# Stats\_C183\_Project\_5\_Charles\_Liu

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### Loading Necessary Packages/Files:

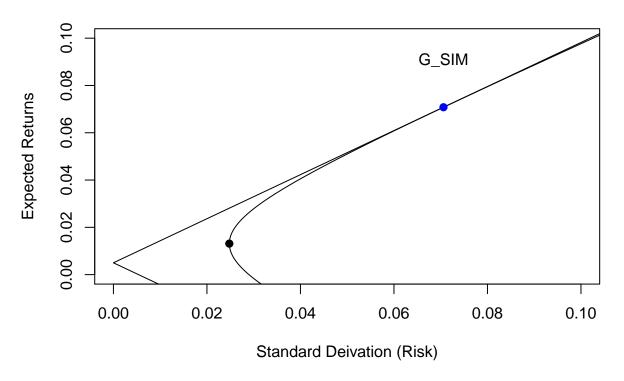
library(readr)

 $E_SIM \leftarrow seq(-5,5,.1)$ 

```
# Loading the data for all stocks:
a <- read.csv("C:/Users/cliuk/Documents/UCLA Works/UCLA Spring 2020/Stats C183/Project/stockData_all.cs
a)
# Convert adjusted close prices into returns:
r \leftarrow (a[-1,3:ncol(a)]-a[-nrow(a),3:ncol(a)])/a[-nrow(a),3:ncol(a)]
#Compute the betas:
means <- colMeans(r)
covmat <- var(r)</pre>
beta \leftarrow covmat[1,-1] / covmat[1,1]
beta
##
          BHP
                     GOLD
                                 VALE
                                             GOOG
                                                                     NFLX
##
   1.03167339 -0.14116913
                           0.96606144
                                      0.99316900
                                                   0.63252260
                                                               1.15829301
##
         AM7.N
                      MCD
                                 TSLA
                                              WMT
                                                           ΚO
                                                                     COST
   1.20704704
              0.60859600
                           1.00025367
                                       0.25604736
                                                   0.66946453
                                                            V
##
          MOX
                      CVX
                                  TRP
                                            BRK.B
                                                                      JPM
##
   1.22485080
               1.17242167
                           0.84150937
                                       0.86005529
                                                   0.95455176
                                                               1.41838539
                                                                       GE
##
           JNJ
                     AMGN
                                              UNP
                                                           BA
                                  CVS
   0.71108254
               0.82660026
                           0.81853689
                                       0.91680193
                                                   1.53779231
                                                               1.33457006
##
##
          DLR
                      BXP
                                    0
                                             MSFT
                                                         AAPL
                                                                     NVDA
   0.07297199 0.97358604 0.57217761 0.91629371
                                                   1.14510667
                                                               1.36654407
# For Beta, only GOLD is negative, so we exclude it in our SIM (Single Index Model)
means_SIM <- colMeans(r[,-c(1,3)]) # 29 stocks instead of 31 (removes GOLD, S&P500)
covmat_SIM <- var(r[,-c(1,3)]) # Removed GOLD as its Beta is NOT positive.
# Set up A - E formulas & column of Ones:
ones_SIM \leftarrow rep(1, 29)
A_SIM <- t(ones_SIM) %*% solve(covmat_SIM) %*% means_SIM
B_SIM <- t(means_SIM) %*% solve(covmat_SIM) %*% means_SIM
C_SIM <- t(ones_SIM) %*% solve(covmat_SIM) %*% ones_SIM
D_SIM <- B_SIM * C_SIM - A_SIM^2
```

```
# Set up the plot:
plot(sqrt(1/C_SIM), A_SIM/C_SIM, xlim=c(0,0.1), ylim=c(0,0.1), pch=19, ylab = "Expected Returns", xlab =
minvar_SIM <- 1/C_SIM</pre>
minE_SIM <- A_SIM/C_SIM
sdeff_SIM \leftarrow seq((minvar_SIM)^0.5, 1, by = 0.0001)
options(warn = -1)
y1_SIM <- (A_SIM + sqrt(D_SIM*(C_SIM*sdeff_SIM^2 - 1)))*(1/C_SIM)</pre>
y2 SIM <- (A SIM - sqrt(D SIM*(C SIM*sdeff SIM^2 - 1)))*(1/C SIM)
options(warn = 0)
points(sdeff_SIM, y1_SIM, type = "1")
points(sdeff_SIM, y2_SIM, type = "1")
# Choose risk-free return:
Rf < -0.005
# Range of expected return:
sigma \leftarrow seq(0,.5,.001)
Rp1_SIM <- Rf + sigma*sqrt(C_SIM*Rf^2-2*Rf*A_SIM+B_SIM)</pre>
Rp2_SIM <- Rf - sigma*sqrt(C_SIM*Rf^2-2*Rf*A_SIM+B_SIM)</pre>
points(sigma, Rp1_SIM, type="1")
points(sigma, Rp2_SIM, type="1")
# Point of tangency:
Ri_SIM <- means_SIM-Rf</pre>
z_SIM <- solve(covmat_SIM) %*% Ri_SIM
x_G_SIM <- z_SIM/sum(z_SIM)</pre>
Ri_bar_SIM <- t(x_G_SIM) %*% means_SIM</pre>
varr_G_SIM <- t(x_G_SIM) %*% covmat_SIM %*% x_G_SIM</pre>
sdev_G_SIM <- varr_G_SIM^.5</pre>
points(sdev_G_SIM, Ri_bar_SIM, pch=19, col = "blue")
text(sdev_G_SIM, Ri_bar_SIM + 0.02, "G_SIM")
```

