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# Portfolio Optimization: A Monte Carlo Study

# Prepared by Tom Miller, June 20, 2025

# Monte Carlo Study Showing Portfolio Opportunity Sets

# Generate above 700 random asset allocation sets: 4-by-1 vectors.
# For each set, the four portfolio asset weights must sum to 1.
# So, we generate three and set the fourth so the sum is 1.
# Weights can be negative (indicating a short position in the asset).
# Many sets of weights will be suboptimal.

# Assume multivariate normal distributions of returns
# with means, standard deviations, and correlations
# as specified in the problem setup.

# functions from 'Modern Applied Statistics in S' (Venables and Ripley 2002)
library(MASS)
library(ggplot2) # visualization of the simulation

## Warning: package 'ggplot2' was built under R version 4.3.3

# mean returns for portfolio assets A, B, C, and D
targetMeanVector <- c(0.1759, 0.2176, 0.6119, 0.2617)

# standard deviations of returns
targetSDVector <- c(0.2822, 0.3229, 0.4390, 0.2640)

# correlation matrix for the four portfolio assets
targetCorMatrix <- matrix(c(
  1.0000, 0.5540, 0.4868, 0.6141,
  0.5540, 1.0000, 0.5767, 0.6477,
  0.4868, 0.5767, 1.0000, 0.5949,
  0.6141, 0.6477, 0.5949, 1.0000),
  nrow = 4, ncol = 4)

# compute the covariance matrix
targetCovMatrix <- diag(targetSDVector) %*% targetCorMatrix %*% diag(targetSDVector)

# number of return sets to generate
sampleSize <- 700
```

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# set seed so results are reproducible across executions
set.seed(1111)
# generate multivariate normal returns
returnsData <- mvrnorm(n = sampleSize,
                      mu = targetMeanVector,
                      Sigma = targetCovMatrix)
returnsDataFrame <- as.data.frame(returnsData)
names(returnsDataFrame) <- c("AAPL", "GOOG", "META", "MSFT")

# check statistics from the generated returns data
print(summary(returnsDataFrame))

```

```

##          AAPL          GOOG          META          MSFT
## Min.      :-0.73804  Min.      :-0.74710  Min.      :-0.7406  Min.      :-0.58119
## 1st Qu.: -0.02278  1st Qu.:  0.00171  1st Qu.:  0.3420  1st Qu.:  0.07259
## Median :  0.16967  Median :  0.21182  Median :  0.6300  Median :  0.26465
## Mean     :  0.16513  Mean      :  0.21387  Mean      :  0.6231  Mean      :  0.25628
## 3rd Qu.:  0.35227  3rd Qu.:  0.44121  3rd Qu.:  0.9385  3rd Qu.:  0.43137
## Max.     :  1.09814  Max.      :  1.23139  Max.      :  1.7923  Max.      :  1.30150

```

```

# check correlation matrix of the generated returns data
cat("\nTarget Correlation of Returns for Generated Data:\n")

```

```

##
## Target Correlation of Returns for Generated Data:

```

```

print(targetCorMatrix)

```

```

##          [,1]  [,2]  [,3]  [,4]
## [1,] 1.0000 0.5540 0.4868 0.6141
## [2,] 0.5540 1.0000 0.5767 0.6477
## [3,] 0.4868 0.5767 1.0000 0.5949
## [4,] 0.6141 0.6477 0.5949 1.0000

```

```

cat("\nActual Correlation of Returns in Generated Data:\n")

```

```

##
## Actual Correlation of Returns in Generated Data:

```

```

print(cor(returnsDataFrame))

```

```

##          AAPL          GOOG          META          MSFT
## AAPL 1.0000000 0.5494631 0.4979419 0.6255798
## GOOG 0.5494631 1.0000000 0.5762297 0.6435262
## META 0.4979419 0.5762297 1.0000000 0.6076079
## MSFT 0.6255798 0.6435262 0.6076079 1.0000000

```

```

# compute covariance matrix for the sample data
# to be used later in portfolio calculations
cat("\nTarget Covariance of Returns for Generated Data:\n")

```

```

##
## Target Covariance of Returns for Generated Data:

```

```

print(targetCovMatrix)

