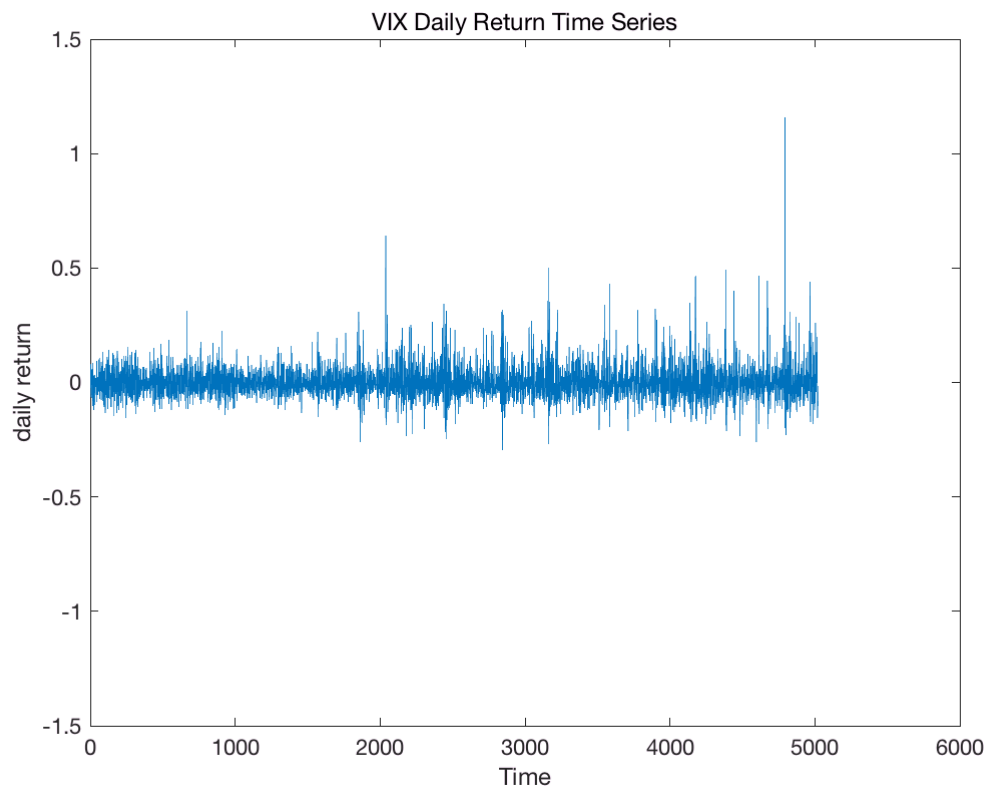


Question 1

```
filename = 'HW1_Due20180207.xlsx';
Q1='DataSource';
[num1,text1,raw1]=xlsread(filename,Q1,'A:B','');
vix_price =cell2mat(raw1(3:end,2));
n1=length(vix_price);
m1=length(vix_price)-1;
v_ret =nan(m1,1);
for i= 1:n1-1
    v_ret(i)=(vix_price(i+1)/vix_price(i))-1;
end
v_ret
```

```
v_ret =
    0
-0.0219
    0.0811
    0.0333
-0.0257
-0.0610
    0.0171
-0.0545
-0.0662
    0.0541
```

```
%%draw daily return time series
plot(v_ret)
ylim([-1.5,1.5])
xlabel('Time')
ylabel('daily return')
title('VIX Daily Return Time Series')
```



```
%%calculate the sample moments (mean,skewness, and kurtosis) for the VIX daily returns
```

```
vix_mean=nanmean(v_ret)
```

```
    vix_mean = 0.0024
```

```
vix_sigma=std(v_ret)
```

```
    vix_sigma = 0.0713
```

```
vix_skewness=skewness(v_ret)
```

```
    vix_skewness = 2.0142
```

```
vix_kurtosis =kurtosis(v_ret)
```

```
    vix_kurtosis = 22.2821
```

```
%%test whether the VIX daily returns were normally distributed or not.
```

