

# STAT 631 Homework 5

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```
load("HW05.Rdata")
```

1)

a)

In the case with short selling there are explicit solutions for the minimum variance portfolio. These are:

$$w_{\min.v} = \frac{\Sigma^{-1}\mathbf{1}}{\mathbf{1}^T \Sigma^{-1} \mathbf{1}}, \text{Mean} = \frac{\mu^T \Sigma^{-1} \mathbf{1}}{\mathbf{1}^T \Sigma^{-1} \mathbf{1}}, \text{Variance} = \frac{1}{\mathbf{1}^T \Sigma^{-1} \mathbf{1}}$$

First we find the mean vector  $\mu$  and covariance matrix  $\Sigma$ .

```
y.mu <- apply(y,MARGIN = 2, FUN = mean)
y.mu
```

|  | AMZN      | KO        | NKE       | PFE       | TSLA      | UNH       | URI       | V         |
|--|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|
|  | 0.4206254 | 0.1692590 | 0.2216723 | 0.1501095 | 0.6957931 | 0.4128102 | 0.5011812 | 0.3970618 |

```
y.S <- var(y)
y.S
```

|      | AMZN      | KO       | NKE       | PFE      | TSLA      | UNH       | URI       |
|------|-----------|----------|-----------|----------|-----------|-----------|-----------|
| AMZN | 17.892637 | 2.431855 | 5.576128  | 2.564539 | 11.972386 | 2.977977  | 8.972939  |
| KO   | 2.431855  | 6.238607 | 3.820411  | 2.972533 | 4.151542  | 3.972146  | 5.440775  |
| NKE  | 5.576128  | 3.820411 | 14.638695 | 2.812862 | 9.371497  | 4.268385  | 10.709588 |
| PFE  | 2.564539  | 2.972533 | 2.812862  | 8.787569 | 3.860339  | 4.452199  | 5.824990  |
| TSLA | 11.972386 | 4.151542 | 9.371497  | 3.860339 | 57.338308 | 6.231686  | 16.666210 |
| UNH  | 2.977977  | 3.972146 | 4.268385  | 4.452199 | 6.231686  | 12.173168 | 7.763183  |

|      |          |          |           |          |           |          |           |
|------|----------|----------|-----------|----------|-----------|----------|-----------|
| URI  | 8.972939 | 5.440775 | 10.709588 | 5.824990 | 16.666210 | 7.763183 | 39.914885 |
| V    | 5.837551 | 3.829269 | 5.747779  | 3.145227 | 7.124705  | 4.292491 | 8.992063  |
|      | V        |          |           |          |           |          |           |
| AMZN | 5.837551 |          |           |          |           |          |           |
| KO   | 3.829269 |          |           |          |           |          |           |
| NKE  | 5.747779 |          |           |          |           |          |           |
| PFE  | 3.145227 |          |           |          |           |          |           |
| TSLA | 7.124705 |          |           |          |           |          |           |
| UNH  | 4.292491 |          |           |          |           |          |           |
| URI  | 8.992063 |          |           |          |           |          |           |
| V    | 9.207900 |          |           |          |           |          |           |

Now we can compute the weights.

```
one_vector <- rep(1, 8)
y.S_inv <- solve(y.S)
w_min.v <- y.S_inv%*%one_vector/as.numeric((t(one_vector)%*%y.S_inv%*%one_vector))
colnames(w_min.v) = "Weights"
w_min.v
```

|      |             |
|------|-------------|
|      | Weights     |
| AMZN | 0.11244506  |
| KO   | 0.42307306  |
| NKE  | 0.08199088  |
| PFE  | 0.26278105  |
| TSLA | -0.01238153 |
| UNH  | 0.07393439  |
| URI  | -0.07294053 |
| V    | 0.13109762  |

The expected return of this portfolio is:

```
expected_return <- as.numeric(t(y.mu)%*%y.S_inv%*%one_vector)/
  as.numeric(t(one_vector)%*%y.S_inv%*%one_vector)
expected_return
```