```

```

##           [,1]      [,2]      [,3]      [,4]
## [1,] 0.07963684 0.05048180 0.06030761 0.04575094
## [2,] 0.05048180 0.10426441 0.08174901 0.05521358
## [3,] 0.06030761 0.08174901 0.19272100 0.06894653
## [4,] 0.04575094 0.05521358 0.06894653 0.06969600

```

```

cat("\nActual Covariance of Returns in Generated Data:\n")

```

```

##
## Actual Covariance of Returns in Generated Data:

```

```

dataCovMatrix <- cov(returnsDataFrame)
print(dataCovMatrix)

```

```

##           AAPL      GOOG      META      MSFT
## AAPL 0.07618120 0.04782908 0.06058624 0.04688190
## GOOG 0.04782908 0.09946246 0.08011184 0.05510543
## META 0.06058624 0.08011184 0.19433135 0.07272662
## MSFT 0.04688190 0.05510543 0.07272662 0.07372201

```

```

# Generate a random sets of weights using
# a uniform distribution from -1 to 1 if
# shortsOK is TRUE. Otherwise use a uniform
# distribution from 0 to 1 with scaling to
# ensure that the weights sum to 1.
makeWeights <- function(shortsOK) {
  if (shortsOK) {
    threeWeights <- runif(3, min = -1, max = 1)
    fourthWeight <- 1 - sum(threeWeights) # ensures sum of 1
    return(c(threeWeights,fourthWeight))
  }
  if (!shortsOK) {
    initialWeights <- runif(4, min = 0, max = 1)
    return(initialWeights/sum(initialWeights)) # ensures sum of 1
  }
}

# generate sets of portfolio weights in shortsOK state
set.seed(9999) # set seed so results are reproducible
weightsMatrix <- matrix(NA, nrow = sampleSize, ncol = 4)
for (iset in 1:sampleSize)

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weightsMatrix[iset,] <- makeWeights(shortsOK = TRUE)

# Compute the portfolio return for each set of weights
# for each set of weights applied to each set of returns,
# storing calculations in a new data frame for the portfolios.
portfolioResults = NULL
for (iset in 1:sampleSize) {
  Positions <- 1 # has shorts is 1
  w1 <- weightsMatrix[iset,1]
  w2 <- weightsMatrix[iset,2]
  w3 <- weightsMatrix[iset,3]
  w4 <- weightsMatrix[iset,4]
  if (w1 > 0 && w2 > 0 && w3 > 0 && w4 > 0) Positions <- 2 # no shorts

  returnVector <- weightsMatrix[iset,] %*% t(returnsData)
  returnMean <- mean(returnVector)
  returnSD <- as.numeric(sqrt(t(weightsMatrix[iset,]) %*% dataCovMatrix %*% weightsMatrix[iset,]))
  thisPortfolioResult <- data.frame(w1, w2, w3, w4, Positions, returnMean, returnSD)
  portfolioResults <- rbind(portfolioResults, thisPortfolioResult)
}

# check portfolio results
cat("\nSummary of portfolio results with short positions allowed:\n")

```

```

##
## Summary of portfolio results with short positions allowed:

```

```

print(summary(portfolioResults))

