# Homework 3

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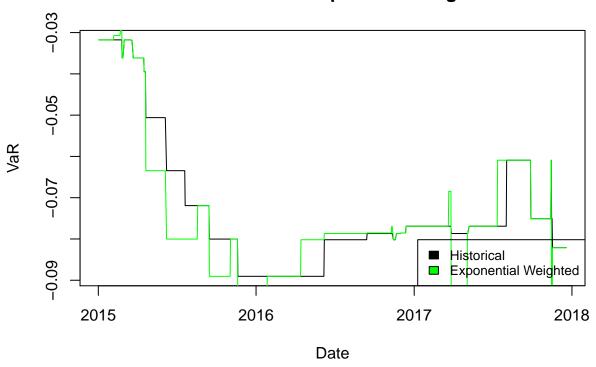
#### Question 1

1.

```
setwd("/Users/auroracappadocian/Downloads/409FRM/HW3")
suppressMessages(require(data.table))
suppressMessages(require(lubridate))
suppressMessages(require(zoo))
suppressMessages(require(ggplot2))
suppressMessages(require(dplyr))
suppressMessages(library(knitr))
suppressMessages(require(quantmod))
returns_data <- as.data.table(read.csv("hw3_returns2.csv"))</pre>
returns_data[,Date:=mdy(Date)]
rollingWindow <- 252</pre>
c < -0.99
lambda <- 0.995
cal_hist_VaR <- function(return, c){</pre>
  sorted_ret <- sort(return)</pre>
  Var_point <- ceiling(length(sorted_ret) * (1-c))</pre>
  var <- sorted_ret[Var_point]</pre>
  return(var)
}
cal_exponential_VaR <- function(return, c){</pre>
  n <- length(return)</pre>
  weight \leftarrow lambda^seq((n-1), 0, -1) * (1-lambda)/(1-lambda<math>^n)
  df <- cbind(weight, return)</pre>
  df_sorted <- df[order(return),]</pre>
  df_sorted <- cbind(df_sorted, cumsum=cumsum(df_sorted[,1]))</pre>
  pos <- which(df_sorted[,3]>(1-c))[1]
  return(df_sorted[pos,2])
}
# calculating historical VaR
returns_data[, historicalVaR := shift(rollapply(returns_data$Return, rollingWindow, function (return){
  return(cal_hist_VaR(return,c))
}, fill=NA, align="right"))
]
# calculating exponential Weighted VaR
returns_data[, exponentialWeightedVaR := shift(rollapply(returns_data$Return, rollingWindow, function(r
  return(cal_exponential_VaR(return, c))
}, fill=NA, align="right"))]
returns_data_2015<- returns_data[Date >= "2015-01-01"]
plot(x = returns_data_2015$Date, y= returns_data_2015$historicalVaR, type = "1",
     main="HistoricalVaRs and ExponentialWeightedVaR", xlab = "Date",
     ylab= "VaR")
```

```
lines(x= returns_data_2015$Date, y= returns_data_2015$exponentialWeightedVaR, col = "green")
legend("bottomright",legend=c("Historical","Exponential Weighted"),fill=c("black","green"), cex = 0.8)
```

## HistoricalVaRs and ExponentialWeightedVaR



The plot of two VaR is shown above.

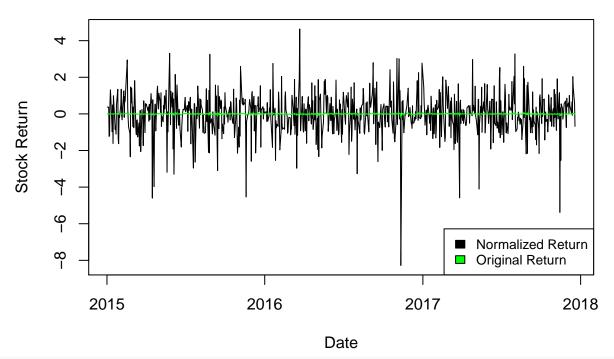
2.

```
data1<-returns_data[Date > "2014-12-31"]
n=length(data1$Date)
m1=sum(data1$Return < (data1$historicalVaR))</pre>
m2=sum(data1$Return < (data1$exponentialWeightedVaR))</pre>
testVal1 <- -2*log(c^(n-m1)*(1-c)^m1)+2*log((1-(m1/n))^(n-m1)*(m1/n)^m1)
testVal2 <- -2*log(c^(n-m2)*(1-c)^m2)+2*log((1-(m2/n))^(n-m2)*(m2/n)^m2)
chisq_val <- qchisq(p=c,df=1)</pre>
backtest=c(testVal1,testVal2,chisq_val)
names(backtest) <- c("Test Val_histVaR", "Test Val_expVaR Val", "Chisq Val")</pre>
exception=c(m1,m2)
names(exception) <- c("Exceptions of histVaR", "Exceptions of expVaR Val")</pre>
print(backtest)
##
      Test Val_histVaR Test Val_expVaR Val
                                                        Chisq Val
##
               4.568662
                                    2.331818
                                                         6.634897
print(exception)
      Exceptions of histVaR Exceptions of expVaR Val
##
##
```

For Historical method, there are 14 exceptions. For exponential weighted method, there are 12 exceptions. Both of them are fine to calculate VaR since the test statistics of them are all within the 1% threshold 6.63.