```
[1] 0.2139305
```

The risk of this portfolio is:

```
risk <- sqrt(1/as.numeric(t(one_vector)%*%y.S_inv%*%one_vector))
risk
```

[1] 2.086775

b)

The tangency portfolio also has an explicit solution with short selling allowed. With  $\mu_{ex} = \mu - \mu_f \mathbf{1}$ :

$$w_T = \frac{\Sigma^{-1} \mu_{ex}}{\mathbf{1}^T \Sigma^{-1} \mu_{ex}}$$

The annual risk-free rate is 4.37%. The weekly risk-free rate is 4.37%/52. The allocation weights of the tangency portfolio are below:

```
weekly_risk_free <- .0437/52
mu_excess = y.mu - weekly_risk_free
w_tangency = y.S_inv %*% mu_excess /
  as.numeric(one_vector^T %*% y.S_inv %*% mu_excess)
colnames(w_tangency) = "Weights"
w_tangency
```

```

              Weights
AMZN  0.200277036
KO    -0.151625522
NKE   -0.184406419
PFE   -0.151336852
TSLA  0.118670461
UNH   0.520386933
URI   0.008547379
V     0.639486985
```

The mean and variance of the tangency portfolio have the explicit solutions:

$$\text{Mean} = \frac{\mu^T \Sigma^{-1} \mu_{ex}}{\mathbf{1}^T \Sigma^{-1} \mu_{ex}}, \text{Variance} = \frac{\mu_{ex}^T \Sigma^{-1} \mu_{ex}}{(\mathbf{1}^T \Sigma^{-1} \mu_{ex})^2}$$

So the expected return of the tangency portfolio is:

```
expected_return_tangency <- as.numeric((t(y.mu) %*% y.S_inv %*% mu_excess)/
  (t(one_vector)%*% y.S_inv %*% mu_excess))
expected_return_tangency
```

```
[1] 0.5505735
```

And the risk of the tangency portfolio is:

```
risk_tangency <- as.numeric(sqrt((t(mu_excess)%*%y.S_inv%*%mu_excess)/
  (t(one_vector)%*%y.S_inv%*%mu_excess)^2))
risk_tangency
```

```
[1] 3.351737
```

c)

To plot the efficient frontier requires us to compute the risk for various expected returns. The explicit solution is below:

$$\text{Amat} = [\mu, 1], H = \text{Amat}^T \Sigma^{-1} \text{Amat} = \begin{bmatrix} C & B \\ B & A \end{bmatrix}, \Delta = \det(H)$$

$$\text{Risk}_{opt} = \sqrt{\frac{Am^2 - 2Bm + C}{\Delta}}$$

