

# 투자론

- R과 Excel을 통한 금융데이터 분석 -

8주차

R의 Collecting Data 및 Portfolio Analysis

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## Unit 03

# Portfolio Analysis with R (2)

# Overview

- Warm up with PortfolioAnalytics package
- Beta Estimation
- Optimal Portfolio Analysis



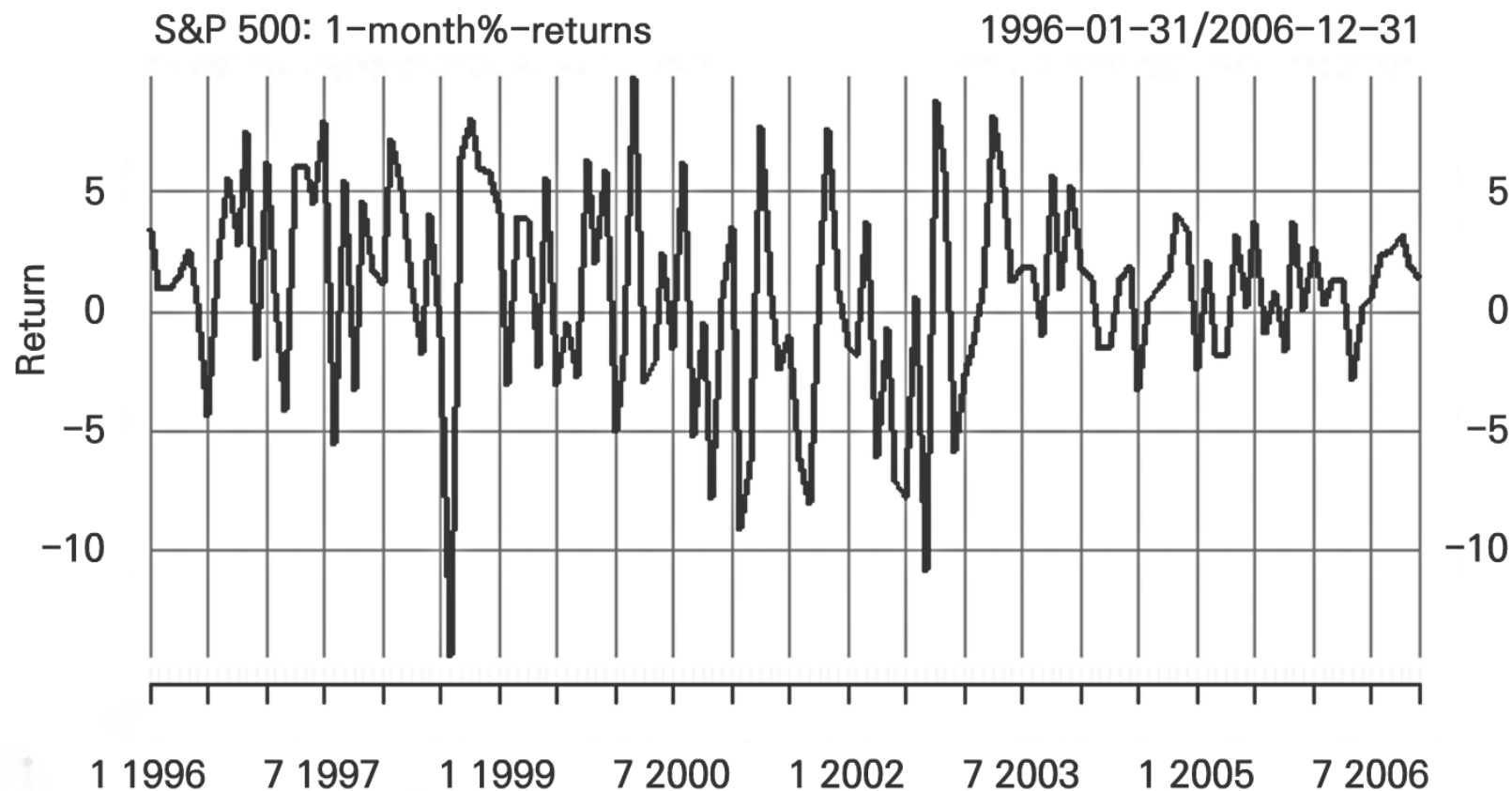
## ◆ Warm up with pre-installed R data set

```
library(PortfolioAnalytics)
data(managers)
sp500.ret <- managers$SP500
plot(sp500.ret*100,
     main="S&P 500: 1-month %-returns",
     ylab="return",
     xlab="date")
```