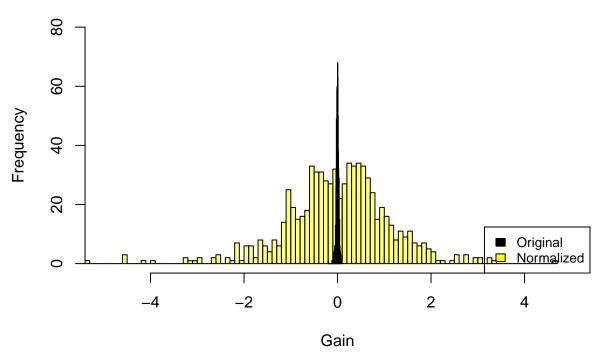
3.

# Normalized Gains using prev. volatility vs. Original Gains



hist(returns\_data\_2015\$normalized,breaks=100,col=scales::alpha("yellow",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(-5,5),ylim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(0,80),hist(returns\_data\_2015\$Return,breaks=100,add=TRUE,col=scales::alpha("blue",0.5),xlim=c(0,80),xlim

# Normalized Gains using prev. volatility vs. Original Gains



The Distribution is shown above.

4. I will use the normalized Return based on the last month volatility and exponetial weighted method to calculate the VaR.

```
rollingWindow <- 252</pre>
lambda <- 0.995
returns_data[, exponentialWeighted_NormVaR := shift(rollapply(returns_data$normalized, rollingWindow, f
  cal_exponential_VaR(return, c)
}, fill=NA, align="right"))]
data1<-returns_data[Date > "2014-12-31"]
m3=sum(data1$normalized < (data1$exponentialWeighted_NormVaR))
testVal3 <- -2*log(c^(n-m3)*(1-c)^m3) + 2*log((1-(m3/n))^(n-m3)*(m3/n)^m3)
chisq_val <- qchisq(p=c,df=1)</pre>
backtest=c(testVal1,testVal2,testVal3,chisq_val)
names(backtest) <- c("Test Val_histVaR","Test Val_expVaR Val","Test Val_NormVaR Val","Chisq Val")</pre>
exception=c(m1,m2,m3)
names(exception) <- c("Exceptions of histVaR", "Exceptions of expVaR Val", "Exceptions of NormVaR Val")
print(backtest)
##
       Test Val_histVaR Test Val_expVaR Val Test Val_NormVaR Val
##
             4.56866243
                                   2.33181815
                                                         0.03570523
##
              Chisq Val
             6.63489660
print(exception)
##
       Exceptions of histVaR Exceptions of expVaR Val
##
                                                      12
## Exceptions of NormVaR Val
##
```

There are 8 exceptions in the new method.

5. There are 748 data entries from 2015 to 2017. By taking 99% VaR on these observation, we will see 8 exceptions. Hence I would suggest to use the method in question4.

### Question 2

- 1. We can use the proof by contradiction. If at most 2 people were born in the same year, for 3 years, there will be at most 6 people. However, there are 8 people. Hence at least 3 of them are born within the same one-year period.
- 2. Assume Normal distribution

$$VaR_5 = \sigma_5 \Phi^{-1}(0.98) = 10$$
  
 $\sigma_5 = 10/\Phi^{-1}(0.98) = 4.869144$   
 $\sigma_{10} = \sigma_5 \sqrt{2}$ 

Hence

$$VaR_{10} = \sigma_{10}\Phi^{-1}(0.99) = \sigma_5\sqrt{2}\Phi^{-1}(0.99) = 16.01925$$

3.

In other other words, we want to calculate the probability of having more than one exception within 21 consecutive days. n is the number of exceptions within the 21 consecutive days.

$$n \sim Bernoulli(21, 1-c)$$
 
$$P(n > 1) = 1 - P(n = 0) - P(n = 1) = 1 - C_{21}^{0}(1-c)^{21} - C_{21}^{1}(1-c)^{20}c^{1}$$

For bunching test, we can calculate the probability of P(n > 1) for each day in a rolling window of 21.

4. We can use VIX to calculate the volatility of return of S&P500  $\sigma$ . We use the mean of previous 1 year return  $\mu$ . Then the 2-day 99% VaR will be

$$VaR = -1(\mu + \sqrt{2}\sigma z(1-c))$$