```

```

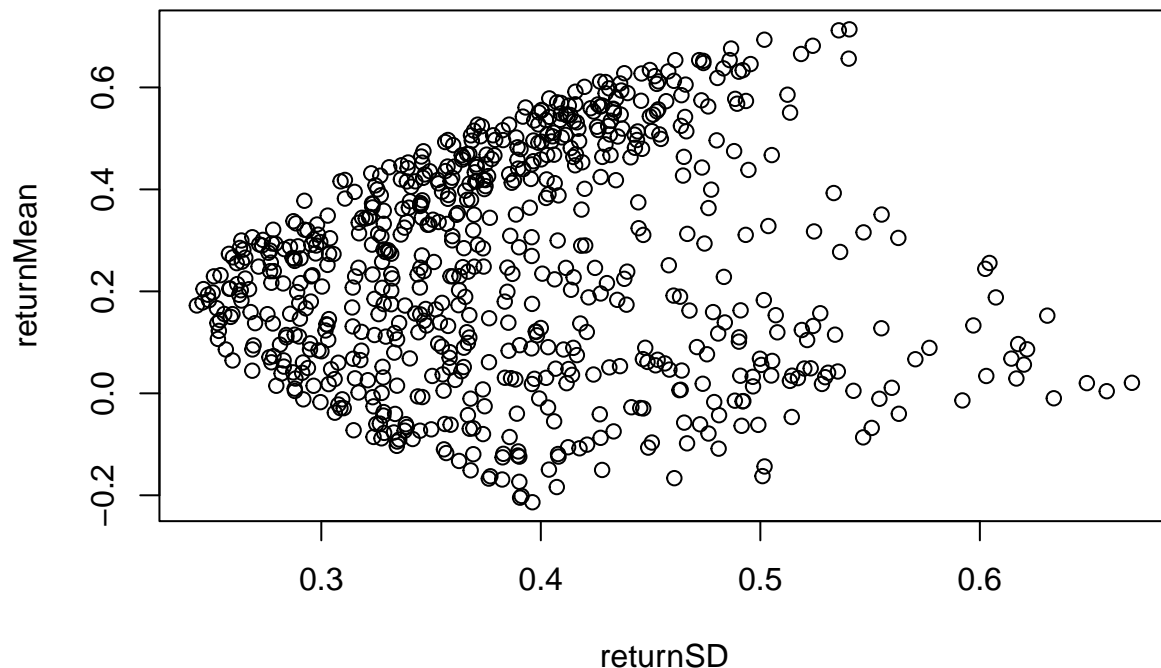
##           w1           w2           w3           w4
## Min.      :-0.99885   Min.      :-0.99972   Min.      :-0.99973   Min.      :-1.4131
## 1st Qu.: -0.53759   1st Qu.: -0.50020   1st Qu.: -0.54154   1st Qu.:  0.2482
## Median:  -0.01290   Median:   0.02385   Median:  -0.05586   Median:   0.9678
## Mean:    -0.01094   Mean:     0.01550   Mean:    -0.03202   Mean:     1.0275
## 3rd Qu.:  0.47658   3rd Qu.:  0.51116   3rd Qu.:  0.46668   3rd Qu.:  1.7545
## Max.      0.99934   Max.       0.99482   Max.       0.99975   Max.       3.6639
## Positions  returnMean  returnSD
## Min.      :1.00   Min.      :-0.21361   Min.      :0.2433
## 1st Qu.: 1.00   1st Qu.:  0.06426   1st Qu.:0.3277
## Median: 1.00   Median:   0.23668   Median:0.3762
## Mean:   1.02   Mean:     0.24487   Mean:   0.3868
## 3rd Qu.: 1.00   3rd Qu.:  0.42767   3rd Qu.:0.4362
## Max.:   2.00   Max.:     0.71392   Max.:   0.6693

```

```

with(portfolioResults, plot(returnSD, returnMean))

```



```

shortsOKResults <- portfolioResults
shortsOKResults$ShortsOK = rep("Shorts OK", times = sampleSize)

# generate sets of portfolio weights with no shorts allowed
set.seed(9999) # set seed so results are reproducible
weightsMatrix <- matrix(NA, nrow = sampleSize, ncol = 4)
for (iset in 1:sampleSize)
  weightsMatrix[iset,] <- makeWeights(shortsOK = FALSE)

# Compute the portfolio return for each set of weights
# for each set of weights applied to each set of returns,
# storing calculations in a new data frame for the portfolios.
portfolioResults = NULL
for (iset in 1:sampleSize) {
  Positions <- 1 # has shorts is 1
  w1 <- weightsMatrix[iset,1]
  w2 <- weightsMatrix[iset,2]
  w3 <- weightsMatrix[iset,3]
  w4 <- weightsMatrix[iset,4]
  if (w1 > 0 && w2 > 0 && w3 > 0 && w4 > 0) Positions <- 2 # no shorts
  returnVector <- weightsMatrix[iset,] %*% t(returnsData)
  returnMean <- mean(returnVector)
  returnSD <- as.numeric(sqrt(t(weightsMatrix[iset,]) %*% dataCovMatrix %*% weightsMatrix[iset,]))
  thisPortfolioResult <- data.frame(w1, w2, w3, w4, Positions, returnMean, returnSD)
  portfolioResults <- rbind(portfolioResults, thisPortfolioResult)
}