## ◆ Warm up with pre-installed R data set

	HAM1	HAM2	HAM3	HAM4	HAM5	HAM6	EDHEC LS EQ	SP500 TR	US 10Y TR	US 3m TR
1996-01-31	0.0074	NA	0.0349	0.0222	NA	NA	NA	0.034000	0.00380	0.00456
1996-02-29	0.0193	NA	0.0351	0.0195	NA	NA	NA	0.009300	-0.03532	0.00398
1996-03-31	0.0155	NA	0.0258	-0.0098	NA	NA	NA	0.009600	-0.01057	0.00371
1996-04-30	-0.0091	NA	0.0449	0.0236	NA	NA	NA	0.014700	-0.01739	0.00428
1996-05-31	0.0076	NA	0.0353	0.0028	NA	NA	NA	0.025800	-0.00543	0.00443
1996-06-30	-0.0039	NA	-0.0303	-0.0019	NA	NA	NA	0.003800	0.01507	0.00412
1996-07-31	-0.0231	NA	-0.0337	-0.0446	NA	NA	NA	-0.044200	-0.00100	0.00454
1996-08-31	0.0395	-0.0001	0.0461	0.0351	NA	NA	NA	0.021100	-0.00448	0.00451
1996-09-30	0.0147	0.1002	0.0653	0.0757	NA	NA	NA	0.056300	0.02229	0.00470
1996-10-31	0.0288	0.0338	0.0395	-0.0180	NA	NA	NA	0.027600	0.02869	0.00428
1996-11-30	0.0156	0.0737	0.0666	0.0458	NA	NA	NA	0.075600	0.02797	0.00427
1996-12-31	0.0176	0.0298	0.0214	0.0439	NA	NA	NA	-0.019800	-0.02094	0.00442
1997-01-31	0.0212	0.0794	0.0771	0.0437	NA	NA	0.0281	0.062500	-0.00055	0.00457

