Portfolio Efficient Set

Niraj Sardar

Loading libraries

```
library(quantmod)
## Warning: package 'quantmod' was built under R version 4.1.3
## Loading required package: xts
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
## Loading required package: TTR
## Registered S3 method overwritten by 'quantmod':
     method
     as.zoo.data.frame zoo
##
library(ggplot2)
## Warning in register(): Can't find generic `scale_type` in package ggplot2 to
## register S3 method.
library(dplyr)
## Attaching package: 'dplyr'
## The following objects are masked from 'package:xts':
##
       first, last
##
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
```

```
## 2010-01-28 19:00:00 0.009416196 -0.08334795 -0.01428166 -0.02711248
## 2010-02-25 19:00:00 -0.064676617 0.09207498 -0.02443185 0.18181821
## 2010-03-30 20:00:00 0.093085173 0.10166519 0.05474668 0.06181321
## 2010-04-29 20:00:00 0.111313801 0.08337308 -0.03451132 0.06446740
## 2010-05-27 20:00:00 -0.088122605 -0.10765161 -0.06233915 0.00931746
## 2010-06-29 20:00:00 -0.062424970 -0.01135612 0.02165291 -0.06562314
```

B. Calculate means and covariance matrix

```
meanret <- colMeans(retout,na.rm = T)</pre>
x1 = round(meanret, 5)
cat("The mean vector :\n")
## The mean vector :
print(x1)
       STZ
               CAT
                       LLY
                               CBRL
## 0.03421 0.00581 0.01275 0.01885
covar <- var(retout)</pre>
x2 = round(covar, 8)
cat("The covariance matrix: \n")
## The covariance matrix:
print(x2)
               STZ
                          CAT
                                      LLY
## STZ 0.00745920 0.00105944 0.00058250 0.00119845
## CAT 0.00105944 0.00699707 0.00031562 0.00100841
## LLY 0.00058250 0.00031562 0.00154920 0.00029047
## CBRL 0.00119845 0.00100841 0.00029047 0.00405279
```

C. Creating one portfolio as an example to showcase method

The weight vector contains weights for each security. Weights add to 1.

```
weight <- c(.2, .3, .3, .2)
```

Calculate portfolio variance Using matrix algebra

```
weight <- as.matrix(weight)
dim(weight)</pre>
```

```
## [1] 4 1
 meanret <- as.matrix(meanret)</pre>
 dim(meanret)
 ## [1] 4 1
 mretp <- t(weight) %*% meanret
 sretp <- sqrt(t(weight) %*% covar %*% weight)</pre>
 cat("The mean and sigma of portfolio returns: ", mretp, sretp)
 ## The mean and sigma of portfolio returns: 0.01617978 0.0416561
D. Run simulation with random weights
First, for niter iterations, let's create random portfolio weights.
 set.seed(12)
 niter <- 500 # Set the number of iterations here
 randomnums <- data.frame(replicate(4, runif(niter, 1, 10)))</pre>
 head(randomnums)
 ##
            X1
                     X2
                               Х3
                                        Х4
 ## 1 1.624248 3.921433 1.545892 6.809350
 ## 2 8.359977 4.151757 2.406675 8.817655
 ## 3 9.483596 5.183035 8.274220 3.444403
 ## 4 3.424437 2.660027 5.040517 1.732409
 ## 5 2.524133 5.642371 5.675101 5.134248
 ## 6 1.305061 7.567241 6.264320 3.112291
 wt_sim <- randomnums / rowSums(randomnums)</pre>
 cat("The weights after normalization are in wt_sim...")
 ## The weights after normalization are in wt_sim...
 head(wt_sim)
              X1
                        X2
                                   Х3
```

```
## 1 0.11684463 0.2820988 0.1112079 0.4898488
## 2 0.35220570 0.1749135 0.1013932 0.3714877
## 3 0.35942787 0.1964368 0.3135926 0.1305427
## 4 0.26633999 0.2068870 0.3920327 0.1347403
## 5 0.13301817 0.2973448 0.2990696 0.2705674
## 6 0.07151443 0.4146680 0.3432709 0.1705466
```

Initializing Variables.

