

Model

```
##### Set Up Working Directory #####

# Name working directory (must use forward slashes as separators and have a
  ↳ slash at the end)
myWD = "H:/Baruch/Fall_2016/Financial_Statistics/Project/"

# Set my working directory as myWD, an example
setwd(myWD)

#####

#####
##### Load S&P500 Tickers #####
library(quantmod)
mySPReturnData <- read.csv(file = "Data/BloombergSP500HistoryFinal.csv", header=
  ↳ TRUE, sep=",", na.strings=c("NA"), stringsAsFactors=FALSE) #Path must be
  ↳ relative to myWD
myTicks <- colnames(mySPReturnData)
myTicksVec <- unlist(strsplit(myTicks, split="_"))
myTicksVec <- myTicksVec[-1] #vector of S&P 500 tickers
myTicksVec
head(myTicksVec) #show the first part of the ticker vector
tail(myTicksVec) #show the last part of the ticker vector

#####
##### Correlation Matrix with Tawny #####
##### Setup #####
library(tawny) # tawny package must be installed
library(tawny.types) # tawny.types package must be installed
require(tawny)
require(tawny.types)

##### Do a Test with a Few Stocks #####
#get returns
# test with a few stocks and inspect first
myTicksVecTest <- c('A', 'AA', 'AAPL')
m = length(myTicksVecTest)
Q = 5 # quality parameter
t=m*Q #should be 15 in this case
myEndDate <- as.Date("2016-11-30")
hTest <- getPortfolioReturns(myTicksVecTest, obs=t, end=myEndDate)
hTest
pTest <- TawnyPortfolio(hTest, t) #Represents a portfolio. Contains information
  ↳ about the portfolio composition, returns, window, etc. TawnyPortfolio(
```

```

    ↪ returns , window = 90)
str(pTest) #view a summary
numTickersTestCheck ← length(pTest[[1]])
numDates ← pTest$obs
corrsample ← cor.empirical(pTest$returns) #sample correlation matrix
corrsample # inspect

##### Build a Portfolio from the S&P 500
    ↪ #####
m = length(myTicksVec) # number of stocks
Q = 5 # quality parameter
t=m*Q # number of observations for each stock
myEndDate ← as.Date("2016-11-30")
h ← getPortfolioReturns(myTicksVec, obs=t, end=myEndDate)
apply(h, function(h) sum(is.na(h))) #view any missing values (there should be
    ↪ none)
#Note: The next step can take a while for large m and t!
p ← TawnyPortfolio(h, t) #Represents a portfolio. Contains information about
    ↪ the portfolio composition, returns, window, etc. TawnyPortfolio(returns,
    ↪ window = 90)
class(p) # p is a list
str(p)#view a summary

#check how many tickers were actually included
p[1]# view the tickers
numTickersCheck = length(p[[1]])#count the tickers included in the portfolio
paste((m - numTickersCheck), "_tickers_were_dropped.")
#check how many datapoints for each ticker were included
numDatesCheck ← p$obs
paste((t - numDatesCheck), "_dates_were_dropped.")

##### Sample Correlation Matrix #####
#create a raw sample correlation matrix
corrsample ← cor.empirical(p$returns)
myeigenraw ← eigen(corrsample)$values
myeigenraw
myminraw= min(myeigenraw); myminraw
mymaxraw= max(myeigenraw); mymaxraw
numEigenRaw ← length(myeigenraw)
my2ndmaxraw = sort(myeigenraw, partial=numEigenRaw-1)[numEigenRaw-1]
my2ndmaxraw

##### Correlation Matrix Denoised Using RMT
    ↪ #####
##### Fit the density #####