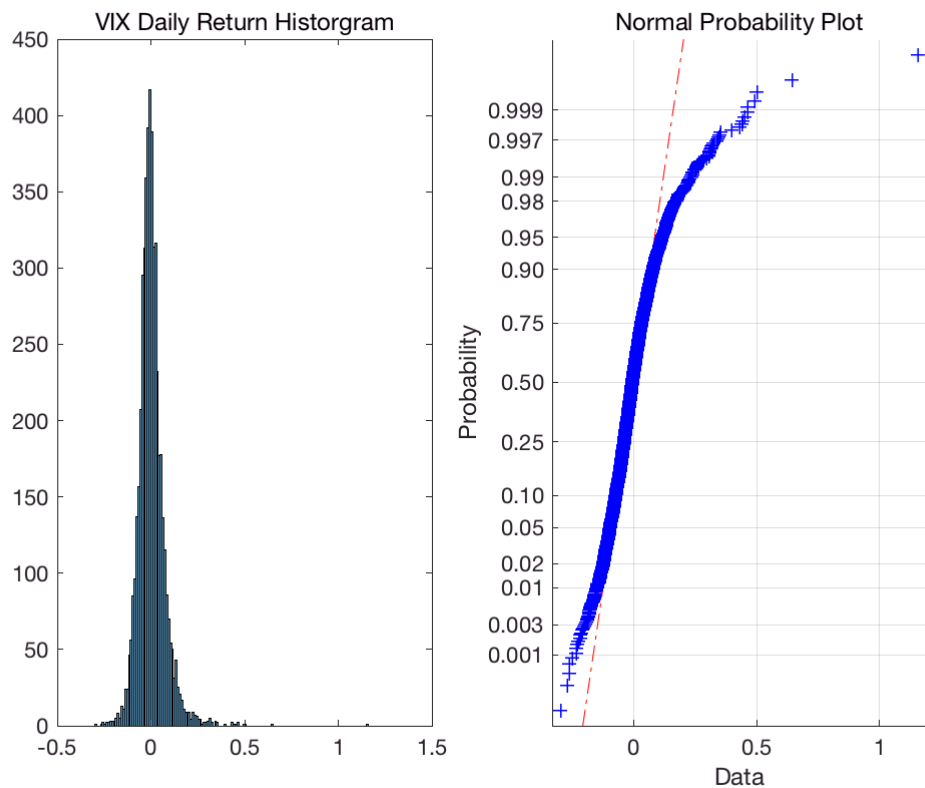
```
subplot(1,2,1);
```

```
histogram(v_ret);
```

```
title('VIX Daily Return Histogram');
```

```
subplot(1,2,2);
```

```
normplot(v_ret);
```



```
%%% Jarque-Bera Test
% h=1 if reject normal distribution
[h,p]=jbtest(v_ret);
```

警告: P is less than the smallest tabulated value, returning 0.001.

```
fprintf('Jarque-Bera Test: h=%d, p-val=%.4f.\n',h,p);
```

Jarque-Bera Test: h=1, p-val=0.0010.

```
% One-sample Kolmogorov-Smirnov test
% h=1 if reject normal distribution
[h,p]=kstest((v_ret-vix_mean)/vix_sigma);
fprintf('KS Test: h=%d, p-val=%.4f.\n',h,p);
```

KS Test: h=1, p-val=0.0000.

```
%%Conclusion
```

Conclusion: the fat tail of Normal probability plot and the two tests which all have h=1, indicating rejection of normality. Thus, the VIX daily return is not normal distributed.

