

Homework6

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Exercise 1

part a

Mean Function A time series defined as an observation of a stochastic process resulting in a set of variables x_1, x_2, \dots, x_n is defined by a joint distribution function $F(c_1, c_2, \dots, c_n) = P(x_{i1} \leq c_1, x_{i2} \leq c_2, \dots, x_{in} \leq c_n)$

Assuming knowledge of such a joint probability distribution, we would derive the marginal probability distributions $f_t(x_t)$

And from such marginal probability distributions, we define the mean function:

$$\mu_x(t) = E(x_t) = \int_{-\infty}^{+\infty} f_t(x_t) dx_t$$

This mean function is different from the mean function of observations of a single random variable, as seen with the classical linear model.

For time series, the observation of x_t is dependent on previous observations of x_{t-1}, x_{t-2}, \dots . That dependency is captured in the joint probability distribution which is unavailable to us, as the time series represents the single instance of the realization of the stochastic process that we are able to observe.

Variance Function For time series defined as described in the mean function discussion above, the variance function, a function of time t , is defined as:

$$\sigma_x(t) = E(x_t - \mu_x(t))^2 = \int_{-\infty}^{+\infty} (x_t - \mu_x)^2 f_t(x_t) dx_t$$

Where $f_t(x_t)$ is the marginal probability distribution of x_t in the stochastic process.

This variance function is also different from the variance of the observations of a single random variable studied with classical linear models, because of the dependency of x_t over x_{t-1}, x_{t-2}, \dots as expressed in the joint probability distribution.

part b

The assumption of strict stationarity is very strong strong assumption of stationarity.

For a given time series, we say that it is **strictly stationary** if its distribution is unchanged for any time shift. i.e. given a joint distribution $F(x_{t1}, x_{t2}, \dots, x_{tn})$ as introduced earlier, a time series $x_{\{t\}}$ is strictly stationary if $F(x_{t1}, x_{t2}, \dots, x_{tn}) = F(x_{t1+m}, x_{t2+m}, \dots, x_{tn+m}), \forall t_1, \dots, t_n$ and m

The assumption of **weak stationarity** (or second order stationarity) is a weaker assumption of stationarity. A time series x_t is weak stationary if its mean and variance stationary and its auto-covariance $Cov(x_t, x_{t+k})$ depends only on the lag k , and is not a function of time t .

The auto-covariance of a time series that is only dependent of lag k is defined as:

$$\gamma_k = E[(x_t - \mu)(x_{t+k} - \mu)]$$

where μ is the stationary mean of the time series.

Exercise 2

Part a

```
rw.wod <- white.noise <- rnorm(500)
for (t in 2:length(rw.wod)) {
  rw.wod[t] <- rw.wod[t - 1] + white.noise[t]
}
```

Part b

Mean of time series The mean of the time series is: 9.838132 The standard deviation of the time series is: 4.723985 The 25th, 50th and 75th quantiles of the time series are: 6.187022 9.525720 13.694139

```
mean(rw.wod)
```

```
## [1] 9.838132
```

```
sd(rw.wod)
```

```
## [1] 4.723985
```

```
quantile(rw.wod)
```

```
##          0%          25%          50%          75%          100%
## -1.278439  6.187022  9.525720 13.694139 20.061148
```

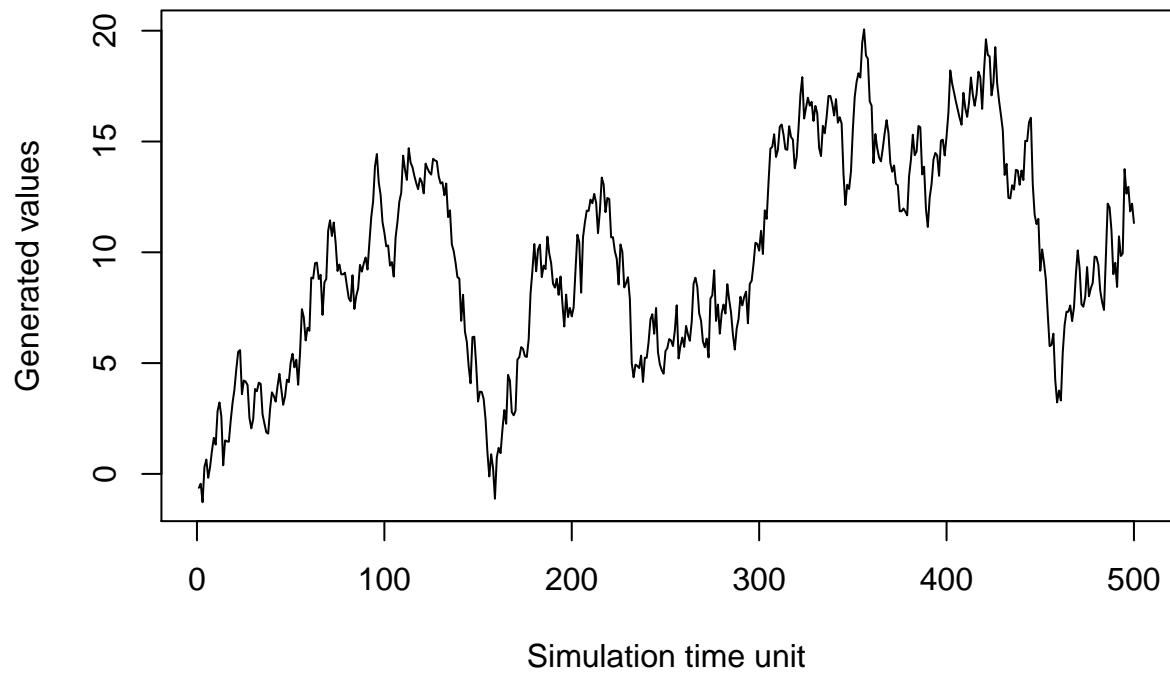
```
describe(rw.wod)
```

```
##   vars   n mean   sd median trimmed  mad   min   max range  skew kurtosis
## 1    1 500 9.84 4.72   9.53     9.9 5.57 -1.28 20.06 21.34 -0.08   -0.83
##      se
## 1 0.21
```

Part c

```
plot.ts(rw.wod, xlab = "Simulation time unit", ylab = "Generated values",
  main = "Random Walk Without Drift Time Series ")
```