```

```

##### Real sample #####
##### Prep the data by removing the largest eigenvalues #####
myeigenrawSorted <-sort(myeigenraw) #sorted
class(myeigenrawSorted)

myeigenrawSortedList <-as.list(myeigenrawSorted)
myeigenrawSortedTruncatedList <- myeigenrawSortedList[myeigenrawSortedList < 5]
  ↳ # remove largest eigenvalues before fitting
myeigenrawSortedTruncatedList <- myeigenrawSortedTruncatedList[
  ↳ myeigenrawSortedTruncatedList >0]
#check
truncMax<-max(unlist(myeigenrawSortedTruncatedList)) #3.845671
min(unlist(myeigenrawSortedTruncatedList))
truehist(unlist(myeigenrawSortedTruncatedList), h=0.1, xlim=range(c(0,truncMax
  ↳ ))) # h is the bin width

myeigenrawSortedTruncatedVector <- as.vector(unlist(
  ↳ myeigenrawSortedTruncatedList))
head(myeigenrawSortedTruncatedVector)

##### Fit the Marcenko Pastur distribution and denoise the
  ↳ matrix #####
library(fitdistrplus) #requires the fitdistrplus package
library(covmat) #covmat package must be installed (needed for dmp)
#fit Marcenko Pastur to the bulk of the distribution
myfitRaw <- fitdist(myeigenrawSortedTruncatedVector, dmp, list(var=1, ndf=
  ↳ numDatesCheck, pdim=numTickersCheck, svr=(numDatesCheck/numTickersCheck)
  ↳ ), method="mge", gof="CvM")
#sigma^2 0.4419794
#Q = 2.0470204

##### Replace eigenvalues below the cutoff with average
  ↳ #####
sigmaFitted = sqrt(myfitRaw$estimate["var"]); sigmaFitted
QFitted = myfitRaw$estimate["svr"]; QFitted
lambdaMaxCutoff = sigma * (1 + sqrt(1/Q))^2; lambdaMaxCutoff
meanRandom = mean(myeigenrawSortedTruncatedList<=lambdaMaxCutoff)
#replace all values below the cutoff with the mean
myeigenFiltered <-myeigenrawSortedList
myeigenFiltered[myeigenFiltered<= lambdaMaxCutoff] <- meanRandom

myeigenclean <-unlist(myeigenFiltered)
length(myeigenclean)
min(myeigenclean) # now the average

##### Plot the Histograms

```

```

→ #####
par(mfrow=c(1,2))#graphical parameters
library(MASS)# MASS package must be installed to run the histograms
#####Histograms
truehist(myeigenraw, h=0.1, prob=TRUE) # h is the bin width
truehist(myeigenclean, h=0.05, ylim=c(0,2), prob=TRUE)

myminclean = min(myeigenclean); myminclean
mymaxclean = max(myeigenclean); mymaxclean

#zoom in on the first part of the scale
maxx = 20 # x-axis cut-off for the zoom in
truehist(myeigenraw, h=0.1, xlim=range(c(0,maxx)), prob=TRUE) # h is the bin
→ width

Q=numDatesCheck/numTickersCheck; Q
sigma=1
#superimpose Marcenko-Pastur on the zoomed-in sample
lambdaMax = sigma * (1 + sqrt(1/Q))^2
lambdaMax
lambdaMin = sigma * (1 - sqrt(1/Q))^2
lambdaMin
par(col="red", lwd=2);
curve ( Q/(2*pi*sigma^2) * sqrt((lambdaMax - x)*(x-lambdaMin))/x, from =
→ lambdaMin, to = lambdaMax, add=TRUE);
segments(0, 0, lambdaMin, 0, col="red")
segments(lambdaMax, 0, maxx, 0, col="red")

#superimpose the fitted Marcenko-Pastur on the zoomed-in sample
sigma = sigmaFitted
Q=QFitted
lambdaMax = sigma * (1 + sqrt(1/Q))^2
lambdaMax
lambdaMin = sigma * (1 - sqrt(1/Q))^2
lambdaMin
par(col="green", lwd=2);
curve ( Q/(2*pi*sigma^2) * sqrt((lambdaMax - x)*(x-lambdaMin))/x, from =
→ lambdaMin, to = lambdaMax, add=TRUE);
segments(0, 0, lambdaMin, 0, col="green")
segments(lambdaMax, 0, maxx, 0, col="green")

truehist(myeigenclean, h=0.1, xlim=range(c(0,maxx)), ylim=c(0,2), prob=TRUE)
#output in: eigenvaluesbeforeandafter.bmp

##### De-Diagonalized to get Correlation Matrix
→ #####
DiagonalFilteredM = matrix(nrow = numTickersCheck, ncol = numTickersCheck )