```
m.R = seq(-.15,1,0.001)
Amat = cbind(y.mu, one_vector)
H = t(Amat)%*%y.S_inv%*%Amat
A = H[2,2];B = H[1,2];C = H[1,1]; Delta = det(H)
sd.R = sqrt((A*m.R^2 - 2*B*m.R + C)/Delta)
```

Our plot is below:

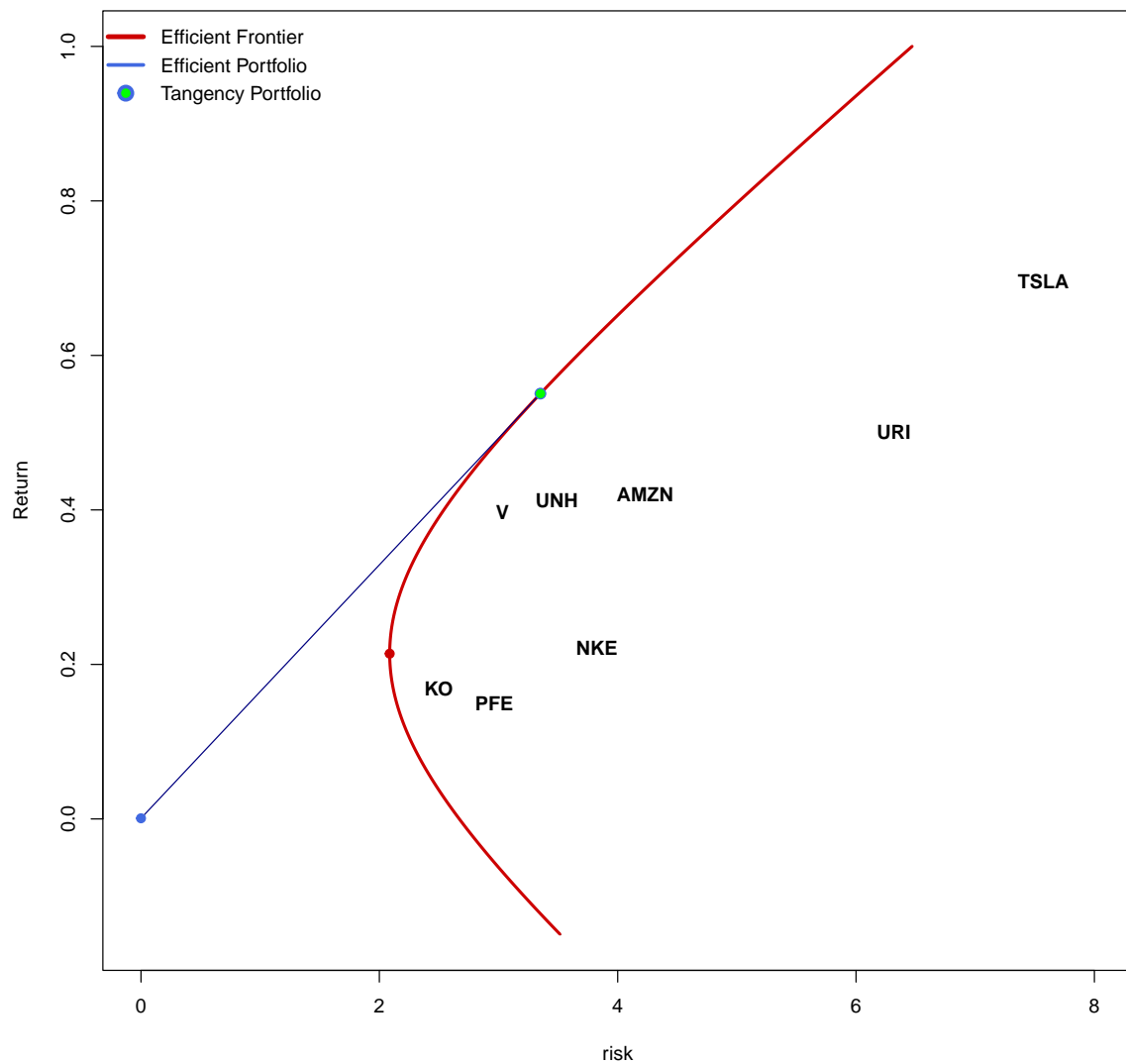
```
mu.min = -.15
y.sd = sqrt(diag(y.S))
plot(sd.R,m.R, type = "l", xlim= c(0, 8), xlab = "risk", ylab = "Return")
lines(sd.R[m.R > mu.min],m.R[m.R > mu.min], lwd = 2.5, col = "red3")
lines(c(0,risk_tangency), c(weekly_risk_free, expected_return_tangency), col = "navy")
points(0, weekly_risk_free, pch = 19, col = "royalblue")
points(risk_tangency,expected_return_tangency, pch = 21, col = "royalblue", bg = "green", ce
```