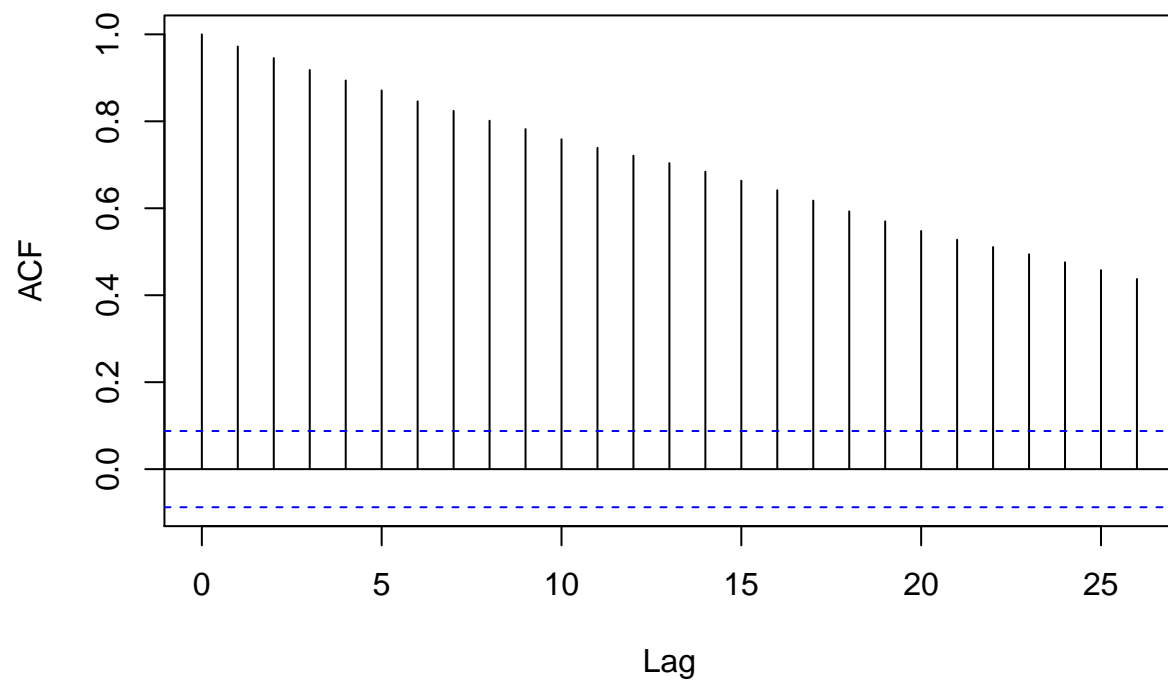
Random Walk Without Drift Time Series



Part d

```
acf(ts(rw.wod), main = "Random Walk Without Drift Time Series")
```

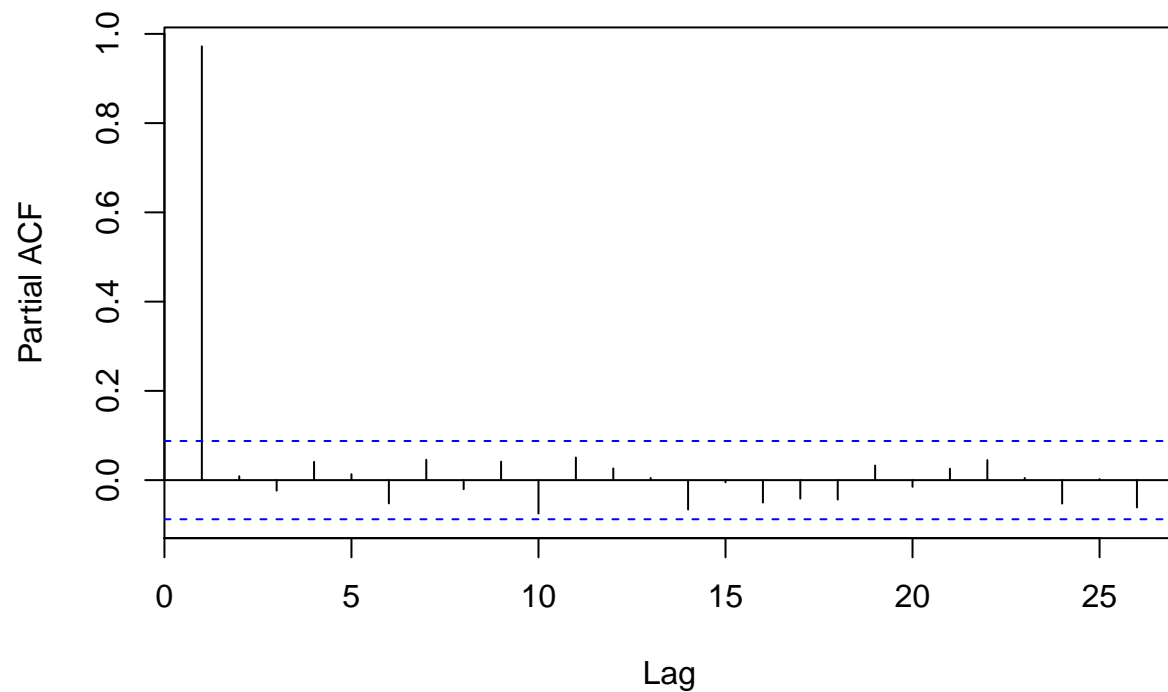
Random Walk Without Drift Time Series



Part e

```
pacf(ts(rw.wod), main = "Random Walk Without Drift Time Series")
```

Random Walk Without Drift Time Series



Exercise 3

Part a

```
rw.wid <- white.noise
for (t in 2:length(rw.wid)) {
  rw.wid[t] <- rw.wid[t - 1] + 0.5 + white.noise[t]
}
```

Mean of time series The mean of the time series is: 134.5881 The standard deviation of the time series is: 74.88504 The 25th, 50th and 75th quantiles of the time series are: 76.3171217 130.3604551 199.4784637

```
mean(rw.wid)
```

```
## [1] 134.5881
```

```
sd(rw.wid)
```

```
## [1] 74.88504
```

```
quantile(rw.wid)
```

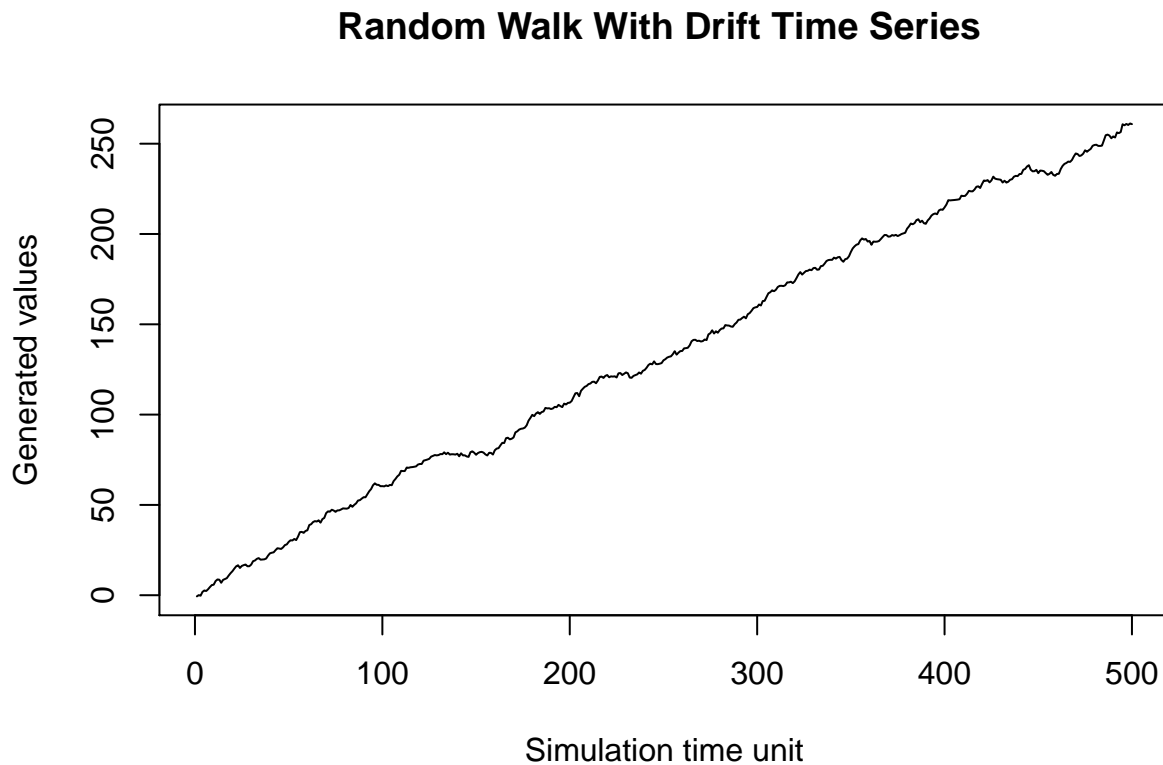
```
##           0%          25%          50%          75%         100%  
## -0.6264538  76.3171217 130.3604551 199.4784637 261.1953089
```

```
describe(rw.wid)
```

```
##   vars   n   mean    sd median trimmed   mad   min   max range skew  
## 1     1 500 134.59  74.89 130.36 135.58  97.01 -0.63 261.2 261.82 -0.03  
##   kurtosis   se  
## 1     -1.23  3.35
```