# Stocks w/ Positive Betas (SIM)



#### # Composition of Optimal Portfolio:

 $x_G_SIM$ 

```
##
                 [,1]
## BHP
         -0.33723772
## VALE
         -0.03448230
## GOOG
         -0.22340547
## T
          0.25570333
## NFLX
          0.02755105
          0.31380130
## AMZN
## MCD
          0.25978438
## TSLA
          0.12209782
## WMT
          0.15861048
## KO
         -0.59357945
## COST
         -0.32149018
         -1.81623157
## XOM
## CVX
          0.94854647
## TRP
          0.15432028
## BRK.B -0.52345533
## V
          0.82362571
## JPM
          0.48436237
## JNJ
          0.50293396
## AMGN
         -0.21516512
## CVS
         -0.10995796
## UNP
          1.16474227
## BA
          0.18991178
```

```
## GE
          -0.12012387
## DLR
          0.15905596
        -1.57521453
## BXP
## 0
          0.98674874
## MSFT -0.03961945
## AAPL
         0.19703010
## NVDA
         0.16113696
Ri_bar_SIM
               [,1]
## [1,] 0.07079141
sdev_G_SIM
                [,1]
## [1,] 0.07059549
b)
# Keep only the stocks with positive betas:
rrr <- r[,-c(1,which(beta<0)+1)]</pre>
# Initialize
beta <- rep(0,ncol(rrr))</pre>
alpha <- rep(0,ncol(rrr))
mse <- rep(0,ncol(rrr))</pre>
Ribar <- rep(0,ncol(rrr))</pre>
Ratio <- rep(0,ncol(rrr))</pre>
stock <- rep(0,ncol(rrr))</pre>
# Risk free asset:
rf < -0.005
# This for loop computes the required inputs:
for(i in 1:ncol(rrr)){
  q <- lm(data=rrr, formula=rrr[,i] ~ r[,1])</pre>
  beta[i] <- q$coefficients[2]</pre>
  alpha[i] <- q$coefficients[1]</pre>
  mse[i] <- summary(q)$sigma^2</pre>
  Ribar[i] <- q$coefficients[1]+q$coefficients[2]*mean(r[,1])</pre>
  Ratio[i] <- (Ribar[i]-rf)/beta[i]</pre>
  stock[i] <- i
# So far we have this table:
xx <- (cbind(stock,alpha, beta, Ribar, mse, Ratio))</pre>
# Order the table based on the excess return to beta ratio:
A_ordered <- xx[order(-xx[,6]),]
col1 <- rep(0,nrow(A_ordered))</pre>
col2 <- rep(0,nrow(A_ordered))</pre>
col3 <- rep(0,nrow(A_ordered))</pre>
```