Question 2

```
filename = 'HW1_Due20180207.xlsx';
Q2='DataSource';
[num,text,row]=xlsread(filename,Q2,'D:E','');
Date = cell2mat(row(3:end,1));
Date =Date +693960;
Date =datestr(Date);

price = cell2mat(row(3:end,2));
price0 =price(2:end,1);
n= length(price);
m=n-1;
dret=zeros(m,1);
pricel = price(1:n-1,1);
for i =1:n-1
    dret(i)=(price(i+1)/price(i))-1;
end
dret
```

```
dret =
    0.0065
    0.0080
   -0.0265
   -0.0023
    0.0102
    0.0182
   -0.0117
    0.0168
    0.0076
   -0.0059
```

```
%%daily volatility
spy_sigma = std(dret)
```

```
spy_sigma = 0.0121
```

```
%%annual volatility
spy_ansigma= spy_sigma*sqrt(252)
```

```
spy_ansigma = 0.1924
```

```
%%plot of dret
x = (1:m)';
y= dret;
subplot(3,1,1)
plot(x,y);
title('SPY daily return time series');
xlabel('Time');
ylabel('daily return of SPY');
%%semi daily volatility
spy_semi=std(dret(dret<0))
```

```
spy_semi = 0.0093
```

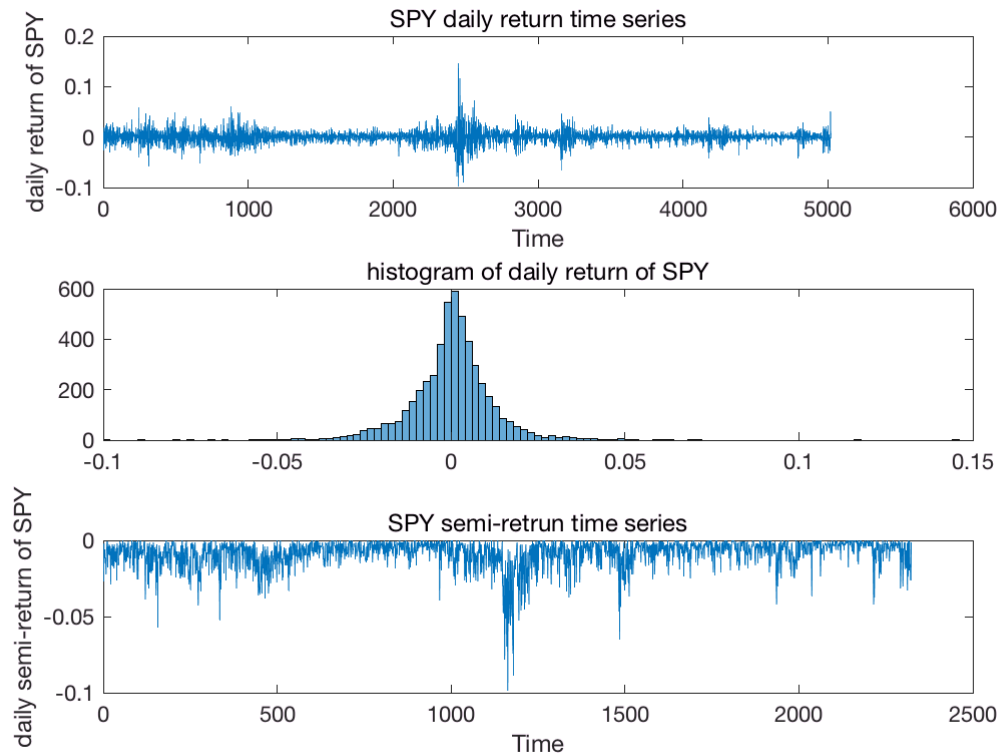
```
%%semi annual volatility
spy_ansemi=spy_semi*sqrt(252)
```

```
spy_ansemi = 0.1471
```

```

%%plot semi-retrun time series
subplot(3,1,3)
plot(dret(dret<0))
title('SPY semi-retrun time series');
xlabel('Time');
ylabel('daily semi-return of SPY');
%% histogram of daily return
subplot(3,1,2);
histogram(dret);
title('histogram of daily return of SPY');

```



```

% set a target amount, e.g., the maximum loss in a month in return
% percentage form, say, -2%
daily_shortfall=-0.02;
% caculate the shortfall probability (realized)
shortfall_prob_realized=sum(dret<daily_shortfall)/length(dret)

shortfall_prob_realized = 0.0464

