76.76801 76.55375 76.54662 76.63994 77.02124 77.70653
##          Sep      Oct      Nov      Dec
## 2018 82.67596 82.94044 82.66783 81.85036
## 2019 80.52631 80.79079 80.51819 79.70071
## 2020 78.37667 78.64115 78.36854 77.55107
```