```

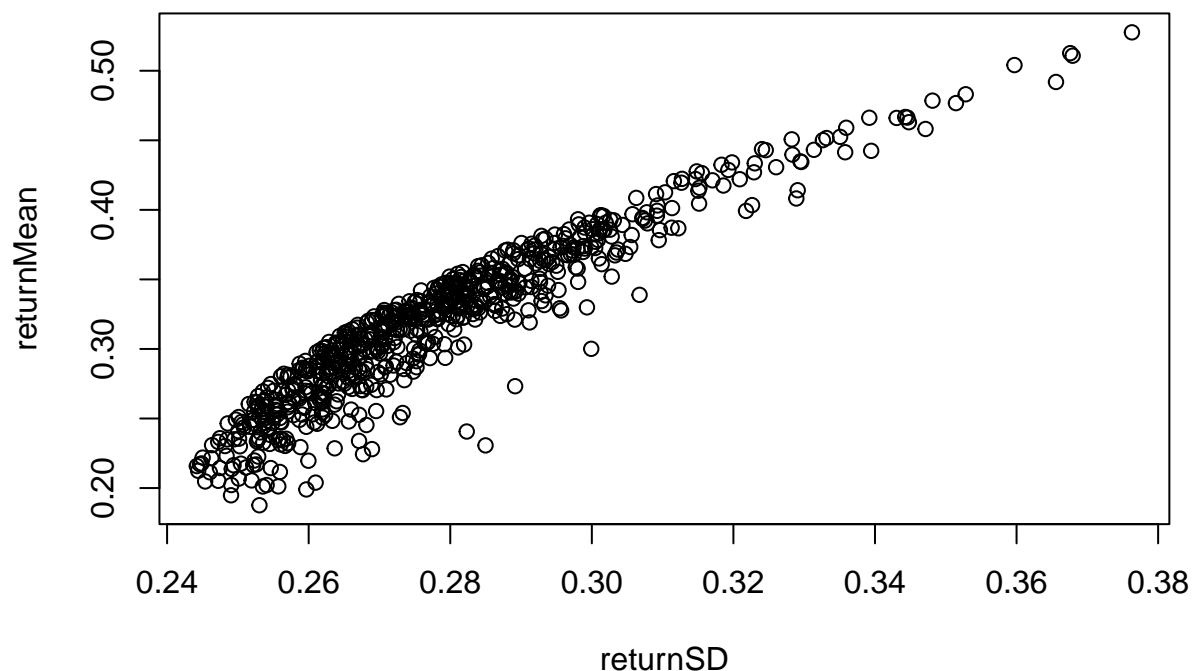
```
# check portfolio results
cat("\nSummary of portfolio results with long positions only:\n")
```

```
##
## Summary of portfolio results with long positions only:
```

```
print(summary(portfolioResults))
```

```
##           w1           w2           w3           w4
## Min.    :0.0000709   Min.    :0.0006501   Min.    :0.001087   Min.    :0.0000689
## 1st Qu.:0.1314990   1st Qu.:0.1367747   1st Qu.:0.160793   1st Qu.:0.1339661
## Median :0.2466478   Median :0.2540729   Median :0.258966   Median :0.2333626
## Mean    :0.2495896   Mean    :0.2467636   Mean    :0.260156   Mean    :0.2434911
## 3rd Qu.:0.3444534   3rd Qu.:0.3356580   3rd Qu.:0.341754   3rd Qu.:0.3381186
## Max.    :0.7673518   Max.    :0.7484792   Max.    :0.751567   Max.    :0.8926828
## Positions  returnMean    returnSD
## Min.      :2   Min.      :0.1876   Min.      :0.2442
## 1st Qu.:2   1st Qu.:0.2749   1st Qu.:0.2625
## Median :2   Median :0.3211   Median :0.2749
## Mean      :2   Mean      :0.3185   Mean      :0.2783
## 3rd Qu.:2   3rd Qu.:0.3519   3rd Qu.:0.2889
## Max.      :2   Max.      :0.5276   Max.      :0.3763
```

```