## ◆ Warm up with pre-installed R data set

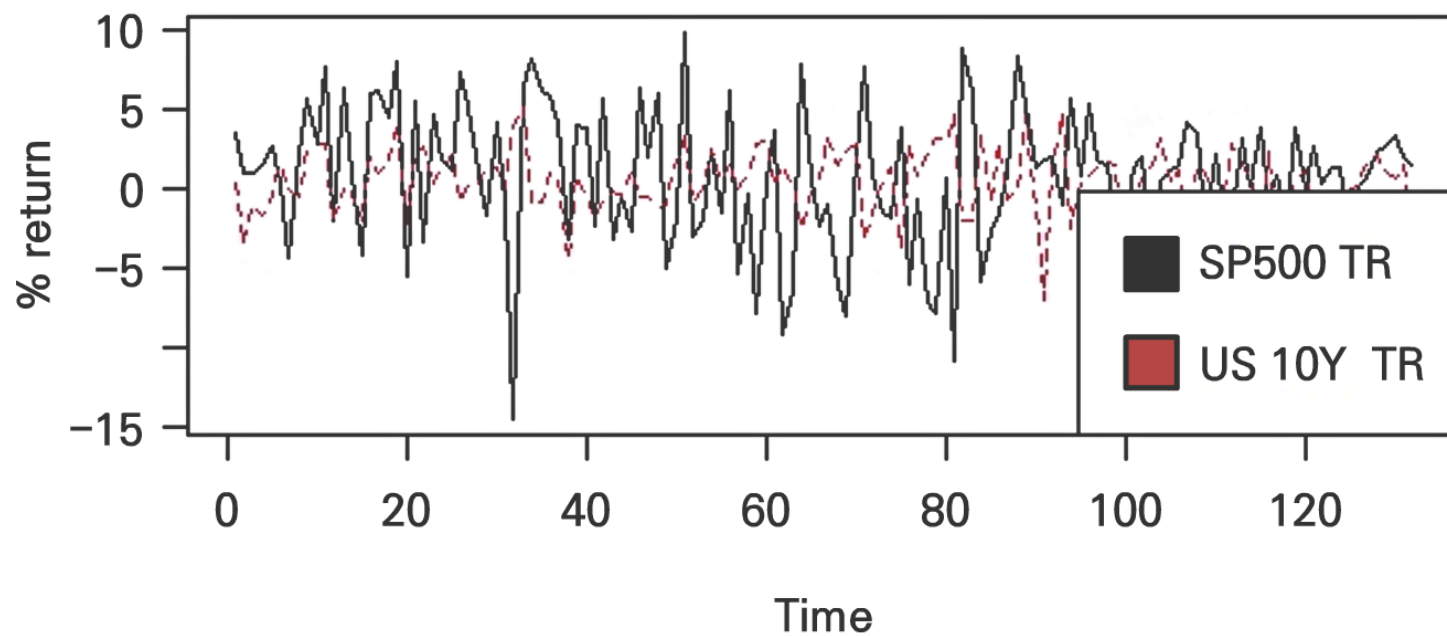


## ◆ Multivariate Plot

```
sp500.10yrT.ret <- managers[,8:9]*100
matplot(sp500.10yrT.ret,
        main="S&P 500 vs. 10-year Treasury",
        ylab="% return",
        xlab="time",
        type='l')
legend("bottomright", colnames(sp500.10yrT.ret),
      col=seq_len(ncol(sp500.10yrT.ret))
      ,cex=0.8,fill=seq_len(sp500.10yrT.ret))
```

## ◆ Multivariate Plot

S&P 500 vs. 10-year Trasury





## ◆ Beta Estimation

### ● Single index model

$$r_{i,t} - r_f = \alpha_i + \beta_i[r_{m,t} - r_f] + \epsilon_{i,t}$$

### ● CAPM

$$E(r_i) = r_f + \beta_i[E(r_m) - r_f]$$

### ● CAPM empirical model: same as single index model

```
ri <- managers$EDHEC  
rm <- managers$SP500
```

### ● ## assume that the riskfree rate is zero

```
reg = lm(ri ~ rm)  
summary(reg)
```

## ◆ Result

```
> reg = lm(ri ~ rm)
> summary(reg)

Call:
lm(formula = ri ~ rm)

Residuals:
    Min       1Q   Median       3Q      Max
-0.039160 -0.008253 -0.000801  0.006788  0.069297

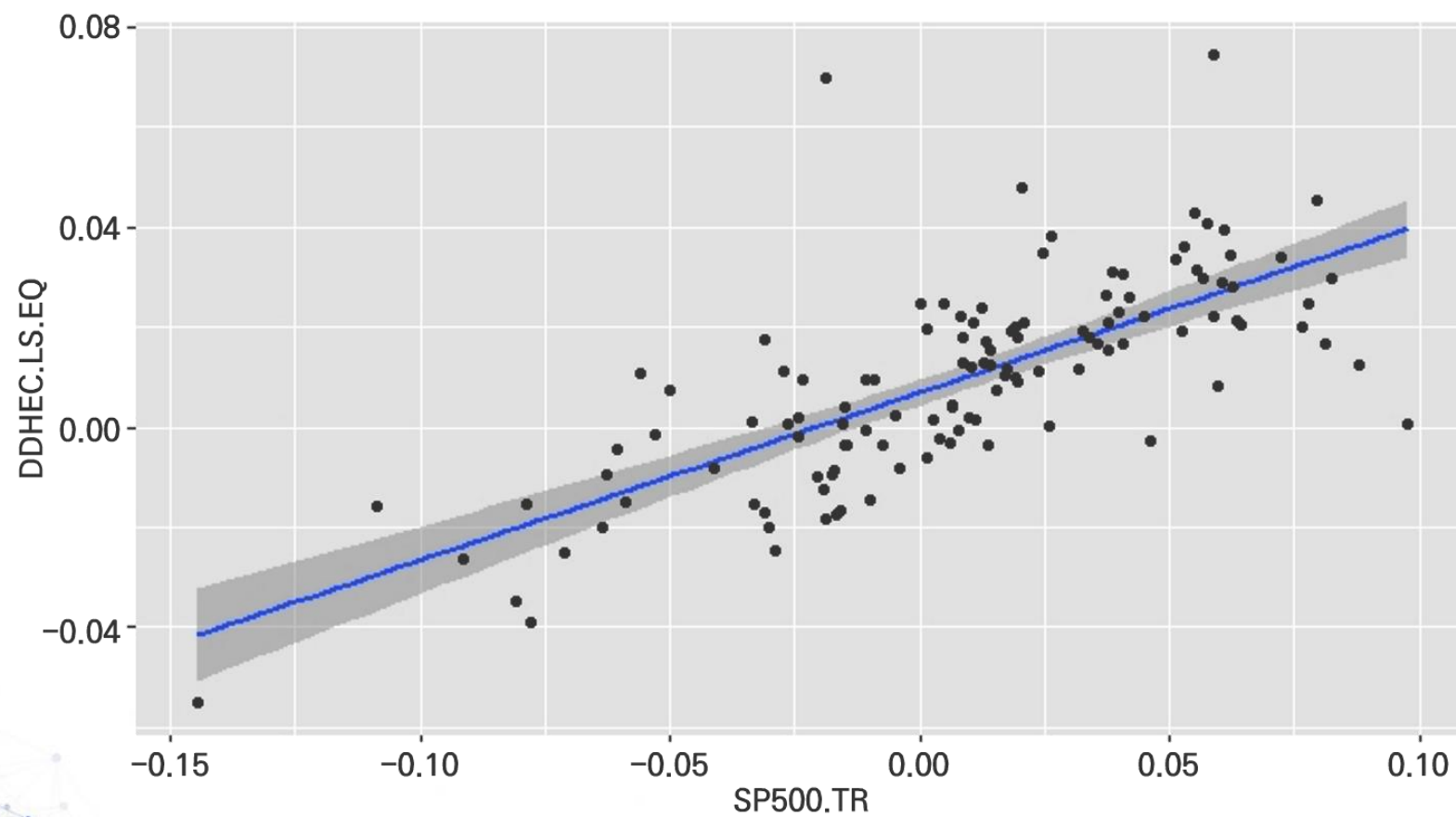
Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.006944   0.001307   5.314 5.14e-07 ***
rm           0.335542   0.029164  11.505 < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.0141 on 118 degrees of freedom
(결측으로 인하여 12개의 관측치가 삭제되었습니다.)
Multiple R-squared:  0.5287,    Adjusted R-squared:  0.5247
F-statistic: 132.4 on 1 and 118 DF,  p-value: < 2.2e-16
```