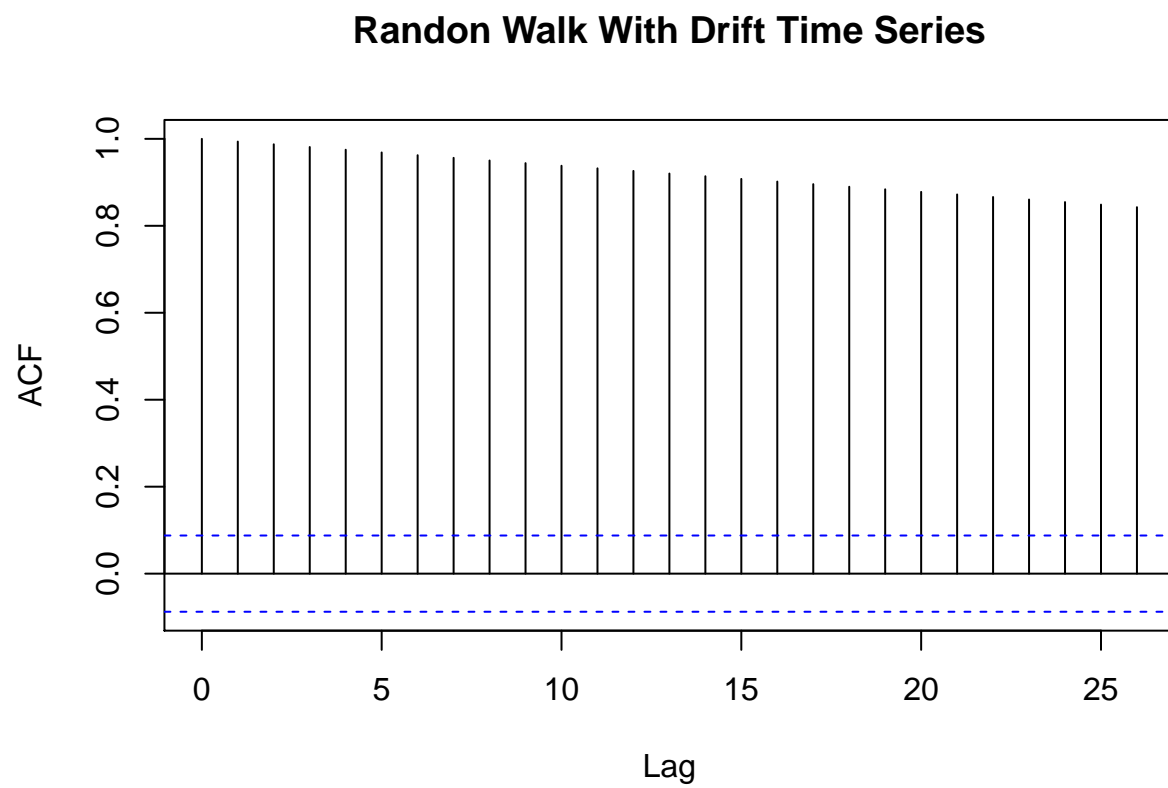
Part c

```
plot.ts(rw.wid, xlab = "Simulation time unit", ylab = "Generated values",  
        main = "Random Walk With Drift Time Series ")
```



Part d

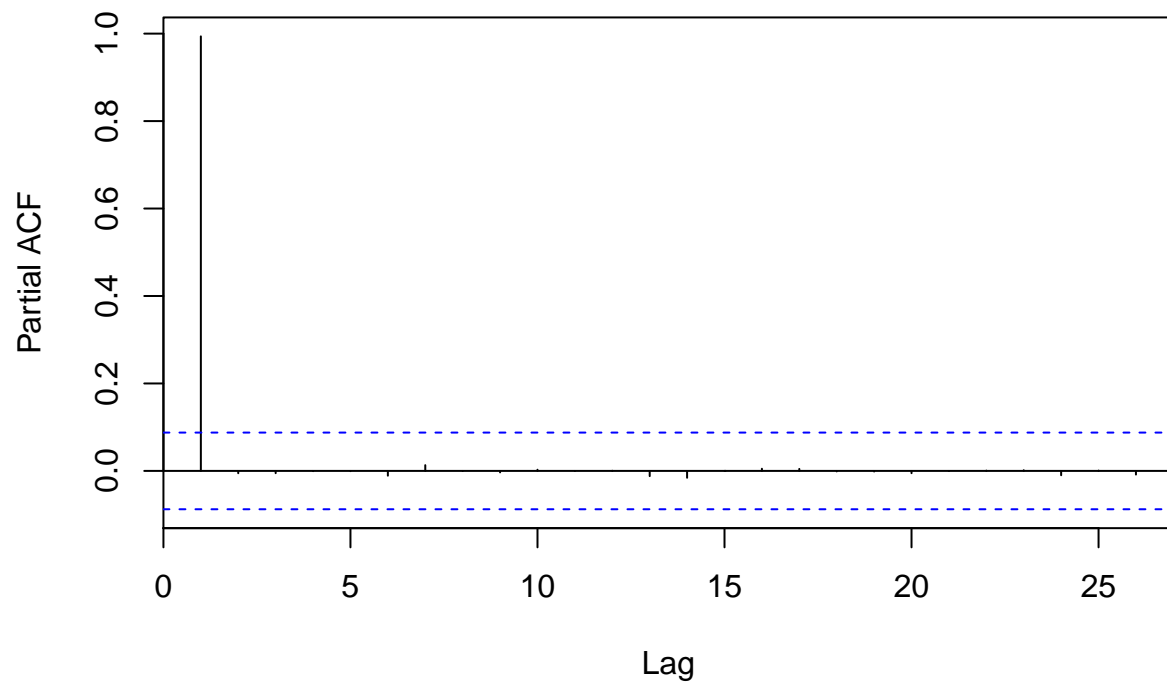
```
acf(ts(rw.wid), main = "Random Walk With Drift Time Series")
```



Part e

```
pacf(ts(rw.wid), main = "Random Walk With Drift Time Series")
```

Random Walk With Drift Time Series



Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.