with(portfolioResults, plot(returnSD, returnMean)) # preliminary plot on console
```



```

noShortsResults <- portfolioResults
noShortsResults$ShortsOK = rep("Long Positions Only", times = sampleSize)

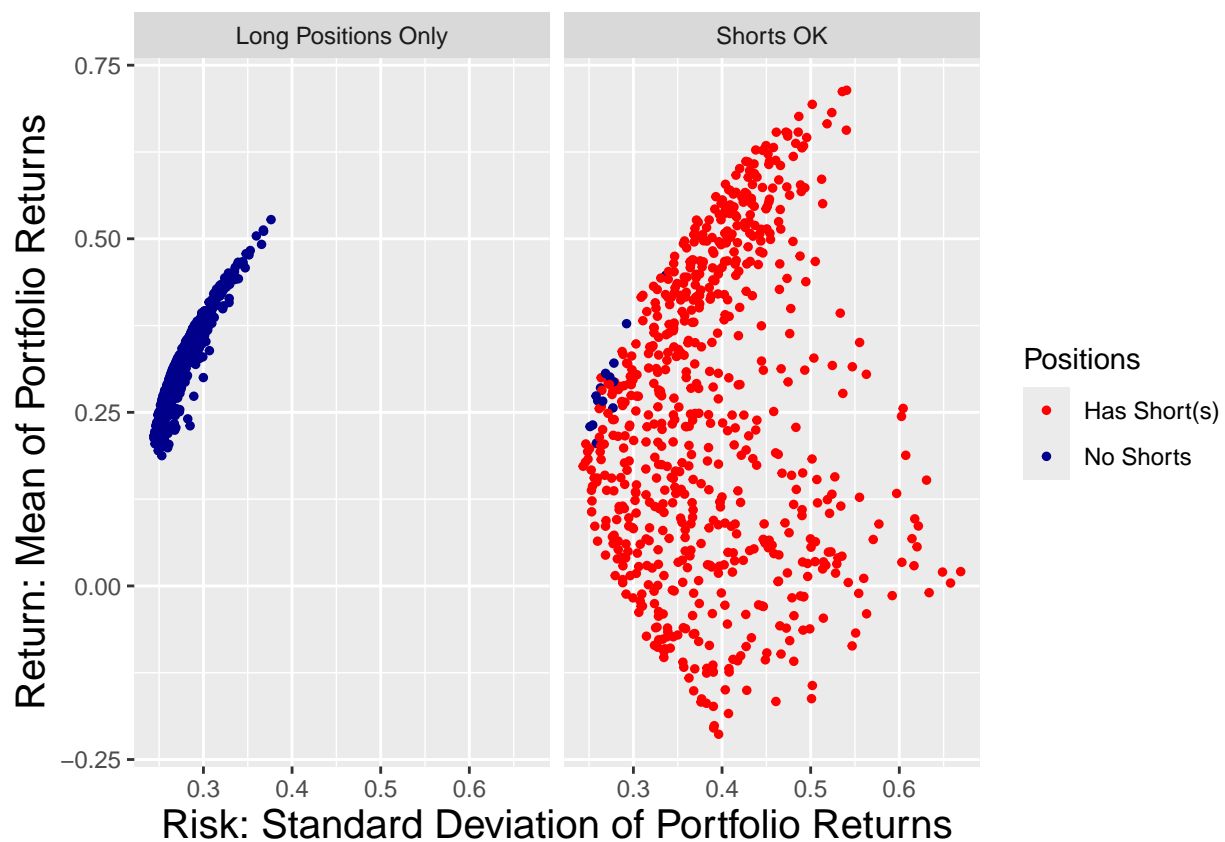
# merge the two data frames
plottingFrame = rbind(shortsOKResults,noShortsResults)
plottingFrame$Positions <- factor(plottingFrame$Positions,
  labels = c("Has Short(s)", "No Shorts"))

plottingFrame$ShortsOK = factor(plottingFrame$ShortsOK)

# complete plot for both conditions: Shorts Allowed and No Shorts
facetPlot <- ggplot(plottingFrame, aes(x=returnSD,y=returnMean,
  colour=Positions)) +
  geom_point(size = 1) +
  xlab("Risk: Standard Deviation of Portfolio Returns") +
  ylab("Return: Mean of Portfolio Returns") +
  scale_color_manual(values = c("red","darkblue")) +
  theme(axis.title = element_text(size=15)) +
  facet_wrap( ~ ShortsOK, ncol = 2)

print(facetPlot)

```



```

# export pdf plot for inclusion in LaTeX document
pdf(file = "451-portfolio-optimization-monte-carlo-figure.pdf", width = 11, height = 8.5)
print(facetPlot)

```

```
dev.off()
```

```
## pdf  
## 2
```

```
cat("\nRun Complete\n")
```

```
##  
## Run Complete
```