Econ 147 Lab 2

0

MSFT is the ticker for Microsoft, a multinational tech company based in Redmond, WA. It represents a share in the company. SBUX is the ticker for Starbucks, a multinational coffehouse chain. FMAGX is a mutual fund run by fidelity. It is a managed fund, meaning there are people that actively choose investments in that fund.

1

```
options(digits=4, width=70)
library(PerformanceAnalytics)
## Loading required package: xts
## Loading required package: zoo
##
## Attaching package: 'zoo'
  The following objects are masked from 'package:base':
##
##
##
       as.Date, as.Date.numeric
##
## Attaching package: 'PerformanceAnalytics'
## The following object is masked from 'package:graphics':
##
##
       legend
library(zoo)
library(tseries)
## Registered S3 method overwritten by 'quantmod':
##
     method
     as.zoo.data.frame zoo
##
```

```
# get monthly adjusted closing price data on MSFT, FMAGX and SBUX from Yahoo
# using the tseries function get.hist.quote(). Set sample to Jan 1998 through
# Dec 2009. Note: if you are not careful with the start and end dates
# or if you set the retclass to "ts" then results might look weird
# look at help on get.hist.quote
?get.hist.quote
# get the adjusted closing prices from Yahoo!
MSFT.prices = get.hist.quote(instrument="msft", start="1998-01-01",
                             end="2009-12-31", quote="AdjClose",
                             provider="yahoo", origin="1970-01-01",
                             compression="m", retclass="zoo")
## 'getSymbols' currently uses auto.assign=TRUE by default, but will
## use auto.assign=FALSE in 0.5-0. You will still be able to use
## 'loadSymbols' to automatically load data. getOption("getSymbols.env")
## and getOption("getSymbols.auto.assign") will still be checked for
## alternate defaults.
##
## This message is shown once per session and may be disabled by setting
## options("getSymbols.warning4.0"=FALSE). See ?getSymbols for details.
## time series ends
                      2009-12-01
# change class of time index to yearmon which is appropriate for monthly data
# index() and as.yearmon() are functions in the zoo package
index(MSFT.prices) = as.yearmon(index(MSFT.prices))
class(MSFT.prices)
## [1] "zoo"
colnames (MSFT.prices)
## [1] "Adjusted"
start(MSFT.prices)
## [1] "Jan 1998"
end(MSFT.prices)
## [1] "Dec 2009"
```

```
FMAGX.prices = get.hist.quote(instrument="fmagx", start="1998-01-01",
                              end="2009-12-31", quote="AdjClose",
                              provider="yahoo", origin="1970-01-01",
                              compression="m", retclass="zoo")
## time series ends
                      2009-12-01
index(FMAGX.prices) = as.yearmon(index(FMAGX.prices))
SBUX.prices = get.hist.quote(instrument="sbux", start="1998-01-01",
                             end="2009-12-31", quote="AdjClose",
                             provider="yahoo", origin="1970-01-01",
                             compression="m", retclass="zoo")
## time series ends
                      2009-12-01
index(SBUX.prices) = as.yearmon(index(SBUX.prices))
# create merged price data
lab4Prices.z = merge(MSFT.prices, FMAGX.prices, SBUX.prices)
# rename columns
colnames(lab4Prices.z) = c("MSFT", "FMAGX", "SBUX")
# calculate cc returns as difference in log prices
lab4Returns.z = diff(log(lab4Prices.z))
# See the document "Working with Time Series in R" on the
# class webpage for more details on zoo objects
# look at the return data
start(lab4Returns.z)
## [1] "Feb 1998"
end(lab4Returns.z)
## [1] "Dec 2009"
colnames(lab4Returns.z)
## [1] "MSFT" "FMAGX" "SBUX"
head(lab4Returns.z)
```

```
## MSFT FMAGX SBUX
## Feb 1998 0.127649 0.073069 0.078859
## Mar 1998 0.054533 0.049066 0.135701
## Apr 1998 0.006959 0.011513 0.060219
## May 1998 -0.060754 -0.045728 -0.002601
## Jun 1998 0.245154 0.043541 0.107312
## Jul 1998 0.014314 -0.007508 -0.243824
```

```
## [1] "matrix" "array"
```

```
colnames(ret.mat)
```

```
## [1] "MSFT" "FMAGX" "SBUX"
```

```
head(ret.mat)
```

```
## MSFT FMAGX SBUX
## [1,] 0.127649 0.073069 0.078859
## [2,] 0.054533 0.049066 0.135701
## [3,] 0.006959 0.011513 0.060219
## [4,] -0.060754 -0.045728 -0.002601
## [5,] 0.245154 0.043541 0.107312
## [6,] 0.014314 -0.007508 -0.243824
```

```
summary(ret.mat)
```

```
MSFT
                       FMAGX
                                         SBUX
##
        :-0.4209 Min.
                         :-0.26430 Min.
## Min.
                                          :-0.4797
  1st Qu.:-0.0555 1st Qu.:-0.02293 1st Qu.:-0.0488
##
## Median: 0.0070 Median: 0.00701 Median: 0.0182
## Mean : 0.0048 Mean :-0.00234 Mean : 0.0113
   3rd Qu.: 0.0590
                   3rd Qu.: 0.03625
                                    3rd Qu.: 0.0868
##
   Max. : 0.3420
                   Max. : 0.13024
                                    Max. : 0.2773
##
```