```

points(risk, expected_return, pch = 19, col = "red3")
for(i in 1:8){
  text(y.sd[i], y.mu[i], syb[i], font = 2)
}

legend("topleft",c("Efficient Frontier", "Efficient Portfolio", "Tangency Portfolio"), lty =
      y.intersp = 1.2, bty = "n", xjust = 5)

```



d)

To find the efficient portfolio with risk of 2.5% we use the fact that:

$$\sigma_p = w_r \sigma_r$$

where  $\sigma_p$  is the risk of the portfolio,  $w_r$  is the weight on the tangency portfolio and  $\sigma_r$  is the risk of the tangency portfolio.

So then:

$$w_r = \frac{\sigma_p}{\sigma_r}$$

Where  $\sigma_p$  is the allowed risk of 2.5% and  $\sigma_r$  is the risk of the tangency portfolio is 3.351737%. So the weight is:

```
w_r = as.numeric(2.5/risk_tangency)
w_r
```

```
[1] 0.7458818
```

With the weight we can find the portfolio needed:

```
port_d <- rbind(1 - w_r, w_r*w_tangency)
rownames(port_d) <- c("Risk Free",syb)
port_d
```

|           | Weights      |
|-----------|--------------|
| Risk Free | 0.254118178  |
| AMZN      | 0.149383001  |
| KO        | -0.113094721 |
| NKE       | -0.137545396 |
| PFE       | -0.112879407 |
| TSLA      | 0.088514139  |
| UNH       | 0.388147154  |
| URI       | 0.006375335  |
| V         | 0.476981717  |

Now with the weight we use the fact that the return of the portfolio is:

$$\mu_p = w_r \mu_r + (1 - w_r) \mu_{rf}$$

where  $w_r$  is the weight on the tangency portfolio,  $\mu_r$  is the return of the tangency portfolio and  $\mu_{rf}$  is the risk free return. So the return of the efficient portfolio with allowed risk 2.5% is:

```
mu_p = w_r*expected_return_tangency + (1 - w_r)*weekly_risk_free
mu_p
```

```
[1] 0.4108763
```

e)

To find the efficient portfolio of a target return 0.55% we reverse the steps of part d. First we find the weight that corresponds to this return through:

$$\mu_p = w_r \mu_r + (1 - w_r) \mu_{rf}$$

$$w_r = \frac{\mu_p - \mu_{rf}}{\mu_r - \mu_{rf}}$$

So the weight on the tangency portfolio is:

```
w_r_e <- (.55 - weekly_risk_free)/(expected_return_tangency - weekly_risk_free)
w_r_e
```

```
[1] 0.9989568
```

With the weight we can find the portfolio:

```
port_e <- rbind(1 - w_r_e, w_r_e*w_tangency)
rownames(port_e) <- c("Risk Free",syb)
port_e
```

|           | Weights      |
|-----------|--------------|
| Risk Free | 0.001043194  |
| AMZN      | 0.200068108  |
| KO        | -0.151467348 |
| NKE       | -0.184214047 |
| PFE       | -0.151178979 |
| TSLA      | 0.118546664  |
| UNH       | 0.519844068  |
| URI       | 0.008538463  |
| V         | 0.638819876  |

So the risk of this portfolio can be computed through:

$$\sigma_p = w_r \sigma_r$$

```
risk_e <- w_r_e*risk_tangency  
risk_e
```

```
[1] 3.348241
```

f)

We go through the same procedure as part e to start:

$$w_r = \frac{\mu_p - \mu_{rf}}{\mu_r - \mu_{rf}}$$

```
w_r_f = (.85 - weekly_risk_free)/(expected_return_tangency - weekly_risk_free)  
w_r_f
```

```
[1] 1.544676
```

We find that in order to get this target return of 0.85% we would need a weight over 1 in the tangency portfolio. This is not feasible as we would need to take out a loan to make up the difference. Therefore we look to the efficient frontier to find a portfolio of risky assets with the desired return. We can find the portfolio by using the two constraints:

$$A_{mat} = \begin{bmatrix} \mu & 1 \end{bmatrix}, b_{vec} = \begin{bmatrix} 0.85\% \\ 1 \end{bmatrix}$$