```

```

% what is the theoretical shortfall probability assuming monthly returns
% follow normal distribution with mean estimated by sample mean, and
% variance estimated by sample variance.
z_value=(daily_shortfall-nanmean(dret))/spy_sigma

z_value = -1.6680

```

```

% the theoretical short fall probability

```

```
shortfall_prob_theoretical=normcdf(z_value)
```

```
shortfall_prob_theoretical = 0.0477
```

```
% Value At Risk - 95%  
% i.e., choose the bottom 5% percentile threshold value  
VaR=prctile(dret,5)
```

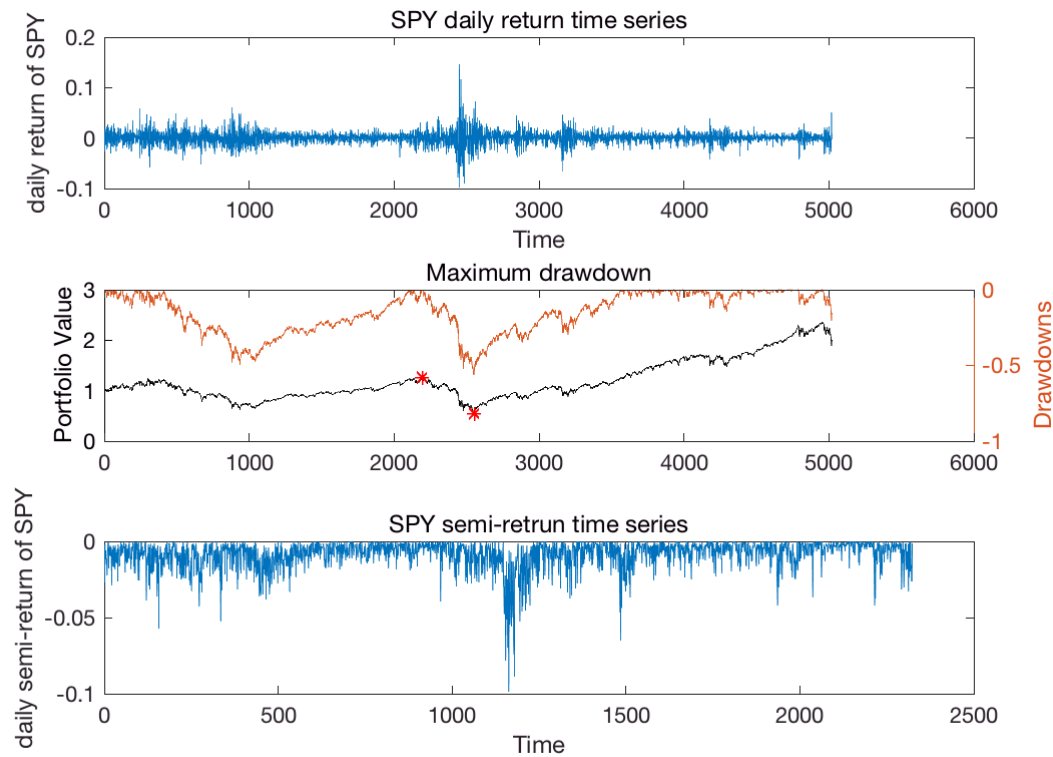
```
VaR = -0.0194
```

```
% CVaR: the mean return for all returns below the VaR threshold  
CVaR=mean(dret(dret<VaR))
```

```
CVaR = -0.0288
```

Maximum drawdown

```
% plot the portfolio value curve for the Emerging Market Equity  
% Benchmark  
portval_vec=[1;cumprod(1+dret)];  
yyaxis left;  
plot(portval_vec);  
title('Maximum drawdown');  
ylabel('Portfolio Value');  
  
% calculate drawdown at each time point  
drawdown_vec=nan(length(portval_vec),1);  
for i=1:length(portval_vec)  
    drawdown_vec(i)=portval_vec(i)/max(portval_vec(1:i))-1;  
end  
yyaxis right  
hold on  
plot(drawdown_vec);  
ylabel('Drawdowns');  
  
% locate the maximum drawdown  
[max_drawdown,pos_right]=min(drawdown_vec);  
[blah,pos_left]=max(portval_vec(1:pos_right));  
hold on  
yyaxis left  
plot([pos_left,pos_right],portval_vec([pos_left,pos_right]),'r*');
```



```
% output the maximum drawdown
fprintf('Max Drawdown = %.1f%%.\n',max_drawdown*100);
```

Max Drawdown = -56.5%.