col4 <- rep(0,nrow(A\_ordered))
col5 <- rep(0,nrow(A\_ordered))</pre>

```
# Create the last 5 columns of the table:
col1 <- (A_ordered[,4]-rf)*A_ordered[,3]/A_ordered[,5]</pre>
col3 <- A_ordered[,3]^2/A_ordered[,5]</pre>
for(i in(1:nrow(covmat))) {
  col2[i] <- sum(col1[1:i])
  col4[i] <- sum(col3[1:i])
}
# Compute the Ci (col5):
for(i in (1:nrow(A_ordered))) {
  col5[i] \leftarrow var(r[,1])*col2[i]/(1+var(r[,1])*col4[i])
}
table1 <- cbind(A_ordered, col1, col2, col3, col4, col5)
# Express the entire table:
table1
##
         stock
                       alpha
                                   beta
                                                Ribar
                                                                            Ratio
##
   [1,]
            24 0.0118107716 0.07297199 0.0123631195 0.0036757586
                                                                     0.1009033728
##
  [2,]
               0.0340296617 1.00025367 0.0416008955 0.0260069622
                                                                     0.0365916135
  [3,]
             5 0.0331898847 1.15829301 0.0419573679 0.0213044632
                                                                     0.0319067522
## [4,]
            29 0.0258610181 1.36654407 0.0362048189 0.0091800225
                                                                     0.0228348428
## [5,]
              6 \quad 0.0176942699 \ 1.20704704 \quad 0.0268307877 \ 0.0045011394 \quad 0.0180861118 \\
## [6,]
             9 0.0076468285 0.25604736 0.0095849313 0.0024518651 0.0179065753
## [7,]
            16 0.0145218598 0.95455176 0.0217471616 0.0013347556 0.0175445295
##
   [8,]
            27 0.0139147478 0.91629371 0.0208504624 0.0022388395
                                                                     0.0172984516
## [9,]
            11 0.0106703983 0.73848356 0.0162602120 0.0016088686
                                                                     0.0152477491
## [10,]
            19 0.0089435868 0.82660026 0.0152003835 0.0031739754
                                                                     0.0123401649
## [11,]
            28 0.0096519746 1.14510667 0.0183196462 0.0041941589
                                                                     0.0116317952
## [12,]
            3 0.0084803876 0.99316900 0.0159979953 0.0022926955
                                                                     0.0110736394
## [13,]
            21 0.0061436706 0.91680193 0.0130832320 0.0019932995
                                                                     0.0088167703
## [14,]
            26 0.0050196653 0.57217761 0.0093506571 0.0040877734
                                                                     0.0076036829
## [15,]
            18 0.0049620446 0.71108254 0.0103444515 0.0011132114
                                                                     0.0075159369
## [16.]
            17 0.0030707954 1.41838539 0.0138069995 0.0022783396
                                                                     0.0062091724
## [17,]
            7 0.0041343280 0.60859600 0.0087409821 0.0013987540 0.0061469055
## [18,]
            22 0.0017475356 1.53779231 0.0133875680 0.0044395254
                                                                    0.0054542918
## [19,]
            15 0.0029891655 0.86005529 0.0094991938 0.0007351871
                                                                     0.0052312844
            20 0.0013003125 0.81853689 0.0074960750 0.0034275847
## [20,]
                                                                     0.0030494350
## [21,]
            14 0.0006211814 0.84150937 0.0069908298 0.0021959108 0.0023657845
## [22,]
            10 0.0013410007 0.66946453 0.0064083877 0.0013278936
                                                                     0.0021037525
## [23,]
            4 0.0008445761 0.63252260 0.0056323382 0.0019083649
                                                                     0.0009997084
## [24,]
            25 -0.0049149844 0.97358604 0.0024543939 0.0021197381 -0.0026146699
## [25,]
            13 \ -0.0074148364 \ 1.17242167 \ \ 0.0014595910 \ \ 0.0019192874 \ \ -0.0030197403
## [26,]
            1 \ -0.0084386036 \ 1.03167339 \ -0.0006295440 \ 0.0056509671 \ -0.0054567115
## [27,]
            23 -0.0124704813 1.33457006 -0.0023687018 0.0044047254 -0.0055214050
## [28,]
             2 -0.0077152214 0.96606144 -0.0004027993 0.0166931433 -0.0055926042
## [29,]
            12 -0.0126171140 1.22485080 -0.0033458340 0.0015371372 -0.0068137557
##
                          col2
               col1
                                      col3
                                                   col4
                                                                 col5
##
  [1,] 0.1461743 0.1461743
                                  1.448656
                                               1.448656 0.0001823629
##
  [2,]
                    1.5538812
                                 38.470752
                                              39.919408 0.0018497973
         1.4077069
  [3,]
         2.0093189
                     3.5632001
                                 62.974724
                                             102.894132 0.0039459469
##
  [4,] 4.6451695 8.2083696 203.424631
                                             306.318763 0.0074188068
```

```
[5,] 5.8542483 14.0626178 323.687499
                                           630.006262 0.0098332099
   [6,] 0.4788027 14.5414205
                                           656.745193 0.0099813875
##
                               26.738931
  [7,] 11.9767488 26.5181693
                              682.648616 1339.393809 0.0123945509
## [8,] 6.4871462 33.0053156
                              375.013117
                                          1714.406926 0.0131259153
   [9,] 5.1685274 38.1738430
                              338.969861
                                          2053.376787 0.0133779712
## [10,] 2.6564918 40.8303348 215.271986 2268.648773 0.0133051695
## [11,] 3.6365851 44.4669199
                              312.641775 2581.290549 0.0131504505
## [12,] 4.7642035 49.2311234 430.229245 3011.519794 0.0129160350
## [13,] 3.7178170 52.9489404
                              421.675616
                                          3433.195409 0.0125077109
## [14,] 0.6089742 53.5579146
                               80.089375 3513.284784 0.0124166548
## [15,] 3.4138586 56.9717732 454.215976 3967.500760 0.0119497574
## [16,] 5.4828171 62.4545903
                              883.018975 4850.519736 0.0110526810
## [17,]
        1.6276963 64.0822866
                              264.799309 5115.319044 0.0108330779
## [18,] 2.9053416 66.9876282
                              532.670721 5647.989765 0.0103887420
## [19,] 5.2633615 72.2509897 1006.131786 6654.121551 0.0096926161
## [20,] 0.5960843 72.8470741
                              195.473695 6849.595246 0.0095228622
## [21,] 0.7629189 73.6099930
                              322.480323 7172.075569 0.0092333535
## [22,] 0.7100461 74.3200391 337.514060 7509.589629 0.0089437713
## [23,] 0.2095869 74.5296259 209.647982 7719.237611 0.0087482802
## [24,] -1.1691853 73.3604406 447.163630 8166.401241 0.0081816050
## [25,] -2.1627049 71.1977357 716.189031 8882.590272 0.0073530880
## [26,] -1.0277623 70.1699735 188.348289 9070.938561 0.0071086657
## [27,] -2.2326134 67.9373601 404.356026 9475.294587 0.0066116492
## [28,] -0.3126695 67.6246906
                               55.907667 9531.202254 0.0065456061
## [29,] -6.6502858 60.9744048 976.008846 10507.211100 0.0053924721
```