Exercise 4

Part a

```
data <- read.csv("INJCJC.csv")
```

```
str(data)
```

```
## 'data.frame': 1300 obs. of 3 variables:
## $ Date : Factor w/ 1300 levels "1-Apr-05","1-Apr-11",...: 1102 143 442 784 483 1271 312 654 498 12
## $ INJCJC : int 355 369 375 345 368 367 348 350 351 349 ...
## $ INJCJC4: num 362 366 364 361 364 ...
```

```
dim(data)
```

```
## [1] 1300 3
```

```
head(data)
```



```
##           Date INJCJC INJCJC4
## 1  5-Jan-90    355  362.25
## 2 12-Jan-90    369  365.75
## 3 19-Jan-90    375  364.25
## 4 26-Jan-90    345  361.00
## 5  2-Feb-90    368  364.25
## 6  9-Feb-90    367  363.75
```

```
tail(data)
```

```
##           Date INJCJC INJCJC4
## 1295 24-Oct-14    288  281.25
## 1296 31-Oct-14    278  279.00
## 1297  7-Nov-14    293  285.75
## 1298 14-Nov-14    292  294.25
## 1299 21-Nov-14    314  294.25
## 1300 28-Nov-14    297  299.00
```

Part b

```
data.ts <- ts(data$INJCJC, frequency = 52, start = c(1990, 1), end = c(2014,
52))
summary(data.ts)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 259.0   324.0   353.5   371.1   406.0   665.0
```

```
quantile(data.ts)
```

```
##      0%   25%   50%   75%  100%
## 259.0 324.0 353.5 406.0 665.0
```

Part c

```
INJCJC.time <- time(data.ts)
```

Part d

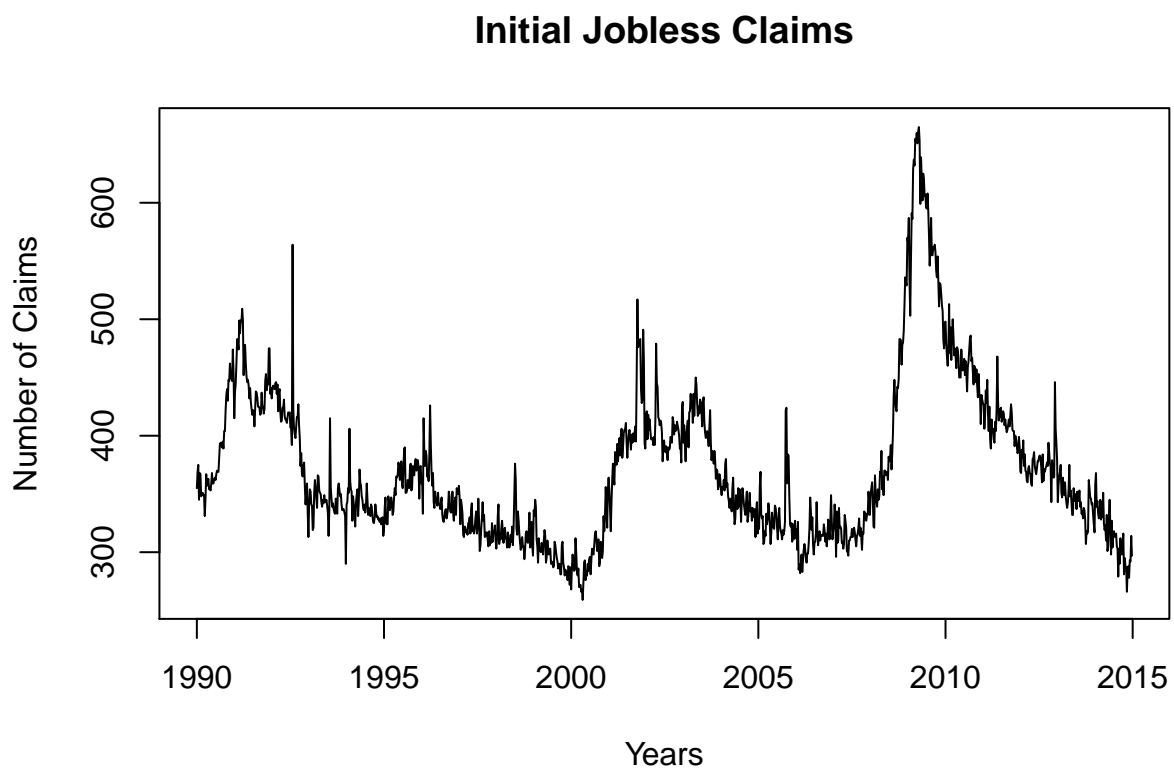
```
head(cbind(INJCJC.time, data.ts), 10)
```

```
##      INJCJC.time data.ts
## [1,]   1990.000    355
## [2,]   1990.019    369
## [3,]   1990.038    375
## [4,]   1990.058    345
```

```
## [5,] 1990.077 368
## [6,] 1990.096 367
## [7,] 1990.115 348
## [8,] 1990.135 350
## [9,] 1990.154 351
## [10,] 1990.173 349
```

Part e1

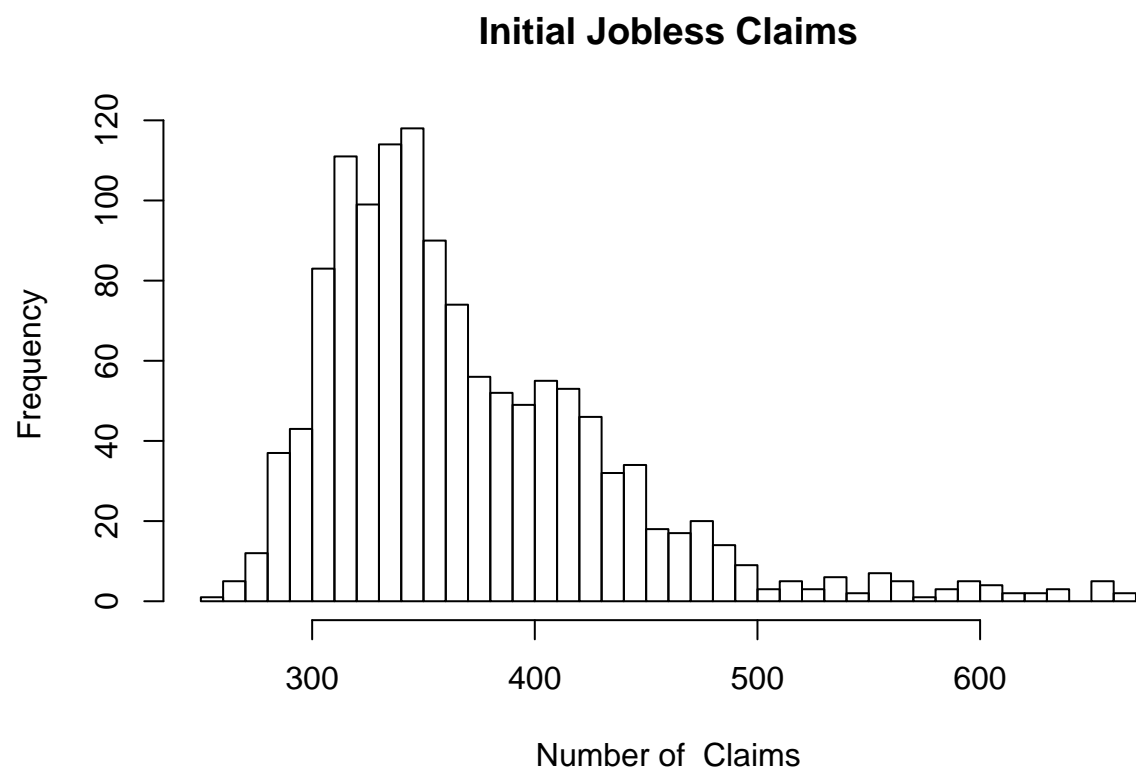
```
plot.ts(data.ts, xlab = "Years", ylab = "Number of Claims", main = "Initial Jobless Claims")
```



Part e2

What the histogram doesn't show is how the values in the distribution occur over time. It does show the distribution of the values over time. The number of bins is selected based on the representation that provides a more visually complete rendering of the distribution of the values of the time series.