Question 3

```
%%Problem 3
%% Data: Multi-Asset Class Monthly Returns on DataSource_CAPMAssetClasses

%%Assumptions 1)- no risk-free asset in the investable universe and all the
risky assets are given on DataSource_CAPMAssetClasses;
%%2)- use historical covariance matrix and historical returns to estimate
covariance matrix and expected return vector (see on lecture 2 notes)
%%3)- no short-sell allowed. (before we have no such constraint)

%%1. Run an optimization to locate the minimum-variance portfolio on the new
frontier.
%%2A. Run optimizations to locate two other efficient portfolios different
than the minimum-variance portfolio
%%2B. Prove or disprove the minimum-variance portfolio can be represented as
a combination of the two portfolios found on 2A.
%%3. Plot both the old efficient frontier without the short-sell constraint
and the new efficient frontier with the short-sell constraint on the same
graph.

% the investable universe
sheet2='DataSource_CAPMAssetClasses'
```

```
sheet2 = 'DataSource_CAPMAssetClasses'
```

```
[num3,text3,row3]=xlsread(filename,sheet2,'A:M','');
```

```
ac_name =text3(1,2:end)
```

```
ac_name = 1×12 cell 数组
```

```
'Fixed-Income: Treasury Inflation Protected Securities (TIPS)' 'Fixed-  
Income: U.S. Treasury Bonds and Investment Grade Corp Bonds' 'Fixed-Income:  
U.S. High Yield Corp Bonds' 'Fixed-Income: Non U.S. Sovereign Bonds'  
'Equities: U.S. Large Cap Growth' 'Equities: U.S. Large Cap Value'  
'Equities: U.S. Small Cap Growth' 'Equities: U.S. Small Cap Value'  
'Equities: Developed Countries Non-US ' 'Equities: Emerging Markets'  
'Commodities' 'Real Estate'
```

```
ac_mret=cell2mat(row3(3:end,2:end));
```

```
% calculate the covariance matrix based on their historical monthly
```

```
% return data
```

```
% note: we annualize the covariance matrix by multiplying the matrix with 12
```

```
Sigma=nancov(ac_mret)*12;
```

```
% let's use the historical annualized as our estimates of asset
```

```
% expected returns
```

```
ExpRet=nan(length(ac_name),1);
```

```
for ii=1:length(ac_name)
```

```
ExpRet(ii)=geomean(1+0.01*ac_mret(:,ii))^12*100-100;
```

```
fprintf('Annualized Return=%.2f%% for %s.\n',ExpRet(ii),ac_name{ii});
```

```
end
```

```
Annualized Return=5.20% for Fixed-Income: Treasury Inflation Protected  
Securities (TIPS).
```

```
Annualized Return=4.55% for Fixed-Income: U.S. Treasury Bonds and Investment  
Grade Corp Bonds.
```

```
Annualized Return=6.30% for Fixed-Income: U.S. High Yield Corp Bonds.
```

```
Annualized Return=3.40% for Fixed-Income: Non U.S. Sovereign Bonds.
```

```
Annualized Return=5.05% for Equities: U.S. Large Cap Growth.
```

```
Annualized Return=6.15% for Equities: U.S. Large Cap Value.
```

```
Annualized Return=6.14% for Equities: U.S. Small Cap Growth.
```

```
Annualized Return=8.23% for Equities: U.S. Small Cap Value.
```

```
Annualized Return=3.52% for Equities: Developed Countries Non-US .
```

```
Annualized Return=8.52% for Equities: Emerging Markets.
```

```
Annualized Return=0.57% for Commodities.
```

```
Annualized Return=9.65% for Real Estate.
```



```

% unit vector
unit_vec=ones(length(ac_name),1);

% calculate the minimum variance portfolio and the pseudo-maximum sharpe
% ratio portfolio
Sigma_inv=inv(Sigma);
w_minv=Sigma_inv*unit_vec/(unit_vec'*Sigma_inv*unit_vec);
w_msr=Sigma_inv*ExpRet/(unit_vec'*Sigma_inv*ExpRet);
fprintf('minv and msr portfolio respectively:(short-sell allowed) ')

minv and msr portfolio respectively:(short-sell allowed)