## ◆ Result

```
testdata <- cbind(rm, ri)
install.packages("ggpubr")
library(ggpubr)
g <- ggplot(data=testdata, aes(x = SP500.TR, y =
EDHEC.LS.EQ)) +
  geom_smooth(method="lm") +
  geom_point() +
  stat_regline_equation(label.x=-0.1, label.y=0.05)
+
  stat_cor(aes(label=..rr.label..), label.x=-0.1,
label.y=0.03)
```

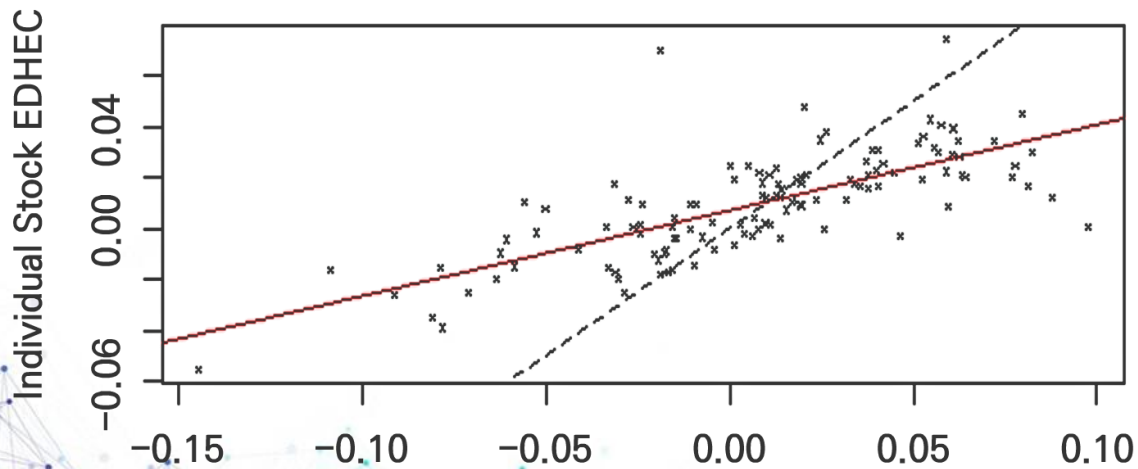
## ◆ Result





## ◆ Another Plotting

```
plot(as.numeric(rm),  
     as.numeric(ri), pch=4, cex=0.3,  
     xlab = "S&P 500",  
     ylab="Individual Stock EDHEC")  
abline(a=0, b=1, lty=2)  
abline(reg, col='red')
```



## ◆ Mean-Variance Analysis

- MV analysis is based on optimization mathematics: maximize expected return & minimize variance
- There are many different optimization technique in R, just like other programming languages. Or you can just write up an optimization function with a programming language that you like to use, i.e., Julia, Matlab, C++, etc.
- However, there are some packages or libraries that we can easily use without worries for most problems.
- In R, we can probably use `slsqp()` from `nloptr` package, `solve.QP()` from `quadprog` package for fundamental computation.
- Or one can use `optimalPortfolio()` from `RiskPortfolios` package.

