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Constrained Portfolio Optimisation

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R Code

The following R code is for solving the unconstrained portfolio optimisation using quadratic programming techniques.

```
library(zoo)
library(PerformanceAnalytics)
library(PortfolioAnalytics)
library(quadprog)
library(quantmod)
#Assets tickers
tickers <- c("FB", "AAPL", "AMZN", "NFLX", "TSLA","GOOG",
             "BA","PYPL","IBM","GS","AIG","MA","JMP","V"
             ,"BABA","INTC","C","AMD","WIX","ACN","BLK"
             ,"WFC","Bk","USB","SIVB","PNC","UBS","WBS",
             "TFC","COO","LH","MD","MDT","MRK","WAT","SYK"
             ,"SM","EOG","SSL","RES","PSX","APA","D","MMM"
             ,"CE","BB","BR","MSFT","EA","PTC","AON","VC"
             ,"HIG","L","CB")
portfolioprices <- NULL
for(ticker in tickers){
  portfolioprices <-
    cbind(portfolioprices,
          getSymbols.yahoo(ticker,from='2016-01-03',
                                          to='2020-10-05',
                                          periodicity ='daily',
                                          auto.assign=FALSE)[,4])
}
#portfolio prices
portfolioprices2 <-data.matrix(as.data.frame(portfolioprices))
#The return matrix.
retmatrix <- data.frame(matrix(NA, nrow = nrow(portfolioprices)-1,
                               ncol = ncol(portfolioprices)))
for(i in 1:(nrow(portfolioprices2)-1) ) {
  for (j in 1:ncol(portfolioprices2)){
    retmatrix[i,j]<-(portfolioprices2[i+1,j]-
                     portfolioprices2[i,j])/portfolioprices2[i,j]
  }
}

covariance <- cov(retmatrix) #The co-variance matrix.
print(covariance)
n <- ncol(retmatrix)
#Matrix defining the constraints.
Amat <- cbind(1, colMeans(retmatrix), diag(n))
#vector appearing in the quadratic function to be minimised.
bvec <- c(1, rep(0, n))

# create the objective matrix
Dmat <-  2*covariance

#Specify the number of equality constraints
```

```
meq <- 1
risk.Vector <- rep(0,100)
count <- 1
minret <- 0 # set min return
# seting max return to be max of average returns
maxret <- max(colMeans(retmatrix))-10^-10
number_points <- 100
return_vals <- seq(from = minret,
                   to = maxret,length.out = number_points)

# Sets of weights
weights_list <- matrix(0, number_points, n)

for ( return_val in return_vals ) {
  #dvec <- colMeans(retmatrix) * return_val
  dvec <- rep(0,n)*return_val
  b.Vector <- c(1, return_val, rep(0,n))
  out <- solve.QP( Dmat, dvec, Amat, bvec = b.Vector, meq )
  risk.Vector[count] <- out$value
  weights_list[count,] <- out$solution
  count <- count + 1
  count

}
#Plotting the efficient frontier.
plot(risk.Vector, return_vals,
     xlab="Risk",ylab="Return",col='red')
```