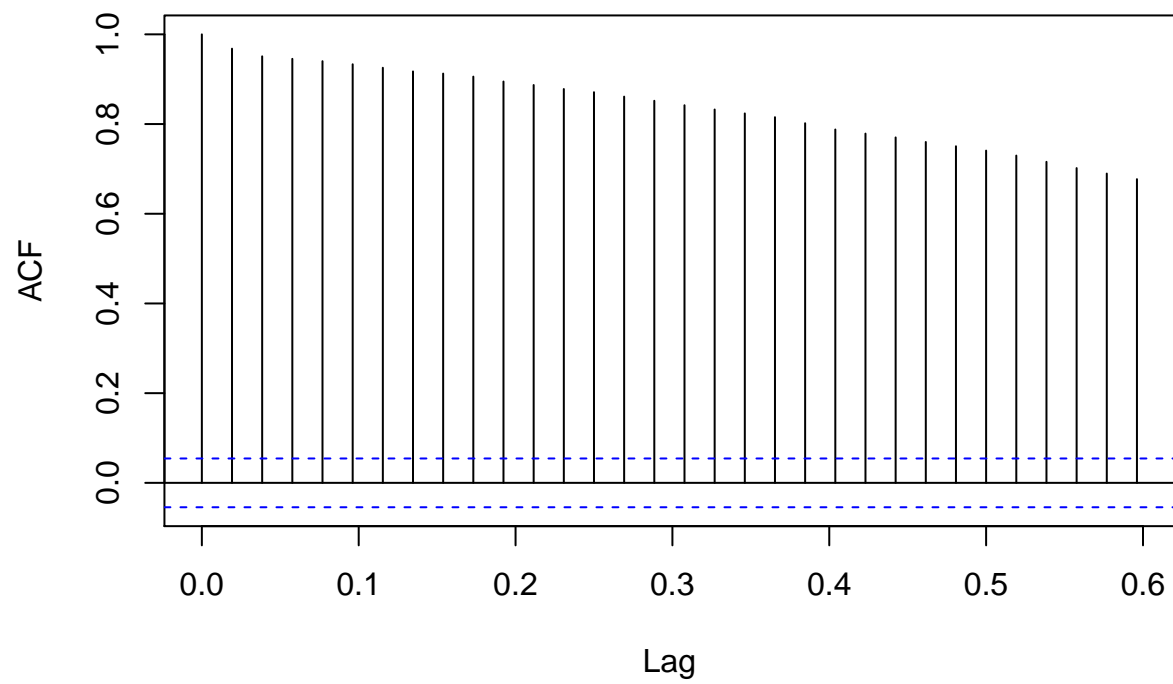
```
hist(data.ts, xlab = "Number of Claims", main = "Initial Jobless Claims",
      breaks = 30)
```