## ◆ What package?

- We first use widely used RiskPortfolios package and then proceed to the package, IntroCompFinR, that is the main package that we use for the whole MV analysis

## ◆ Import Data

```
individual_ret <- read.csv("KSE_listed_returns.csv")
mkt_ret <- read.csv("kospi_ret_all.csv")
colnames(individual_ret) <- individual_ret[1,]
individual_ret <- individual_ret[-1, ]
individual_ret[,2] <- as.Date(individual_ret[,2])
riskfree <- read_excel("KTreasury_3yr.xlsx")
riskfree <- transpose(riskfree[,5:ncol(riskfree)])
# Note that riskfree rate for 2022 Sep. does not exist yet
```



## ◆ Data Management

### ● Do this

```
class(individual_ret), class(individual_ret[,10])
```

### ● Convert individual\_ret to numerical values

```
colchange <- c(3:ncol(individual_ret))  
for (i in colchange){  
  individual_ret[,i] <- as.numeric(individual_ret[,i])  
}
```

## ◆ MV Analysis with RiskPortfolios

```
library(PortfolioAnalytics)
library(RiskPortfolios)

• # we randomly select 5 individual stocks → 5 can vary

set.seed(80) ## For the reproducibility
picker <- floor(runif(6,3,ncol(individual_ret)))
stocks_ret <- individual_ret[,picker]
stocks_ret1 <- stocks_ret[ ,
colSums(is.na(stocks_ret))==0]
```

## ◆ Construct a Risky Stock Universe

```
risky_stocks <-  
cbind(individual_ret[,2],stocks_ret1,mkt_ret[,3])  
colnames(risky_stocks) <- c("Dates",  
colnames(stocks_ret1), "KOSPI Ret")  
stockdates <- risky_stocks[,1]  
risky_stocks <-  
risky_stocks[,2:ncol(risky_stocks)]/100  
Note: we convert % numbers into decimal
```

- Note: we convert % numbers into decimal

## ◆ Function: optimalPortfolio

```
covmat <- cov(risky_stocks) # create cov-var  
matrix  
w_mvp <- optimalPortfolio(cov.mat,  
                           control = list(type =  
'minvol')) %>%  
  round(., 4) %>%  
  setNames(colnames(risky_stocks))  
print(w_mvp)
```

```
> print(w_mvp)
```

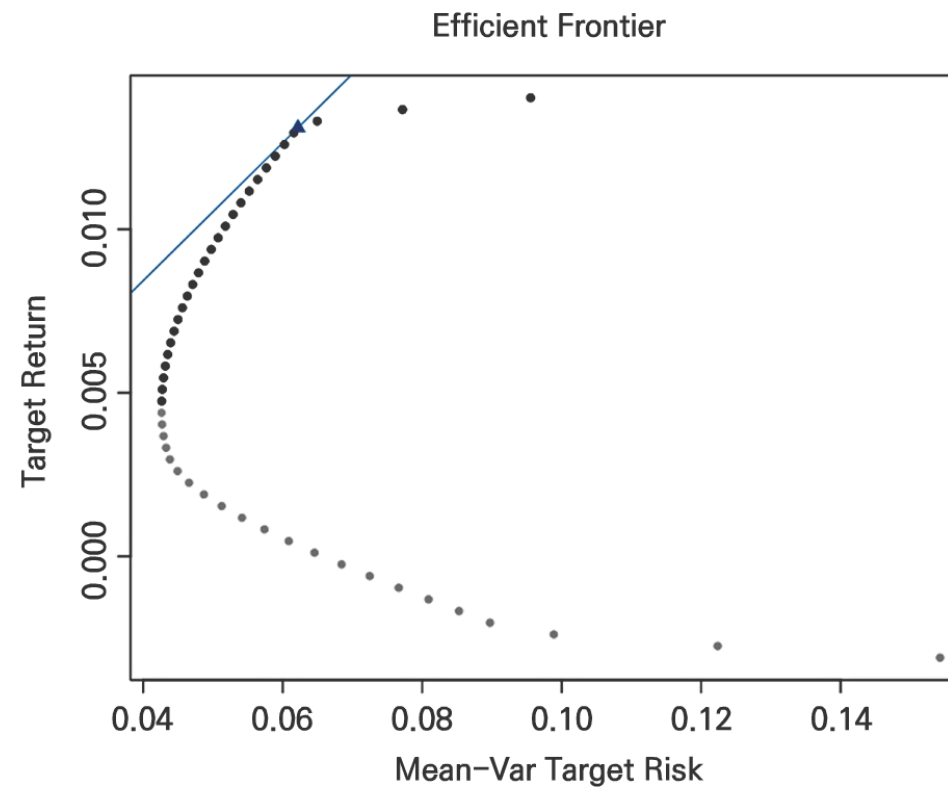
KSS해운	웅진	LG디스플레이	삼성SDI	KOSPI	Ret
0.1335	-0.0077	-0.0699	0.0220		0.9222