(a). Compute estimates of IID normal model parameters

```
muhat.vals = apply(ret.mat, 2, mean)
muhat.vals
```

```
##
        MSFT
                 FMAGX
                            SBUX
## 0.004846 -0.002342 0.011318
sigma2hat.vals = apply(ret.mat, 2, var)
sigma2hat.vals
##
       MSFT
               FMAGX
                         SBUX
## 0.011206 0.003705 0.014254
sigmahat.vals = apply(ret.mat, 2, sd)
sigmahat.vals
##
     MSFT
             FMAGX
                      SBUX
## 0.10586 0.06086 0.11939
cov.mat = var(ret.mat)
cov.mat
##
             MSFT
                     FMAGX
                               SBUX
## MSFT 0.011206 0.003519 0.004235
## FMAGX 0.003519 0.003705 0.003189
## SBUX 0.004235 0.003189 0.014254
cor.mat = cor(ret.mat)
cor.mat
##
           MSFT FMAGX
                         SBUX
## MSFT 1.0000 0.5461 0.3351
## FMAGX 0.5461 1.0000 0.4388
## SBUX 0.3351 0.4388 1.0000
covhat.vals = cov.mat[lower.tri(cov.mat)]
rhohat.vals = cor.mat[lower.tri(cor.mat)]
names(covhat.vals) <- names(rhohat.vals) <-</pre>
 c("MSFT,FMAGX","MSFT,SBUX","FMAGX,SBUX")
covhat.vals
## MSFT,FMAGX MSFT,SBUX FMAGX,SBUX
     0.003519
              0.004235
##
                           0.003189
rhohat.vals
## MSFT, FMAGX MSFT, SBUX FMAGX, SBUX
```

0.3351

0.4388

0.5461

##

cbind(muhat.vals,sigma2hat.vals,sigmahat.vals)

```
## muhat.vals sigma2hat.vals sigmahat.vals

## MSFT 0.004846 0.011206 0.10586

## FMAGX -0.002342 0.003705 0.06086

## SBUX 0.011318 0.014254 0.11939
```

```
cbind(covhat.vals,rhohat.vals)
```

```
## covhat.vals rhohat.vals
## MSFT,FMAGX 0.003519 0.5461
## MSFT,SBUX 0.004235 0.3351
## FMAGX,SBUX 0.003189 0.4388
```

Microsoft stock is less volatile than Starbucks but more than FMAGX, as shown by its variance. Fidelity is relatively stable with low variance but shows a negative mu value. Starbucks has the highest mu and the highest variance, suggesting room to grow.