Part e3

```
acf(data.ts)
```

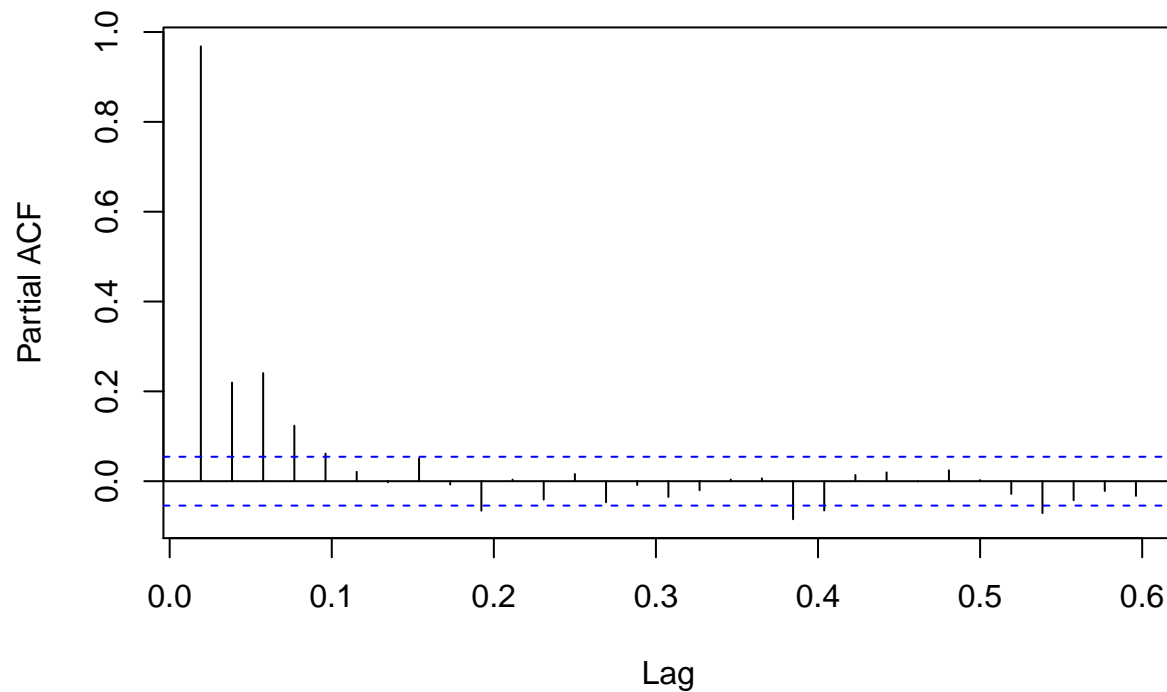
Series data.ts



Part e4

```
pacf(data.ts)
```

Series data.ts



Part e5

```
lag.plot(data.ts, lags = 9, layout = c(3, 3), diag = TRUE, disg.col = "red",  
         main = "Autocorrelation between Initial Jobless Claims and its own lags")
```

```
## Warning in plot.window(...): "disg.col" is not a graphical parameter
```

```
## Warning in plot.xy(xy, type, ...): "disg.col" is not a graphical parameter
```

```
## Warning in title(...): "disg.col" is not a graphical parameter
```

```
## Warning in box(...): "disg.col" is not a graphical parameter
```

```
## Warning in axis(side, ..., xpd = NA): "disg.col" is not a graphical  
## parameter
```

```
## Warning in axis(side, ..., xpd = NA): "disg.col" is not a graphical  
## parameter
```

```
## Warning in plot.window(...): "disg.col" is not a graphical parameter
```

```
## Warning in plot.xy(xy, type, ...): "disg.col" is not a graphical parameter
```

```
## Warning in title(...): "disg.col" is not a graphical parameter

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## Warning in plot.xy(xy, type, ...): "disg.col" is not a graphical parameter

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## Warning in box(...): "disg.col" is not a graphical parameter

## Warning in axis(side, ..., xpd = NA): "disg.col" is not a graphical
## parameter

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## Warning in title(...): "disg.col" is not a graphical parameter

## Warning in box(...): "disg.col" is not a graphical parameter

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## Warning in title(...): "disg.col" is not a graphical parameter

## Warning in box(...): "disg.col" is not a graphical parameter

## Warning in axis(side, ..., xpd = NA): "disg.col" is not a graphical
## parameter

## Warning in plot.window(...): "disg.col" is not a graphical parameter

## Warning in plot.xy(xy, type, ...): "disg.col" is not a graphical parameter

## Warning in title(...): "disg.col" is not a graphical parameter
```

```
## Warning in box(...): "disg.col" is not a graphical parameter

## Warning in axis(side, ..., xpd = NA): "disg.col" is not a graphical
## parameter

## Warning in plot.window(...): "disg.col" is not a graphical parameter

## Warning in plot.xy(xy, type, ...): "disg.col" is not a graphical parameter

## Warning in title(...): "disg.col" is not a graphical parameter

## Warning in box(...): "disg.col" is not a graphical parameter

## Warning in axis(side, ..., xpd = NA): "disg.col" is not a graphical
## parameter

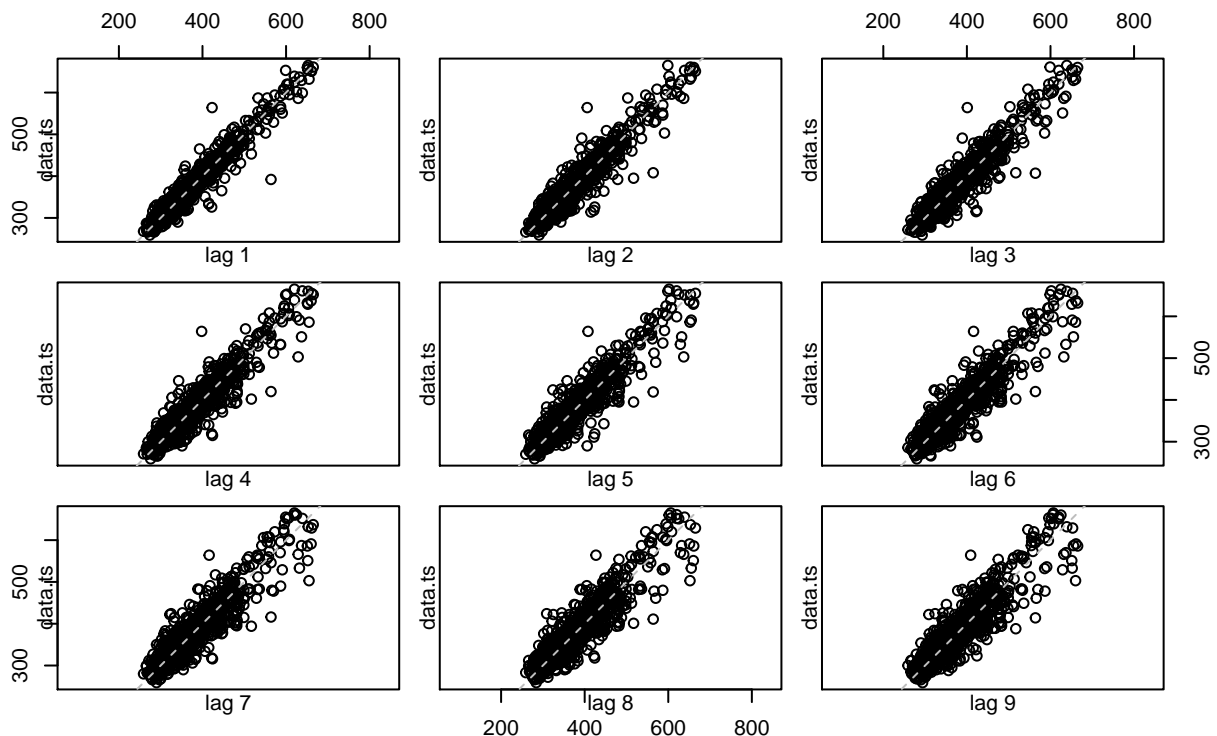
## Warning in plot.window(...): "disg.col" is not a graphical parameter

## Warning in plot.xy(xy, type, ...): "disg.col" is not a graphical parameter

## Warning in title(...): "disg.col" is not a graphical parameter

## Warning in box(...): "disg.col" is not a graphical parameter
```

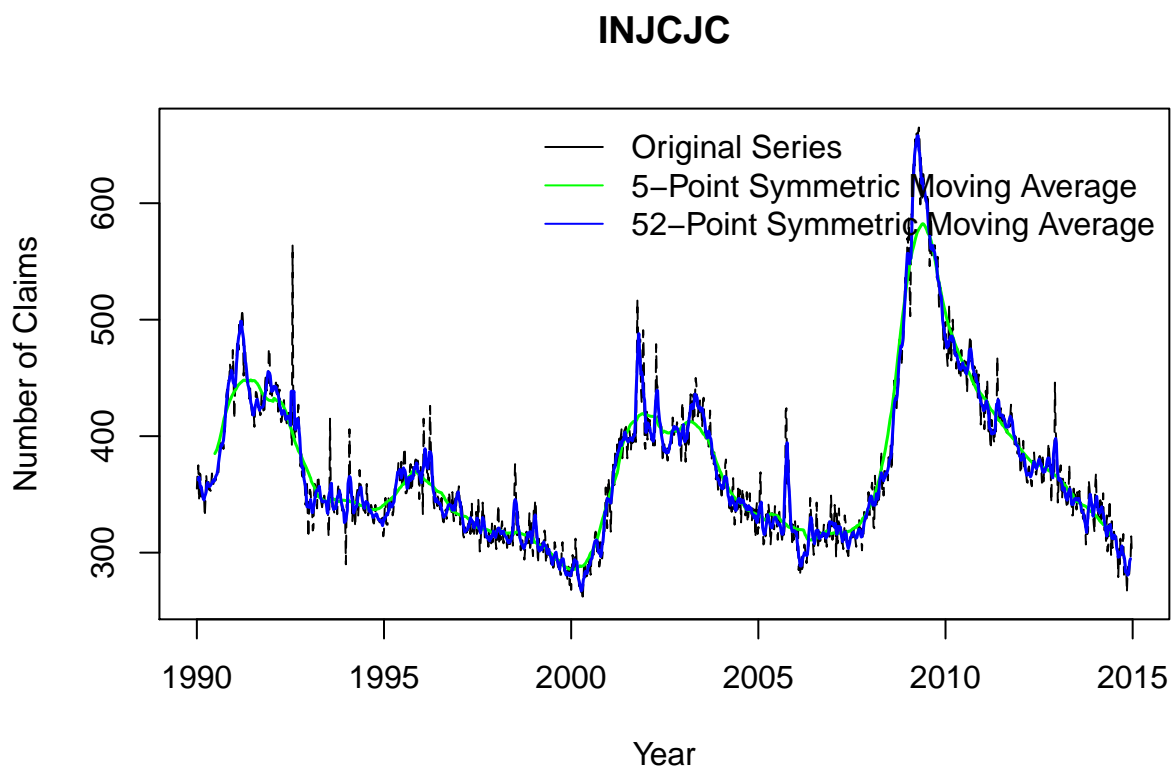
Autocorrelation between Initial Jobless Claims and its own lags



Part f1

```
# Create a 52 weeks and a 5 weeks moving average smoother
mo52 = filter(data.ts, sides = 2, rep(1, 52)/52)
mo5 = filter(data.ts, sides = 2, rep(1, 5)/5)

plot(data.ts, main = "INJCJC", pch = 4, lty = 5, lwd = 1, xlab = "Year",
      ylab = "Number of Claims")
lines(mo52, lty = 1, lwd = 1.5, col = "green")
lines(mo5, lty = 1, lwd = 1.5, col = "blue")
# Add Legend
leg.txt <- c("Original Series", "5-Point Symmetric Moving Average", "52-Point Symmetric Moving Average")
legend("topright", legend = leg.txt, lty = c(1, 1, 1), col = c("black",
  "green", "blue"), bty = "n", cex = 1, merge = TRUE, bg = 336)
```