### $\mathbf{c})$

0.031211982

CVX

##

0.114097428

TRP

```
# SHORT SALES ALLOWED:
z_short <- (A_ordered[,3]/A_ordered[,5])*(A_ordered[,6]-col5[nrow(A_ordered)])</pre>
x_short <- z_short/sum(z_short)</pre>
x_short <- x_short[order(A_ordered[,1])]</pre>
names(x_short) \leftarrow names(a)[-c(1,2,3,5)]
Rbar_short <- t(x_short) %*% Ribar</pre>
sdev_short <- sqrt(t(x_short) %*% covmat_SIM %*% x_short)</pre>
#SHORT SALES NOT ALLOWED:
table1 <- cbind(A_ordered, col1, col2, col3, col4, col5)
table2 <- table1[1:which(col5==max(col5)), ]</pre>
z_{no\_short} \leftarrow (table2[,3]/table2[,5])*(table2[,6]-max(col5))
x_no_short <- z_no_short/sum(z_no_short)</pre>
names(x_no_short) \leftarrow names(means_SIM)[c(24,8,5,29,6,9,16,27,11)]
Rbar no short \leftarrow t(x no short) %*% Ribar[c(24,8,5,29,6,9,16,27,11)]
sdev_no_short \leftarrow sqrt(t(x_no_short) %*% covmat_SIM[c(24,8,5,29,6,9,16,27,11), c(24,8,5,29,6,9,16,27,11)]
# Composition of Point of Tangency (Short Sales Allowed):
x_short
                                                                                    AMZN
             BHP
                                         GOOG
                                                           Τ
                                                                      NFLX
##
                          VALE
   -0.188334201 -0.060448063
                                 0.234006026 -0.138441138
                                                             0.137069353
                                                                            0.323668611
##
             MCD
                          TSLA
                                          WMT
                                                         ΚO
                                                                      COST
                                                                                     XUM
```