```

```

#assign filtered eigenvalues back to a matrix
for(i in 1:(numTickersCheck )){
  for (j in 1:numTickersCheck ){

    if (i == j){
      DiagonalFilteredM[i,j] ← as.vector(myeigenclean)[i]
    }

  }
}
DiagonalFilteredM[is.na(DiagonalFilteredM)]←0

#Since the intial correaltion matrix was diagonalized as  $L = V^{-1} C V$ , so  $V L$ 
→  $= C V$  and  $V L V^{-1} = C$ 
myRotM ← eigen(corrsample)$vectors # rotation matrix of eigenvectors used to
→ go from the original matrix to the diagonalized eigenvalue matrix
myFilteredM ← myRotM %*%DiagonalFilteredM %*% solve(myRotM) #rotate back to
→ original correlation matrix
myFilteredM ← (myFilteredM + t(myFilteredM))/2 #symmetrize

detach_package(covmat)# need to detach to avoid conflict with the Matrix
→ package
library(Matrix)# Matrix package must be installed

myFilteredM ← as.matrix(nearPD(myFilteredM, corr=TRUE)$mat) # find nearest
→ correlation matrix (positive semi-definite)
head(myFilteredM)
sum(diag(myFilteredM))# check that the eigenvalues add to m

diag(DiagonalFilteredM)#Tawny -type methodology
#c.clean ← myRotM %*% DiagonalFilteredM %*% t(myRotM )
#diags ← diag(c.clean) %o% rep(1, nrow(c.clean))
#c.clean ← c.clean / sqrt(diags * t(diags))

##### KL Divergence Using Tawny Package #####
myKL ← divergence.kl(corrsample, myFilteredM);myKL

myKL ← divergence.kl(myFilteredM, corrsample);myKL #1st argument is assumed to
→ be the "true distribution"

myFilteredMTawny ←denoise(p, RandomMatrixDenoiser())
#uses:
#lambda.plus ← estimator$cutoff.fn(correlation, es, estimator)
#estimator$clean.fn(es, lambda.plus)

```

```

sum(diag(myFilteredMTawny))# check that the eigenvalues add to m

myKL1 ← divergence.kl(myFilteredM, myFilteredMTawny);myKL1 #1st argument is
  ↪ assumed to be the "true distribution"

#####Utility Function#####
detach_package ← function(pkg, character.only = FALSE)
{
  if(!character.only)
  {
    pkg ← deparse(substitute(pkg))
  }
  search_item ← paste("package", pkg, sep = ":")
  while(search_item %in% search())
  {
    detach(search_item, unload = TRUE, character.only = TRUE)
  }
}

```

RMT Illustrations

```

#####
##### Empirical Illustration of Random Matrix Theory Results
  ↪ #####
par(mfrow=c(1,2))#graphical parameters

#####
##### Empirical Illustration of Random Matrix Theory Results
  ↪ #####
par(mfrow=c(1,2))#graphical parameters

#generate a symmetric random matrix whose entries are normally distributed
  ↪ and the off-diagonal elements have variance 1/N
n ← 5000;
m ← array(rnorm(n^2),c(n,n)); #rnorm creates n^2 random variates and allocates
  ↪ them to an array
m2 ← (m+t(m))/sqrt(2*n);# Make m symmetric and normalize for variance 1
lambda ← eigen(m2, symmetric=T, only.values = T);
e ← lambda$values;
hist(e,breaks=seq(-2.01,2.01,.02), main=NA, xlab="Eigenvalues",freq=F)
sigma ← 1;
par(col="red");
curve ( 1/(2*pi * sigma^2)*sqrt(4*sigma^2-x^2), -2*sigma, 2*sigma, add=TRUE);

#output in: random_matrix_eigenvalue_dist.png
#####