Part f2

```
wk = time(data.ts) - mean(time(data.ts))
wk2 = wk^2
wk3 = wk^3
cs = cos(2 * pi * wk)
sn = sin(2 * pi * wk)
```

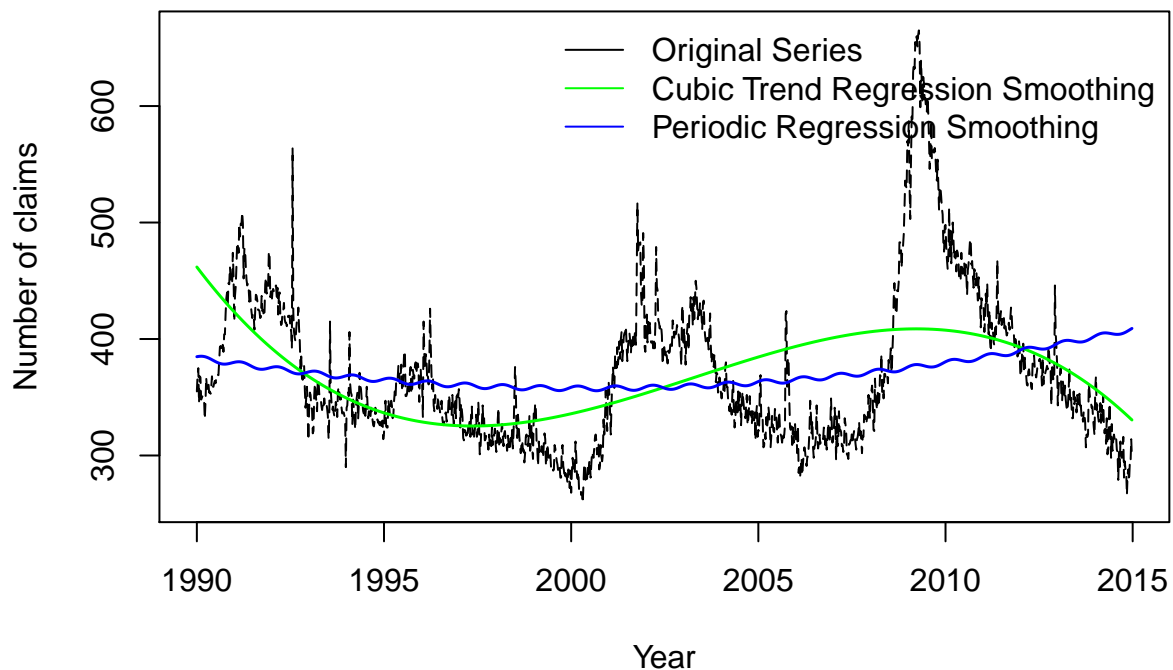


```

reg1 = lm(data.ts ~ wk + wk2 + wk3, na.action = NULL)
reg2 = lm(data.ts ~ wk + wk2 + cs + sn, na.action = NULL)
plot(data.ts, main = "Initial Jobless Claims (Weekly Series) and Regression Smoothing",
     pch = 4, lty = 5, lwd = 1, xlab = "Year", ylab = "Number of claims")
lines(fitted(reg1), lty = 1, lwd = 1.5, col = "green")
lines(fitted(reg2), lty = 1, lwd = 1.5, col = "blue")
# Add Legend
leg.txt <- c("Original Series", "Cubic Trend Regression Smoothing", "Periodic Regression Smoothing")
legend("topright", legend = leg.txt, lty = c(1, 1, 1), col = c("black",
    "green", "blue"), bty = "n", cex = 1, merge = TRUE, bg = 336)

```

Initial Jobless Claims (Weekly Series) and Regression Smoothing



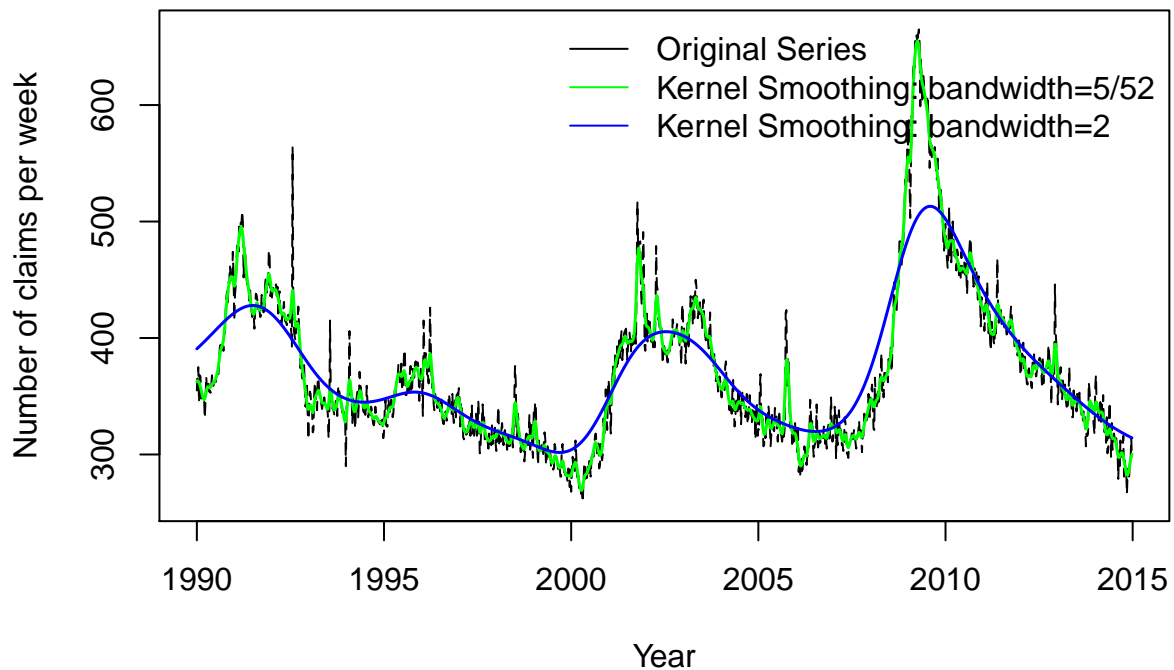
Part f3

```

plot(data.ts, main = "Initial Jobless Claims (Weekly Series) and Kernel Smoothing",
     pch = 4, lty = 5, lwd = 1, xlab = "Year", ylab = "Number of claims per week")
lines(ksmooth(time(data.ts), data.ts, "normal", bandwidth = 5/52), lty = 1,
     lwd = 1.5, col = "green")
lines(ksmooth(time(data.ts), data.ts, "normal", bandwidth = 2), lty = 1,
     lwd = 1.5, col = "blue")
# Add Legend
leg.txt <- c("Original Series", "Kernel Smoothing: bandwidth=5/52", "Kernel Smoothing: bandwidth=2")
legend("topright", legend = leg.txt, lty = c(1, 1, 1), col = c("black",
    "green", "blue"), bty = "n", cex = 1, merge = TRUE, bg = 336)