V

0.430132025 -0.924835985

JNJ

JPM

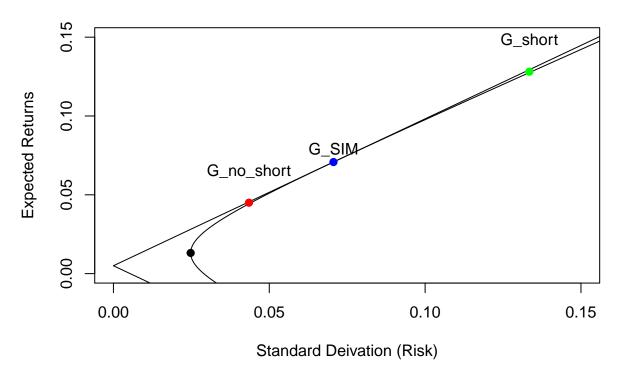
0.124261375 -0.157653576

BRK.B

```
## -0.488614940 -0.110287057 -0.017929700 0.826342765 0.048344961 0.128973398
##
                         CVS
                                       UNP
                                                     ΒA
                                                                               DI.R.
           AMGN
                                                                   GF.
## 0.172046276 -0.053203709 0.149757036 0.002036109 -0.314422985 0.180291319
            BXP
                                      MSFT
##
                           0
                                                   AAPL
                                                                 NVDA
## -0.349689307 0.029429771 0.463329090 0.161976207 0.246886927
# Composition of Point of Tangency (Short Sales NOT Allowed):
x_no_short
                                           NVDA
##
          DLR
                    TSLA
                                NFLX
                                                       AMZN
                                                                   WMT
## 0.14215012 0.07304117 0.08241355 0.11516787 0.10328924 0.03868940 0.24376936
                    COST
         MSFT
## 0.13126682 0.07021246
# Previous Project Plot:
plot(sqrt(1/C SIM), A SIM/C SIM, xlim=c(0,0.15), ylim=c(0,0.15),pch=19, ylab = "Expected Returns", xla
minvar_SIM <- 1/C_SIM</pre>
minE_SIM <- A_SIM/C_SIM
sdeff_SIM \leftarrow seq((minvar_SIM)^0.5, 1, by = 0.0001)
options(warn = -1)
Rbar_short <- 0.1281
Rbar_no_short <- 0.045
y1_SIM <- (A_SIM + sqrt(D_SIM*(C_SIM*sdeff_SIM^2 - 1)))*(1/C_SIM)
y2_SIM <- (A_SIM - sqrt(D_SIM*(C_SIM*sdeff_SIM^2 - 1)))*(1/C_SIM)
options(warn = 0)
points(sdeff_SIM, y1_SIM, type = "1")
points(sdeff_SIM, y2_SIM, type = "1")
points(sigma, Rp1_SIM, type="1")
points(sigma, Rp2_SIM, type="l")
Ri_SIM <- means_SIM-Rf</pre>
z SIM <- solve(covmat SIM) %*% Ri SIM
x_G_SIM <- z_SIM/sum(z_SIM)</pre>
Ri_bar_SIM <- t(x_G_SIM) %*% means_SIM
varr_G_SIM <- t(x_G_SIM) %*% covmat_SIM %*% x_G_SIM</pre>
sdev_G_SIM <- varr_G_SIM^.5</pre>
points(sdev_G_SIM, Ri_bar_SIM, pch=19, col = "blue")
text(sdev_G_SIM, Ri_bar_SIM + 0.008, "G_SIM")
# Plot it with Short Sales' and No Short Sales' Tangency Point:
points(sdev_short, Rbar_short, pch=19, col = "green") # Short Sales Allowed
text(sdev_short, Rbar_short + 0.02, "G_short")
points(sdev_no_short, Rbar_no_short, pch=19, col = "red") # Short Sales NOT Allowed
```

text(sdev\_no\_short, Rbar\_no\_short + 0.02, "G\_no\_short")