(b),(c). Compute standard errors for estimated parameters

estimated standard error for mean

```
nobs = nrow(ret.mat)
nobs
```

```
## [1] 143
```

```
se.muhat = sigmahat.vals/sqrt(nobs)
se.muhat
```

```
## MSFT FMAGX SBUX
## 0.008853 0.005090 0.009984
```

```
cbind(muhat.vals,se.muhat)
```

compute approx 95% confidence intervals

```
mu.lower = muhat.vals - 2*se.muhat
mu.upper = muhat.vals + 2*se.muhat
cbind(mu.lower,mu.upper)
```

```
## mu.lower mu.upper

## MSFT -0.01286 0.022551

## FMAGX -0.01252 0.007838

## SBUX -0.00865 0.031286
```

compute estimated standard errors for variance and sd

```
se.sigma2hat = sigma2hat.vals/sqrt(nobs/2)
se.sigma2hat
```

```
## MSFT FMAGX SBUX
## 0.0013253 0.0004381 0.0016858
```

```
se.sigmahat = sigmahat.vals/sqrt(2*nobs)
se.sigmahat
```

```
## MSFT FMAGX SBUX
## 0.006260 0.003599 0.007060
```

```
cbind(sigma2hat.vals,se.sigma2hat)
```

```
## sigma2hat.vals se.sigma2hat

## MSFT 0.011206 0.0013253

## FMAGX 0.003705 0.0004381

## SBUX 0.014254 0.0016858
```

```
cbind(sigmahat.vals,se.sigmahat)
```

compute approx 95% confidence intervals

```
sigma2.lower = sigma2hat.vals - 2*se.sigma2hat
sigma2.upper = sigma2hat.vals + 2*se.sigma2hat
cbind(sigma2.lower,sigma2.upper)
```

```
## sigma2.lower sigma2.upper

## MSFT 0.008556 0.013857

## FMAGX 0.002828 0.004581

## SBUX 0.010883 0.017626
```

```
sigma.lower = sigmahat.vals - 2*se.sigmahat
sigma.upper = sigmahat.vals + 2*se.sigmahat
cbind(sigma.lower, sigma.upper)
```

```
## sigma.lower sigma.upper

## MSFT 0.09334 0.11838

## FMAGX 0.05367 0.06806

## SBUX 0.10527 0.13351
```

```
# compute estimated standard errors for correlation
se.rhohat = (1-rhohat.vals^2)/sqrt(nobs)
se.rhohat
```

```
## MSFT,FMAGX MSFT,SBUX FMAGX,SBUX
## 0.05869 0.07423 0.06752
```

```
cbind(rhohat.vals,se.rhohat)
```

```
# compute approx 95% confidence intervals
rho.lower = rhohat.vals - 2*se.rhohat
rho.upper = rhohat.vals + 2*se.rhohat
cbind(rho.lower,rho.upper)
```

```
## rho.lower rho.upper

## MSFT,FMAGX 0.4287 0.6635

## MSFT,SBUX 0.1866 0.4836

## FMAGX,SBUX 0.3038 0.5739
```

Since we are using a 95% confidence interval, we can say that we are 95% confident that mu, sigma^2, sigma, and correlation are between the above confidence intervals.

(d). Compute 5% and 1% Value at Risk

```
# function to compute Value-at-Risk
# note: default values are selected for
# the probability level (p) and the initial
# wealth (w). These values can be changed
# when calling the function. Highlight the entire
# function, right click and select run line or selection

Value.at.Risk = function(x,p=0.05,w=100000) {
    x = as.matrix(x)
    q = apply(x, 2, mean) + apply(x, 2, sd)*qnorm(p)
    VaR = (exp(q) - 1)*w
    VaR
}
# 5% and 1% VaR estimates based on W0 = 100000

Value.at.Risk(ret.mat,p=0.05,w=100000)
```

```
## MSFT FMAGX SBUX
## -15573 -9738 -16895
```

```
Value.at.Risk(ret.mat,p=0.01,w=100000)
```

```
## MSFT FMAGX SBUX
## -21449 -13406 -23389
```

Fidelity has the lowest VaR, consistent with its low variance. Microsoft has the median VaR, consistent with is median variance. Starbucks has the highest VaR, consistent with its high variance.