```

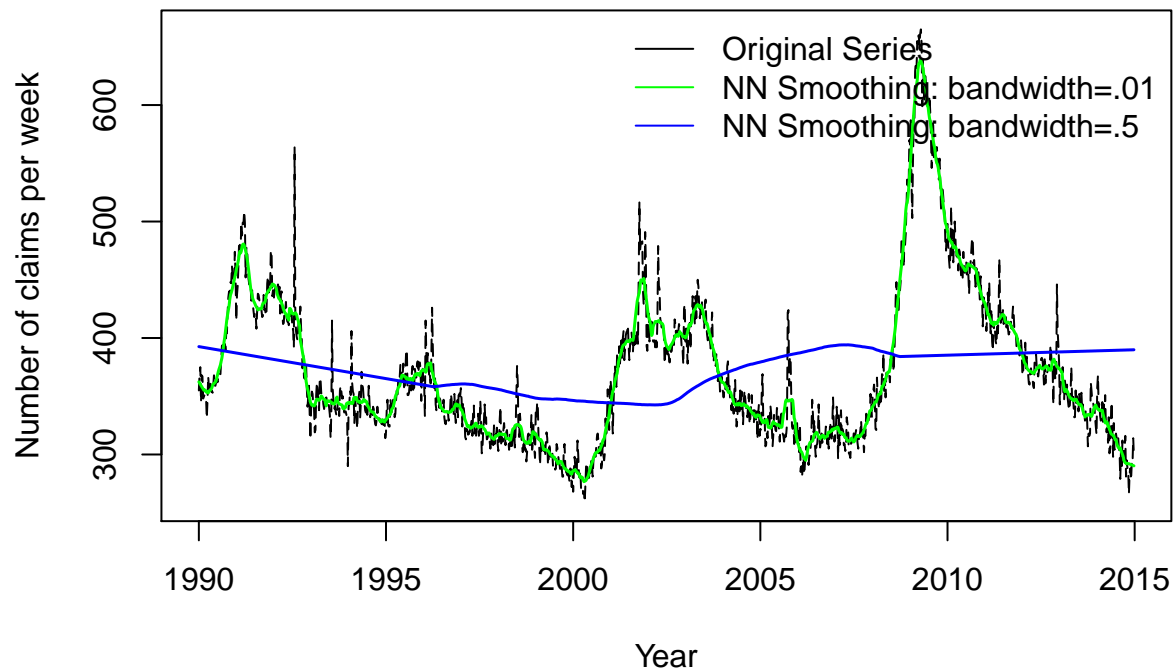
Initial Jobless Claims (Weekly Series) and Kernel Smoothing



Part f4

```
plot(data.ts, main = "Initial Jobless Claims (Weekly Series) and Nearest Neighborhood Smoothing",
     pch = 4, lty = 5, lwd = 1, xlab = "Year", ylab = "Number of claims per week")
lines(supsmu(time(data.ts), data.ts, span = 0.01), lty = 1, lwd = 1.5,
     col = "green")
lines(supsmu(time(data.ts), data.ts, span = 0.5), lty = 1, lwd = 1.5, col = "blue")
# Add Legend
leg.txt <- c("Original Series", "NN Smoothing: bandwidth=.01", "NN Smoothing: bandwidth=.5")
legend("topright", legend = leg.txt, lty = c(1, 1, 1), col = c("black",
  "green", "blue"), bty = "n", cex = 1, merge = TRUE, bg = 336)
```

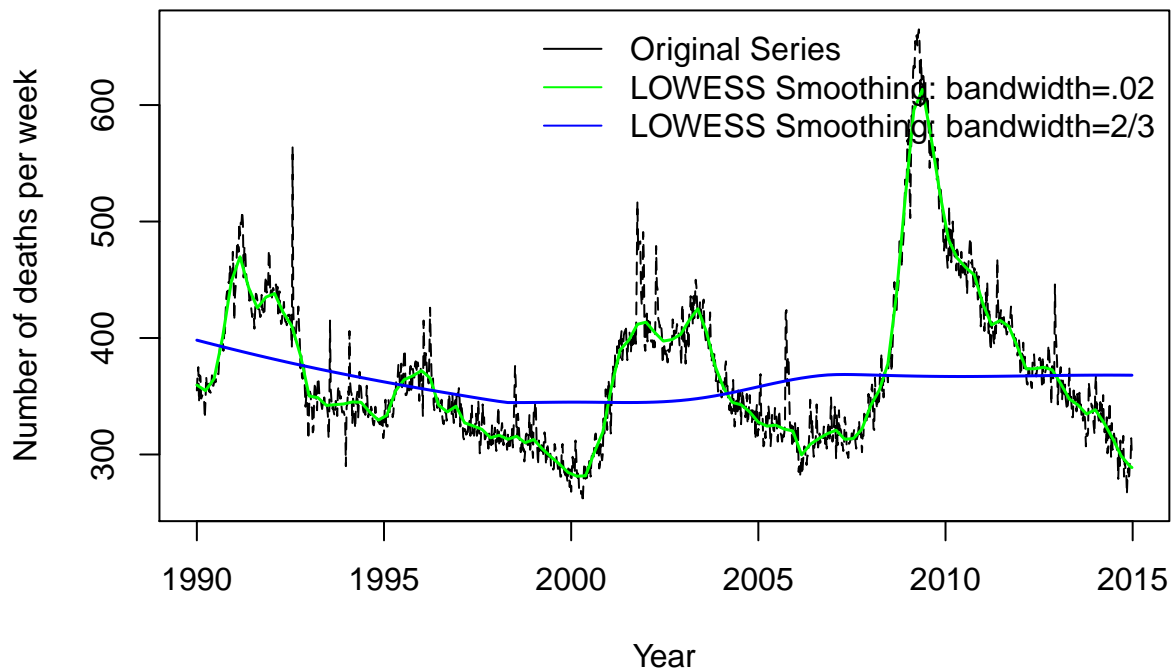
Initial Jobless Claims (Weekly Series) and Nearest Neighborhood Smoo



Part f5

```
plot(data.ts, main = "Initial Jobless Claims (Weekly Series) and LOWESS Smoothing",
     pch = 4, lty = 5, lwd = 1, xlab = "Year", ylab = "Number of deaths per week")
lines(lowess(data.ts, f = 0.02), lty = 1, lwd = 1.5, col = "green")
lines(lowess(data.ts, f = 2/3), lty = 1, lwd = 1.5, col = "blue")
# Add Legend
leg.txt <- c("Original Series", "LOWESS Smoothing: bandwidth=.02", "LOWESS Smoothing: bandwidth=2/3")
legend("topright", legend = leg.txt, lty = c(1, 1, 1), col = c("black",
  "green", "blue"), bty = "n", cex = 1, merge = TRUE, bg = 336)
```

Initial Jobless Claims (Weekly Series) and LOWESS Smoothing



Part f6

```
plot(data.ts, main = "Initial Jobless Claims (Weekly Series) and Smoothing Splines",
     pch = 4, lty = 5, lwd = 1, xlab = "Year", ylab = "Number of claims per week")
lines(smooth.spline(time(data.ts), data.ts, spar = 0.05), lty = 1, lwd = 1.5,
     col = "green")
lines(smooth.spline(time(data.ts), data.ts, spar = 0.9), lty = 1, lwd = 1.5,
     col = "blue")
# Add Legend
leg.txt <- c("Original Series", "Spline: Smoothing Parameter=.05", "Spline: Smoothing Parameter=0.8")
legend("topright", legend = leg.txt, lty = c(1, 1, 1), col = c("black",
"green", "blue"), bty = "n", cex = 1, merge = TRUE, bg = 336)
```

Initial Jobless Claims (Weekly Series) and Smoothing Splines