## Stocks w/ Positive Betas (SIM)

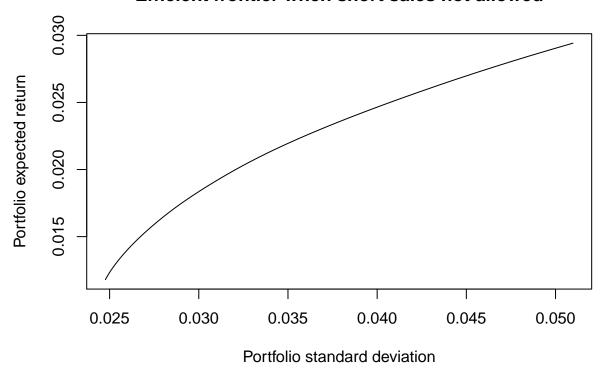


d)

```
# Initialize for the for() loop:
Rfr \leftarrow seq(-0.05,.01,0.0005)
rbar_opt <- rep(0,length(Rfr))</pre>
risk_opt <- rep(0,length(Rfr))
# Start the for() loop:
for(l in 1:length(Rfr)){
  rf <- Rfr[1]
  beta <- rep(0,ncol(rrr))</pre>
  alpha <- rep(0,ncol(rrr))</pre>
  mse <- rep(0,ncol(rrr))</pre>
  Ribar <- rep(0,ncol(rrr))</pre>
  Ratio <- rep(0,ncol(rrr))</pre>
  stocknum <- rep(0,ncol(rrr))</pre>
  for(i in 1:ncol(rrr)){
    q <- lm(data=rrr, formula=rrr[,i] ~ r[,1])</pre>
    beta[i] <- q$coefficients[2]</pre>
    alpha[i] <- q$coefficients[1]</pre>
    mse[i] <- summary(q)$sigma^2</pre>
    Ribar[i] <- q$coefficients[1]+q$coefficients[2]*mean(r[,1])</pre>
    Ratio[i] <- (Ribar[i]-rf)/beta[i]</pre>
```

```
stocknum[i] <- i
  }
  xx <- (data.frame(stocknum,alpha, beta, Ribar, mse, Ratio))
  A \leftarrow xx[order(-xx[,6]),]
  col1 <- rep(0,nrow(A))</pre>
  col2 <- rep(0,nrow(A))</pre>
  col3 <- rep(0,nrow(A))</pre>
  col4 \leftarrow rep(0,nrow(A))
  col5 \leftarrow rep(0,nrow(A))
  col1 \leftarrow (A[,4]-rf)*A[,3]/A[,5]
  col3 <- A[,3]^2/A[,5]
  for(i in(1:nrow(A))) {
    col2[i] <- sum(col1[1:i])</pre>
    col4[i] <- sum(col3[1:i])
  cbind(A, col1, col2, col3, col4)
  for(i in (1:nrow(A))) {
    col5[i] \leftarrow var(r[,1])*col2[i]/(1+var(r[,1])*col4[i])
  B <- cbind(A, col1, col2, col3, col4, col5)
  rownames(B) <- NULL
  table2 <- B[1:which(col5==max(col5)), ]
  z_{no\_short} \leftarrow (table2[,3]/table2[,5])*(table2[,6]-max(col5))
  x_no_short <- z_no_short/sum(z_no_short)</pre>
  r1 <- data.frame(rrr[,table2[,1]])
  beta1 <- rep(0,ncol(r1))</pre>
  sigma_e1 \leftarrow rep(0,ncol(r1))
  alpha1 <- rep(0,ncol(r1))
  for(i in 1:ncol(r1)){
    q1<- lm(r1[,i] ~ r[,1])
    beta1[i] <- q1$coefficients[2]</pre>
    sigma_e1[i] <- summary(q1)$sigma^2
    alpha1[i] <- q1$coefficients[1]</pre>
  means1 <- colMeans(r1)</pre>
  xx \leftarrow rep(0,ncol(r1)*(ncol(r1)))
  varcovar <- matrix(xx,nrow=ncol(r1),ncol=ncol(r1))</pre>
  for (i in 1:ncol(r1)){
    for (j in 1:ncol(r1)){
      varcovar[i,j]=beta1[i]*beta1[j]*var(r[,1])
      if(i==j){varcovar[i,j]=beta1[i]^2*var(r[,1])+ sigma_e1[i]}
    }
  }
  rbar_opt[1] <- t(x_no_short) %*% means1</pre>
  risk_opt[1] <- ( t(x_no_short) %*% varcovar %*% x_no_short )^.5
# Create the Plot for Efficient Frontier When Short Sales NOT Allowed:
plot(risk_opt, rbar_opt, type="l", main="Efficient frontier when short sales not allowed", ylab="Portfo
```