```

```
[w_minv,w_msr]
```

```

ans =
    -0.2648    -0.1828
     1.2018     1.1267
     0.0495     0.0833
    -0.0480    -0.0695
     0.0311     0.0653
     0.0757     0.0430
    -0.0032    -0.0586
     0.0372     0.1238
    -0.0335    -0.1366
    -0.0242     0.0446

```

```

% min variance portfolio without shortsell

% we'll formulate a portfolio by adding constraints to the min-variance
portfolio(no shortselling)
% put return target on constraint
target_ret=0.7;

% objective function
fun=@(w) w'*Sigma*w;

% initial guess
w0=ones(length(ac_name));
w0=w0/sum(w0);

% constraint 1: target return
% Ax<=b
A=-ExpRet';
b=-target_ret;

% constraint 2: ub and lb for no shortsell
ub=ones(12,1);
lb=zeros(12,1);

% constraint 3: fully invested
Aeq=ones(1,length(ac_name));
beq=1;

% run the optimization
w_new_minv=fmincon(fun,w0,A,b,Aeq,beq,lb,ub);

Local minimum found that satisfies the constraints.

Optimization completed because the objective function is non-decreasing in
feasible directions, to within the default value of the optimality tolerance,
and constraints are satisfied to within the default value of the constraint

```

tolerance.