```
library(quadprog)  
A_mat = cbind(y.mu, one_vector)  
b_vec = c(.85,1)  
zeros = rep(0,8)  
out = solve.QP(Dmat = y.S, dvec = zeros, A_mat = A_mat, bvec = b_vec, meq = 2)  
w_f = out$solution; names(w_f) = syb  
cat("Portfolio:"); w_f
```

Portfolio:



|  | AMZN       | KO          | NKE         | PFE         | TSLA       | UNH        |
|--|------------|-------------|-------------|-------------|------------|------------|
|  | 0.27839904 | -0.66279023 | -0.42135306 | -0.51967333 | 0.23523444 | 0.91748341 |
|  | URI        | V           |             |             |            |            |
|  | 0.08102667 | 1.09167306  |             |             |            |            |

We can find the risk from what we computed in part c to plot the efficient frontier. The details of the computation are in part c. The risk is:

```
sd.R[which(m.R == .85)]
```

```
[1] 5.377229
```

2)

a)

In the case without short selling there is no explicit solution for the minimum variance portfolio. We set two constraints:

$$A_{\text{mat}} = \begin{bmatrix} 1 & I_n \end{bmatrix}, b_{\text{vec}} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

```
Amat = cbind(one_vector, diag(8))
bvec = c(1, zeros)
out = solve.QP(Dmat = y.S, dvec = zeros, Amat = Amat, bvec = bvec, meq = 1)
w.min = out$solution; w.min = w.min*(abs(w.min) > 10e-7); names(w.min) = syb;
mu.min = sum(w.min*y.mu); sd.min = sqrt(2*out$val)
w.min
```

|  | AMZN       | KO         | NKE        | PFE        | TSLA       | UNH        | URI        |
|--|------------|------------|------------|------------|------------|------------|------------|
|  | 0.09798502 | 0.43290178 | 0.05251957 | 0.25712960 | 0.00000000 | 0.05812796 | 0.00000000 |
|  | V          |            |            |            |            |            |            |
|  | 0.10133607 |            |            |            |            |            |            |

```
c(return = mu.min, risk = sd.min)
```

|  | return    | risk      |
|--|-----------|-----------|
|  | 0.2289597 | 2.1267106 |

b)

In the case without short selling there is also no explicit solution for the tangency portfolio. We find a portfolio  $w_*$  and then re-scale so the sum of weights is equal to 1. So we can choose the below constraints to find  $w_*$ :

$$A_{\text{mat}} = \begin{bmatrix} \mu - \mu_f 1 & I_n \end{bmatrix}, b_{\text{vec}} = \begin{bmatrix} m \\ 0 \end{bmatrix}$$

```
amat = cbind(y.mu - weekly_risk_free, diag(8))
bvec = c(0.4, zeros)
out = solve.QP(Dmat = y.S, dvec = zeros, Amat = amat, bvec = bvec, meq = 1)
w.star = out$solution; names(w.star) = syb
w.T = w.star/sum(w.star)
w.T = w.T * (abs(w.T) > 10e-7)
cat("Portfolio: ");w.T
```

Portfolio:

|   | AMZN       | KO         | NKE        | PFE        | TSLA       | UNH        | URI        |
|---|------------|------------|------------|------------|------------|------------|------------|
|   | 0.15136990 | 0.00000000 | 0.00000000 | 0.00000000 | 0.08657754 | 0.35365583 | 0.00000000 |
| V |            |            |            |            |            |            |            |
|   | 0.40839673 |            |            |            |            |            |            |

```
mu.T = sum(w.T*y.mu); s.T = sqrt(2*out$value)/sum(out$solution)
c(return = mu.T, risk = s.T)
```

|  | return    | risk      |
|--|-----------|-----------|
|  | 0.4320615 | 2.7162153 |

c)

First we need to create the risk-return sets. We do this by finding the portfolios with pre-specified returns ranging from the lowest per stock return (PFE with return 0.15%) and the highest per stock return (TSLA with return 0.7%) and then computing the risk of each portfolio.