### Efficient frontier when short sales not allowed



**e**)

```
# Convert adjusted close prices into returns:
r \leftarrow (a[-1, 3:ncol(a)] - a[-nrow(a), 3:ncol(a)]) / a[-nrow(a), 3:ncol(a)]
# Compute inputs:
stock_CCM <- 1:30
Ribar_CCM <- colMeans(r)</pre>
diff_CCM <- Ribar_CCM - Rf</pre>
covmat_CCM <- cov(r)</pre>
stdev_CCM <- diag(covmat)^.5</pre>
ratio_CCM <- diff_CCM / stdev_CCM
cormat_CCM <- cor(r)</pre>
avg_cor_CCM <- (sum(cormat_CCM) - 30) / (30*29)</pre>
# Table Construction
xx_CCM <- (cbind(stock_CCM, Ribar_CCM, diff_CCM, stdev_CCM, ratio_CCM))</pre>
A_CCM <- xx_CCM[order(-xx_CCM[,5]),]
col1 <- rep(0, nrow(A_CCM))</pre>
col2 <- rep(0, nrow(A_CCM))</pre>
col3 <- rep(0, nrow(A_CCM))</pre>
col1 <- avg_cor_CCM / (1 - avg_cor_CCM + stock_CCM*avg_cor_CCM)</pre>
for(i in(1:nrow(A CCM))){
col2[i] <- sum(A_CCM[1:i, 5])</pre>
}
```

```
col3 <- col1 * col2 # Ci
table3 <- cbind(A_CCM, col1, col2, col3)
# Express the Table:
table3
##
          stock_CCM
                        Ribar_CCM
                                        diff_CCM stdev_CCM
                                                               ratio_CCM
                                                                                col1
## V
                 18
                     0.0217471616
                                    0.0167471616 0.04959639
                                                              0.33766894 0.26907667
                                    0.0312048189 0.10686145
## NVDA
                  1
                     0.0362048189
                                                              0.29201193 0.21202554
## MSFT
                 29
                     0.0208504624
                                    0.0158504624 0.05714113
                                                              0.27739146 0.17493488
                                    0.0218307877 0.07921927
## AMZN
                     0.0268307877
                                                              0.27557422 0.14888900
                  8
## NFLX
                  7
                     0.0419573679
                                    0.0369573679 0.15086966
                                                              0.24496222 0.12959389
## COST
                     0.0162602120
                                    0.0112602120 0.04768530
                                                              0.23613594 0.11472609
                 13
## TSLA
                     0.0416008955
                                    0.0366008955 0.16428425
                                                              0.22279005 0.10291864
## GOOG
                  5
                     0.0159979953
                                    0.0109979953 0.05917663
                                                              0.18585032 0.09331481
## AAPL
                 30
                     0.0183196462
                                    0.0133196462 0.07609063
                                                              0.17504975 0.08535036
## AMGN
                     0.0152003835
                                    0.0102003835 0.06320778
                                                              0.16137861 0.07863853
## UNP
                 23
                     0.0130832320
                                    0.0080832320 0.05498422
                                                              0.14701002 0.07290536
## JNJ
                 20
                     0.0103444515
                                    0.0053444515 0.04163769
                                                              0.12835610 0.06795134
## JPM
                 19
                     0.0138069995
                                    0.0088069995 0.06905999
                                                              0.12752680 0.06362775
                     0.0123631195
                                    0.0073631195 0.06036985
## DLR
                 26
                                                              0.12196682 0.05982145
## BRK.B
                     0.0094991938
                                    0.0044991938 0.04064605
                                                              0.11069203 0.05644483
                 17
## BA
                 24
                     0.0133875680
                                    0.0083875680 0.08572843
                                                              0.09783882 0.05342904
## WMT
                     0.0095849313
                                    0.0045849313 0.05008520
                                                              0.09154264 0.05071917
                 11
## MCD
                  9
                     0.0087409821
                                   0.0037409821 0.04297973
                                                              0.08704061 0.04827091
## X.GSPC
                     0.0075693137
                                    0.0025693137 0.03535294
                                                              0.07267609 0.04604812
                  1
## 0
                 28
                     0.0093506571
                                    0.0043506571 0.06674436
                                                              0.06518389 0.04402103
## CVS
                 22
                     0.0074960750
                                    0.0024960750 0.06503567
                                                              0.03838009 0.04216489
## TRP
                     0.0069908298
                                    0.0019908298 0.05530213
                                                              0.03599916 0.04045895
                                    0.0014083877 0.04329384
## KO
                                                              0.03253090 0.03888567
                 12
                     0.0064083877
## T
                  6
                     0.0056323382
                                    0.0006323382 0.04887463
                                                              0.01293796 0.03743017
                  3 -0.0001718661 -0.0051718661 0.13304506 -0.03887304 0.03607970
## GOLD
                  4 -0.0004027993 -0.0054027993 0.13299430 -0.04062429 0.03482329
## VALE
## BXP
                 27
                     0.0024543939 -0.0025456061 0.05729365 -0.04443086 0.03365144
## CVX
                     0.0014595910 -0.0035404090 0.06014552 -0.05886405 0.03255588
                 15
                  2 -0.0006295440 -0.0056295440 0.08320437 -0.06765924 0.03152942
## BHP
                 25 -0.0023687018 -0.0073687018 0.08115024 -0.09080321 0.03056570
## GE
                 14 -0.0033458340 -0.0083458340 0.05827830 -0.14320654 0.26907667
## XOM
##
               co12
                           co13
## V
          0.3376689 0.09085883
          0.6296809 0.13350843
## NVDA
## MSFT
          0.9070723 0.15867859
## AMZN
          1.1826465 0.17608306
## NFLX
          1.4276088 0.18500938
## COST
          1.6637447 0.19087493
## TSLA
          1.8865348 0.19415959
## GOOG
          2.0723851 0.19338421
## AAPL
          2.2474348 0.19181936
## AMGN
          2.4088134 0.18942554
## UNP
          2.5558235 0.18633323
## JNJ
          2.6841796 0.18239360
## JPM
          2.8117064 0.17890255
## DLR
          2.9336732 0.17549657
```

```
## BRK.B 3.0443652 0.17183869
## BA
          3.1422040 0.16788495
## WMT
          3.2337467 0.16401293
## MCD
          3.3207873 0.16029741
## X.GSPC 3.3934634 0.15626261
          3.4586473 0.15225323
## N
## CVS
          3.4970273 0.14745178
          3.5330265 0.14294253
## TRP
## KO
          3.5655574 0.13864910
## T
          3.5784954 0.13394371
## GOLD
          3.5396223 0.12770853
## VALE
          3.4989980 0.12184663
## BXP
          3.4545672 0.11625115
## CVX
          3.3957031 0.11055012
## BHP
          3.3280439 0.10493128
## GE
          3.2372407 0.09894852
## XOM
          3.0940341 0.83253240
f)
# SHORT SALES ALLOWED:
z_short_CCM <- (A_CCM[,5] - col3[nrow(A_CCM)]) / ((1 - avg_cor_CCM)*A_CCM[,4])</pre>
x_short_CCM <- z_short_CCM/sum(z_short_CCM)</pre>
names(x_short_CCM) <- names(a)[-c(1,2)]</pre>
# SHORT SALES NOT ALLOWED:
table3 <- cbind(A_CCM, col1, col2, col3)
table4 <- table3[1:which(col3 == max(col3)), ]</pre>
z_no_short_CCM <- (table4[,5] - col3[nrow(table4)]) / ((1 - avg_cor_CCM)*table4[,4])</pre>
x_no_short_CCM <- z_no_short_CCM/sum(z_no_short_CCM)</pre>
x_no_short_CCM <- x_no_short_CCM[order(A_CCM[,1])]</pre>
x_no_short_CCM <- ifelse(is.na(x_no_short_CCM), 0, x_no_short_CCM)</pre>
names(x_no_short_CCM) <- names(a)[-c(1,2)]</pre>
# Composition of short sales and no short sales:
x_short_CCM
##
       X.GSPC
                                 GOLD
                                            VALE
                                                        GOOG
                                                                       Τ
                                                                               NFLX
## 0.02704427 0.01370980 0.02633264 0.01905600 0.01055596 0.03389929 0.01005983
                      MCD
                                TSLA
                                             WMT
                                                          ΚO
                                                                   COST
                                                                                MOX
## 0.02961970 0.02342034 0.02878005 0.03379275 0.04583896 0.02766980 0.03190245
          CVX
                      TRP
                               BRK.B
                                               V
                                                         JPM
                                                                     JNJ
                                                                               AMGN
## 0.04813521 0.02322851 0.04009986 0.04701313 0.05825671 0.03116148 0.03309727
          CVS
## 0.03903931 0.05008466 0.04545227 0.01775258 0.01779505 0.04148728 0.04017054
## 0.02932438 0.03083968 0.04538024
x_no_short_CCM
##
       X.GSPC
                      BHP
                                 GOLD
                                            VALE
                                                        GOOG
                                                                               NFI.X
```