<stopping criteria details>

```
[w_new_minv,w_minv]
```

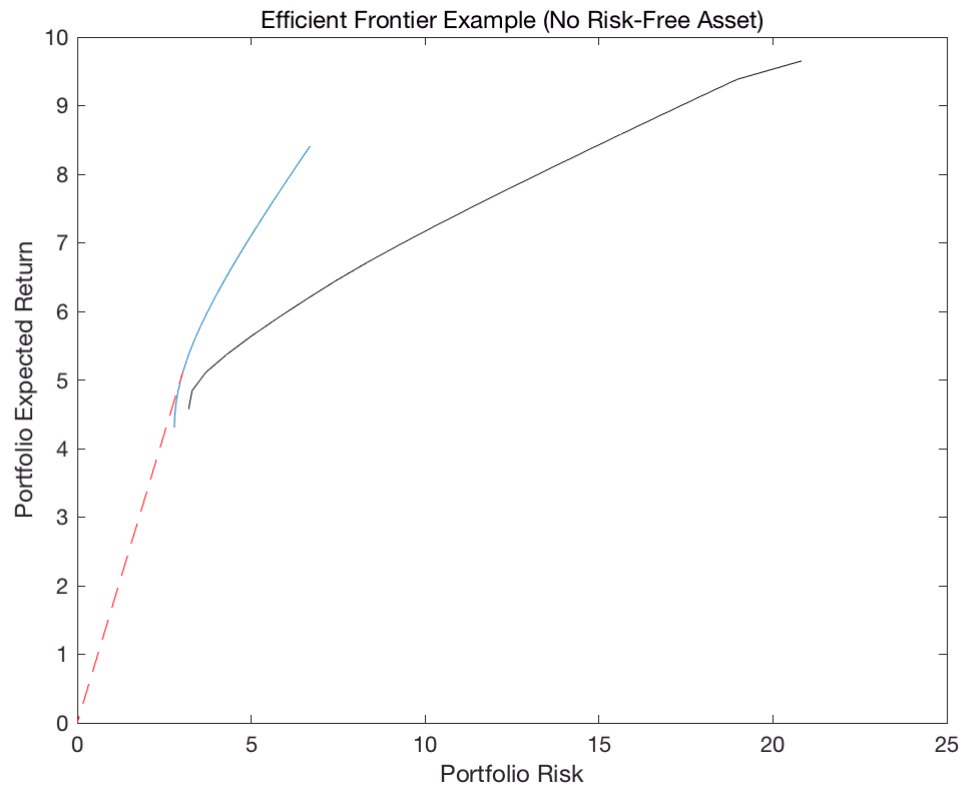
```
ans =  
    0.0000   -0.2648  
    0.9265    1.2018  
    0.0000    0.0495  
    0.0000   -0.0480  
    0.0130    0.0311  
    0.0388    0.0757  
    0.0081   -0.0032  
    0.0000    0.0372  
    0.0000   -0.0335  
    0.0000   -0.0242
```

```
% build the efficient frontier  
  
% define a function for weight vector in terms a combination of two known  
% portfolios  
wt_func=@(c) (1-c)*w_minv+c*w_msr;  
% portfolio return and volatility function  
pret_func=@(c) ExpRet'*wt_func(c);  
prisk_func=@(c) sqrt(wt_func(c)'*Sigma*wt_func(c));  
  
%without short sell, efficient frontier  
%%2A. Run optimizations to locate two other efficient portfolios different than  
the minimum-variance portfolio  
Numassets =length(ac_name)
```

```
Numassets = 12
```

```
p = Portfolio();  
p = setDefaultConstraints(p,Numassets);  
p = setInitPort(p,1/12);  
p = setAssetMoments(p,ExpRet,Sigma);  
  
pwgt = estimateFrontier(p,20);  
[prsk,pret] = estimatePortMoments(p,pwgt);  
  
% Plot efficient frontier  
  
% efficient frontier with short selling  
c=0:0.2:5;  
close();  
plot(arrayfun(prisk_func,c),arrayfun(pret_func,c));% draw EF with short  
selling  
title('Efficient Frontier Example (No Risk-Free Asset)')  
xlabel('Portfolio Risk')  
ylabel('Portfolio Expected Return')  
hold on  
plot([0,prisk_func(1)],[0,pret_func(1)], '--','Color','r');%%get tangent  
portfolio
```

```
hold on
plot(prsk,pret,'Color','black') %%draw EF without shorsell
```



```
% EF with shortsell allowed(blue curve), EF without shortsell(black
curve)
```

```
%-----%
-----%
```

```
%2A. Run optimizations to locate two other efficient portfolios
different than the minimum-variance portfolio, the weights in pwgt
are all the optimized portfolios on the new efficient frontier, so I
choose two as the optimized portfolios other than min-variance
portfolio.
```

```
awgt= pwgt(:,1)
```

```
awgt =
    0
    0.9265
    0
    0
    0.0130
    0.0388
    0.0081
    0
    0
    0
```

```
bwgt =pwgt(:,2)
```

```
bwgt =  
      0  
      0.8922  
      0.0426  
      0  
      0  
      0.0095  
      0  
      0.0557  
      0  
      0
```

```
aret= ExpRet'*awgt
```

```
aret = 4.5768
```

```
arsk=awgt'*Sigma*bwgt
```

```
arsk = 10.3420
```

```
bret= ExpRet'*awgt
```

```
bret = 4.5768
```

```
brsk=awgt'*Sigma*bwgt
```

```
brsk = 10.3420
```

```
[w_new_minv,awgt,bwgt]
```

```
ans =  
      0.0000      0      0  
      0.9265      0.9265      0.8922  
      0.0000      0      0.0426  
      0.0000      0      0  
      0.0130      0.0130      0  
      0.0388      0.0388      0.0095  
      0.0081      0.0081      0  
      0.0000      0      0.0557  
      0.0000      0      0  
      0.0000      0      0
```

```
%It's easy to see these two portfolios(not the min-variance portfolio) can not  
be combined to formulate the min-variance portfolio
```