```
m.R = seq(round(min(y.mu)+.005,3), round(max(y.mu)-.005,3), 0.001);
Amat = cbind(y.mu, one_vector, diag(8)) ## for positive w
sd.R = c();
for(i in 1:length(m.R)){
```

```

bvec = c(m.R[i],1, zeros) ## for nonnegative w
out= solve.QP(y.S, dvec = zeros, Amat=Amat, bvec = bvec, meq = 2);
sd.R[i] = sqrt(2*out$value)
}

```

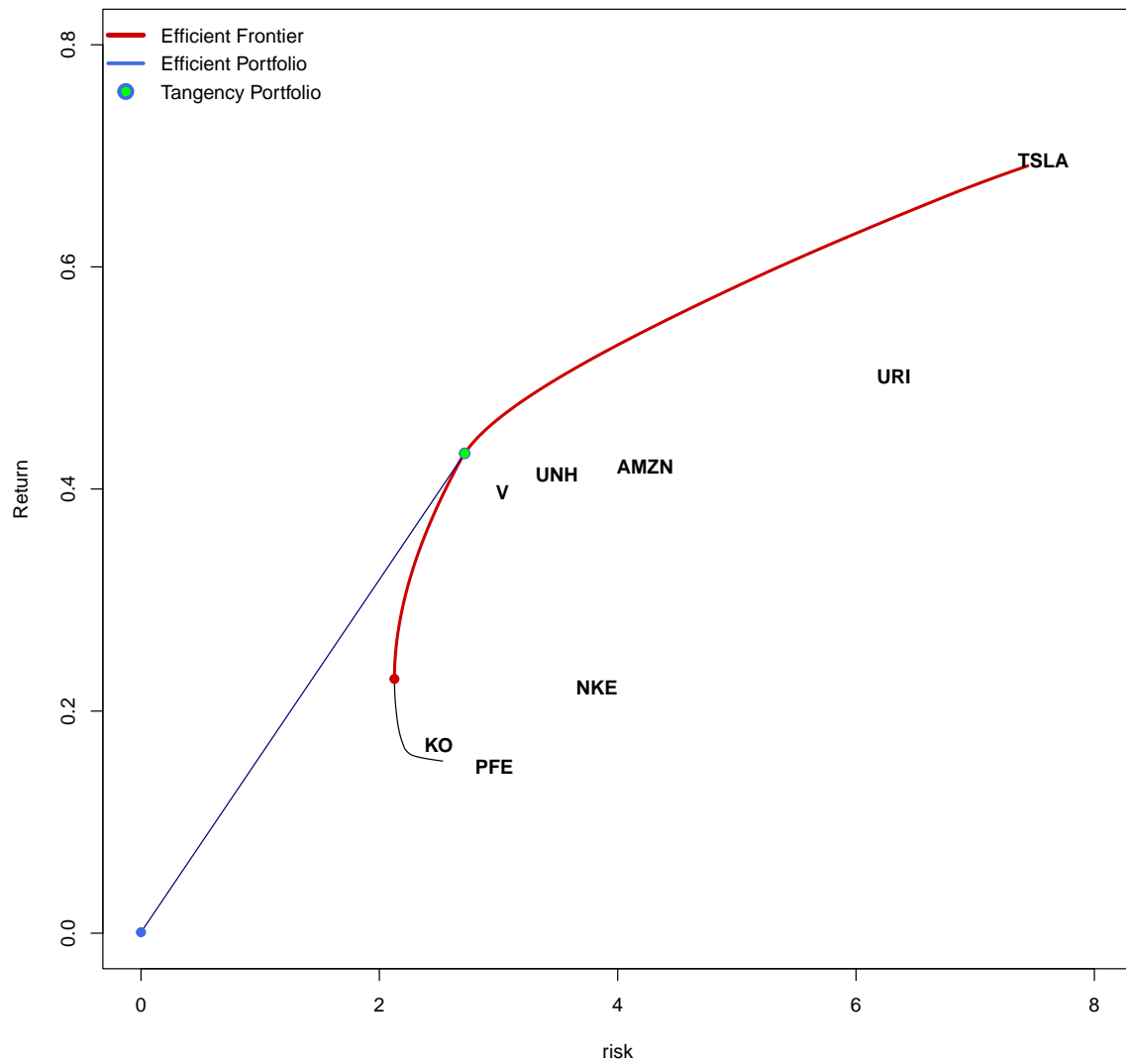
Now we can create our plot:

```

plot(sd.R,m.R, type = "l", xlim= c(0, 8), ylim = c(0, .8), xlab = "risk", ylab = "Return")
lines(sd.R[m.R > mu.min],m.R[m.R > mu.min], lwd = 2.5, col = "red3")
lines(c(0,s.T), c(weekly_risk_free, mu.T), col = "navy")
points(0, weekly_risk_free, pch = 19, col = "royalblue")
points(s.T,mu.T, pch = 21, col = "royalblue", bg = "green", cex = 1.2)
points(sd.min, mu.min, pch = 19, col = "red3")
for(i in 1:8){
  text(y.sd[i], y.mu[i], syb[i], font = 2)
}

legend("topleft",c("Efficient Frontier", "Efficient Portfolio", "Tangency Portfolio"), lty =
      y.intersp = 1.2, bty = "n", xjust = 5)

```



d)

To find the portfolio with allowed risk 2.5% we follow a procedure similar to the case with short selling:

$$\sigma_p = w_r \sigma_r$$

$$w_r = \sigma_p / \sigma_r$$

So we have the weights below:

```
w_r_d <- 2.5/s.T
c(Risk_Free = 1 - w_r_d, Tangency_Portfolio = w_r_d)
```

|  | Risk_Free  | Tangency_Portfolio |
|--|------------|--------------------|
|  | 0.07960167 | 0.92039833         |

```
weights_d <- c(1 - w_r_d, w_r_d*w.T)
names(weights_d) = c("Risk Free", syb)
weights_d
```

|  | Risk Free  | AMZN       | KO         | NKE        | PFE        | TSLA       | UNH        |
|--|------------|------------|------------|------------|------------|------------|------------|
|  | 0.07960167 | 0.13932061 | 0.00000000 | 0.00000000 | 0.00000000 | 0.07968582 | 0.32550423 |
|  | URI        | V          |            |            |            |            |            |
|  | 0.00000000 | 0.37588767 |            |            |            |            |            |

And the portfolio's return and risk:

```
c(return_d = as.numeric(t(weights_d)%*% c(weekly_risk_free, y.mu)),
  risk_d = 2.5)
```

|  | return_d  | risk_d    |
|--|-----------|-----------|
|  | 0.3977356 | 2.5000000 |

e)

Since the return of 0.55% is greater than the 0.43%, the return of the tangency portfolio, we need to find a portfolio off of the efficient frontier. The weight, return and risk are below:

```
m.e = 0.55
Amat = cbind(y.mu, one_vector, diag(8))
bvec = c(m.e, 1, zeros)
out = solve.QP(y.S, dvec = zeros, Amat=Amat, bvec = bvec, meq = 2);
sd.e = sqrt(2*out$value)
portfolio_weights = out$solution;names(portfolio_weights) = syb;
portfolio_weights = portfolio_weights*(abs(portfolio_weights) > 10e-7)
cat("Portfolio Weights:");portfolio_weights
```

Portfolio Weights:

| AMZN       | KO         | NKE        | PFE        | TSLA       | UNH        | URI        |
|------------|------------|------------|------------|------------|------------|------------|
| 0.03685150 | 0.00000000 | 0.00000000 | 0.00000000 | 0.45942698 | 0.30673284 | 0.09598675 |
| V          |            |            |            |            |            |            |
| 0.10100193 |            |            |            |            |            |            |

```
cat("Return: 0.55%", "Risk: ", sd.e )
```

Return: 0.55% Risk: 4.370181

f)

The portfolio with the highest return that we can construct is one holding solely the highest return stock, in this case holding TSLA with a return of .7%. Since the target return of 0.85% is larger than the return of a portfolio holding only TSLA it can not be constructed without short selling.