## 0.01370980 0.05825671 0.02932438 0.01775258 0.01779505 0.02961970 0.04545227

```
##
         AMZN
                      MCD
                                TSLA
                                             WMT
                                                          ΚO
                                                                   COST
                                                                                MOX
## 0.01055596 0.01905600 0.04701313 0.01005983 0.04009986 0.05008466 0.03389929
                               BRK.B
                      TRP
                                               V
                                                         JPM
                                                                     JNJ
                                                                               AMGN
## 0.04538024 0.04017054 0.03903931 0.04813521 0.02704427 0.02766980 0.04583896
          CVS
                      UNP
                                  BA
                                              GE
                                                         DLR
                                                                    BXP
## 0.02878005 0.03309727 0.03379275 0.02322851 0.03083968 0.03190245 0.04148728
                     AAPL
## 0.03116148 0.02633264 0.02342034
# Plot it all:
mean_short <- t(x_short_CCM) %*% means</pre>
sd_short <- sqrt(t(x_short_CCM) %*% covmat %*% x_short_CCM)</pre>
mean_no_short <- t(x_no_short_CCM) %*% means</pre>
sd_no_short <- sqrt(t(x_no_short_CCM) %*% covmat %*% x_no_short_CCM)</pre>
ones <- matrix(1, nrow = 31, ncol = 1)</pre>
A2 <- t(ones) %*% solve(covmat) %*% means
B <- t(means) %*% solve(covmat) %*% means
C <- t(ones) %*% solve(covmat) %*% ones
D <- B*C - A2^2
mean short <- 0.04
mean no short <- 0.04
plot(stdev_CCM, Ribar_CCM, xlab = "Standard Deviation", ylab = "Expected Return", pch = 16,
xlim = c(0,0.15), ylim = c(0, 0.04))
# Efficient frontier:
minvar <- 1/C
minE <- A2/C
sdeff \leftarrow seq((minvar)^0.5, 1, by = 0.0001)
options(warn = -1)
y1 \leftarrow (A2 + sqrt(D*(C*sdeff^2 - 1)))*(1/C)
y2 \leftarrow (A2 - sqrt(D*(C*sdeff^2 - 1)))*(1/C)
options(warn = 0)
points(sdeff, y1, type = "1")
points(sdeff, y2, type = "1")
points(sd_short, mean_short, pch = 16, col = "red")
points(sd_no_short, mean_no_short, pch = 16, col = "orange")
```