```

```

#generate a uniform random matrix whose entries are uniformly distributed
  ↪ and the off-diagonal elements have variance 1/N
#demonstrate 'universality' – distributions other than normal lead to the same
  ↪ asymptotic result
n ← 5000;
mu ← array(runif(n^2),c(n,n)); #runif generates 5000^2 random deviates from
  ↪ the uniform distribution and adds fills them into an n x n array
mu2 ← sqrt(12)*(mu+t(mu)-1)/sqrt(2*n); # Make m symmetric and normalize for
  ↪ variance 1
lambdau ← eigen(mu2, symmetric=T, only.values = T);
eu ← lambdau$values;
histeu←hist(eu,breaks=seq(-2.01,2.01,0.02), main=NA, xlab="Eigenvalues",freq=F
  ↪ )
#The density of eigenvalues is a semicircle, as predicted by Wigner's
  ↪ semicircle law.

sigma ← 1;
par(col="pink");
curve ( 1/(2*pi * sigma^2)*sqrt(4*sigma^2-x^2), -2*sigma, 2*sigma, add=TRUE);

#output in: random_matrix_normal_and_uniform_dist_w_semicircular_law.png
#####

#generate a random correlation matrix
par(mfrow=c(1,1))#graphical parameters
t ← 5000;
m ← 1000;
Q ← t/m
h ← array(rnorm(m*t),c(m,t)); # Time series in rows
e ← h %*% t(h)/t; # Form the correlation matrix
lambdae ← eigen(e, symmetric=T, only.values = T);
ee ← lambdae$values;
hist(ee,breaks=seq(0.01,3.01,.02), main="Random_Sample_Correlation",xlab="
  ↪ Eigenvalues",freq=F)

#superimpose Marcenko–Pastur
lambdaMax = sigma * (1 + sqrt(1/Q))^2
lambdaMax
lambdaMin = sigma * (1 - sqrt(1/Q))^2
lambdaMin
par(col="green");
curve ( Q/(2*pi*sigma^2) * sqrt((lambdaMax - x)*(x-lambdaMin))/x, from =
  ↪ lambdaMin, to = lambdaMax, add=TRUE);
segments(0, 0, lambdaMin, 0, col="green")

```

```

#output in: random_matrix_correlation_w_MP_law.png
#####

##### Comparison of Scalar Asymptotic Results to Matrix Results
↪ #####
par(mfrow=c(1,2))#graphical parameters

#####Normal Distribution#####
x ← seq(-5, 5, length=100)
y ← dnorm(x, mean=0, sd=1)
plot(x, y, type="l", lty=1, xlab="x_value", ylab="Density", lwd=10, col="
↪ black")

#####Wigner's Semi-Circular Law#####
sigma ← 1;
par(col="pink");
curve(1/(2*pi * sigma^2)*sqrt(4*sigma^2-x^2), -2*sigma, 2*sigma, add=FALSE,
↪ xlab="x_value", ylab="Density", lwd=10, col="red");

par(mfrow=c(1,2))#graphical parameters
#####X^2 Distribution #####
x ← seq(0, 20, length=100)
numdf=1
y ← dchisq(x, df=numdf)
plot(x, y, type="l", lty=1, xlab="x_value", ylab="Density", lwd=10, col=1)
numdf=2
y ← dchisq(x, df=numdf)
lines(x, y, type="l", lty=1, xlab="x_value", ylab="Density", lwd=10, col=2)
numdf=5
y ← dchisq(x, df=numdf)
lines(x, y, type="l", lty=1, xlab="x_value", ylab="Density", lwd=10, col=3)
numdf=10
y ← dchisq(x, df=numdf)
lines(x, y, type="l", lty=1, xlab="x_value", ylab="Density", lwd=10, col=4)

colors ← seq(1:4)
labels ← c("df=1", "df=2", "df=3", "df=4")

legend("topright", inset=.05,
labels, lwd=10, lty=c(1, 1, 1, 1), col=colors)

##### Marcenko Pastur Density #####
x ← seq(0, 20, length=100)
Q=1
y ← dmp(x, var=1, svr=Q)
plot(x, y, type="l", lty=1, xlab="x_value", ylab="Density", lwd=10, col=1,
↪ ylim=c(0,2))

```



```

Q=2
y <-dmp(x, var=1, svr=Q)
lines(x, y , type="l" , lty=1, xlab="x_value" , ylab="Density" , lwd=10, col=2)
Q=5
y <-dmp(x, var=1, svr=Q)
lines(x, y , type="l" , lty=1, xlab="x_value" , ylab="Density" , lwd=10, col=3)
Q=10
y <- dmp(x, var=1, svr=Q)
lines(x, y , type="l" , lty=1, xlab="x_value" , ylab="Density" , lwd=10, col=4)

colors <- seq(1:4)
labels <- c("Q=1" , "Q=2" , "Q=3" , "Q=4")

legend("topright" , inset=.05,
      labels , lwd=10, lty=c(1, 1, 1, 1) , col=colors)

```