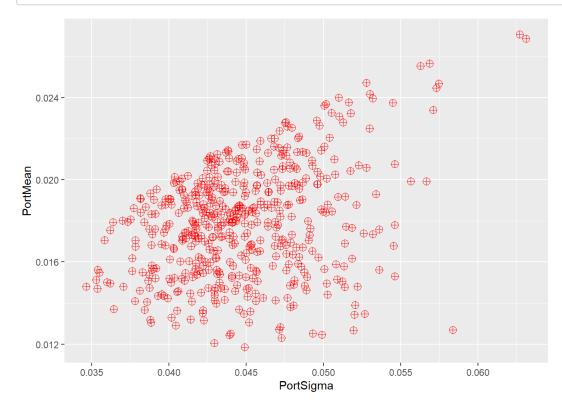
```
# initialize weight and Results matrices
weight <- matrix(data = NA, nrow = length(ticks), ncol = 1)</pre>
Results <- matrix(data = NA, nrow = niter, ncol = 6)</pre>
```

Run the simulations - this means, do portfolio calculations for each simulated portfolio.

```
## STZ CAT LLY CBRL PortMean PortSigma
## 1 0.11684463 0.2820988 0.1112079 0.4898488 0.01628776 0.04693199
## 2 0.35220570 0.1749135 0.1013932 0.3714877 0.02136079 0.04862590
## 3 0.35942787 0.1964368 0.3135926 0.1305427 0.01989614 0.04430372
## 4 0.26633999 0.2068870 0.3920327 0.1347403 0.01785103 0.04003521
## 5 0.13301817 0.2973448 0.2990696 0.2705674 0.01519099 0.04083742
## 6 0.07151443 0.4146680 0.3432709 0.1705466 0.01244703 0.04391907
```

Plotting the results.

```
ggplot(data = Results , aes(x = PortSigma, y = PortMean)) +
geom_point(pch = 10, colour = "red", size = 3)
```



E. Optimization

Again, risk is bad while return is good, so one way to select the best portfolio(s) is to identify a constraint based on one of these dimensions, and locate the best portfolio using the other.

```
## # A tibble: 17 x 6
##
      ints
                     lowerval sig_optim retn_optim
                                                     numb portID
##
      <fct>
                        <dbl>
                                   <dbl>
                                              <dbl>
                                                     <int>
                                                            <int>
                        0.011
    1 (0.011,0.012]
                                  0.0449
                                             0.0119
##
                                                         1
                                                              446
    2 (0.012,0.013]
                                  0.0405
                                             0.0129
                                                              407
##
                        0.012
                                                        11
    3 (0.013,0.014]
                        0.013
                                  0.0364
                                             0.0137
                                                        20
                                                              315
    4 (0.014,0.015]
                        0.014
                                  0.0347
                                             0.0148
                                                        45
                                                              231
    5 (0.015,0.016]
                        0.015
                                             0.0151
##
                                  0.0353
                                                        57
                                                              242
                        0.016
                                             0.0162
                                                              352
##
    6 (0.016,0.017]
                                  0.0376
                                                        48
                                             0.0171
    7 (0.017,0.018]
                        0.017
                                  0.0358
                                                        76
                                                              282
    8 (0.018, 0.019]
                        0.018
                                  0.0370
                                             0.0180
                                                              345
   9 (0.019,0.02]
                        0.019
                                             0.0191
##
                                  0.0382
                                                        68
                                                              403
## 10 (0.02,0.021]
                        0.02
                                             0.0201
                                  0.0404
                                                        42
                                                              218
## 11 (0.021,0.022]
                        0.021
                                  0.0426
                                             0.0211
                                                        28
                                                              485
## 12 (0.022,0.023]
                        0.022
                                  0.0466
                                             0.0220
                                                        14
                                                               76
## 13 (0.023,0.024]
                        0.023
                                  0.0500
                                             0.0236
                                                        10
                                                              139
## 14 (0.024,0.025]
                        0.024
                                  0.0528
                                             0.0247
                                                         4
                                                              495
## 15 (0.025,0.026]
                        0.025
                                  0.0563
                                             0.0255
                                                         2
                                                               36
## 16 (0.026,0.027]
                        0.026
                                  0.0631
                                             0.0269
                                                         1
                                                              198
## 17 (0.027,0.028]
                        0.027
                                  0.0627
                                             0.0271
                                                              120
```

```
ggplot(data = optim , aes(x = sig_optim, y = retn_optim)) +
  geom_point(pch = 10, colour = "red", size = 3)
```