## ◆ Efficient Frontier

```
tp <-  
tangencyPortfolio(as.timeSeries  
(risky_stocks))  
frontier <-  
portfolioFrontier(as.timeSeries  
(risky_stocks))  
plot(frontier)
```

### ● What is Caveat?



## ◆ Package: IntroCompFinR

```
install.packages("IntroCompFinR", repos="http://R-  
Forge.R-project.org")  
library(IntroCompFinR)
```

### ⦿ Details of the Package

- # -- getPortfolio: create portfolio object
- # -- globalMin.portfolio: compute global minimum variance portfolio
- # -- efficient.portfolio: compute mvp subject to target return
- # -- tangency.portfolio: compute tangency portfolio
- # -- efficient.frontier: compute efficient frontier of risky assets

## ◆ Package: IntroCompFinR

```
er <- colMeans(risky_stocks) # average stock returns -- E(r)
rfree <- mean(riskfree[,1]/1200) # r_f -- set risk free asset: monthly and
decimal
ew <- rep(1,ncol(risky_stocks))/ncol(risky_stocks) # equal weights portfolio
equalWeights.portfolio = getPortfolio(er=er, cov.mat=covmat, weights=ew)
class(equalWeights.portfolio)
```

```
> equalWeights.portfolio
```

Call:

```
getPortfolio(er = er, cov.mat = covmat, weights = ew)
```

Portfolio expected return: 0.004885229

Portfolio standard deviation: 0.06268657

Portfolio weights:

KSS해운	웅진	LG디스플레이	삼성SDI	KOSPI	Ret
0.2	0.2	0.2	0.2	0.2	

## ◆ GMVP

```
gmin.port <- globalMin.portfolio(er, covmat)
gmin.port
```

```
> gmin.port
```

```
Call:
```

```
globalMin.portfolio(er = er, cov.mat = covmat)
```

```
Portfolio expected return: 0.004973064
```

```
Portfolio standard deviation: 0.04221948
```

```
Portfolio weights:
```

KSS해운	웅진	LG디스플레이	삼성SDI	KOSPI Ret
0.1335	-0.0077	-0.0699	0.0220	0.9222



## ◆ Efficient Portfolio with Target Return

```
target.ret <- er[1]
eff.port <- efficient.portfolio(er, covmat, target.ret)
eff.port
```

```
> eff.port
```

Call:

```
efficient.portfolio(er = er, cov.mat = covmat, target.return = target.ret)
```

Portfolio expected return: 0.0125

Portfolio standard deviation: 0.05581672

Portfolio weights:

KSS해운	웅진	LG디스플레이	삼성SDI	KOSPI	Ret
0.4949	-0.0448	-0.2262	0.3085		0.4676

## ◆ Tangent Portfolio

```
tan.port <- tangency.portfolio(er, covmat, rfree)
tan.port
```

```
> tan.port
```

Call:

```
tangency.portfolio(er = er, cov.mat = covmat, risk.free = rfree)
```

Portfolio expected return: 0.02957863

Portfolio standard deviation: 0.1266004

Portfolio weights:

KSS해운	웅진	LG디스플레이	삼성SDI	KOSPI	Ret
1.3150	-0.1289	-0.5807	0.9586		-0.5640

## ◆ Efficient Frontier

```
ef <- efficient.frontier(er, covmat, alpha.min=-2,
                        alpha.max=1.5, nport=30)
```

```
ef
```

```
> ef
```

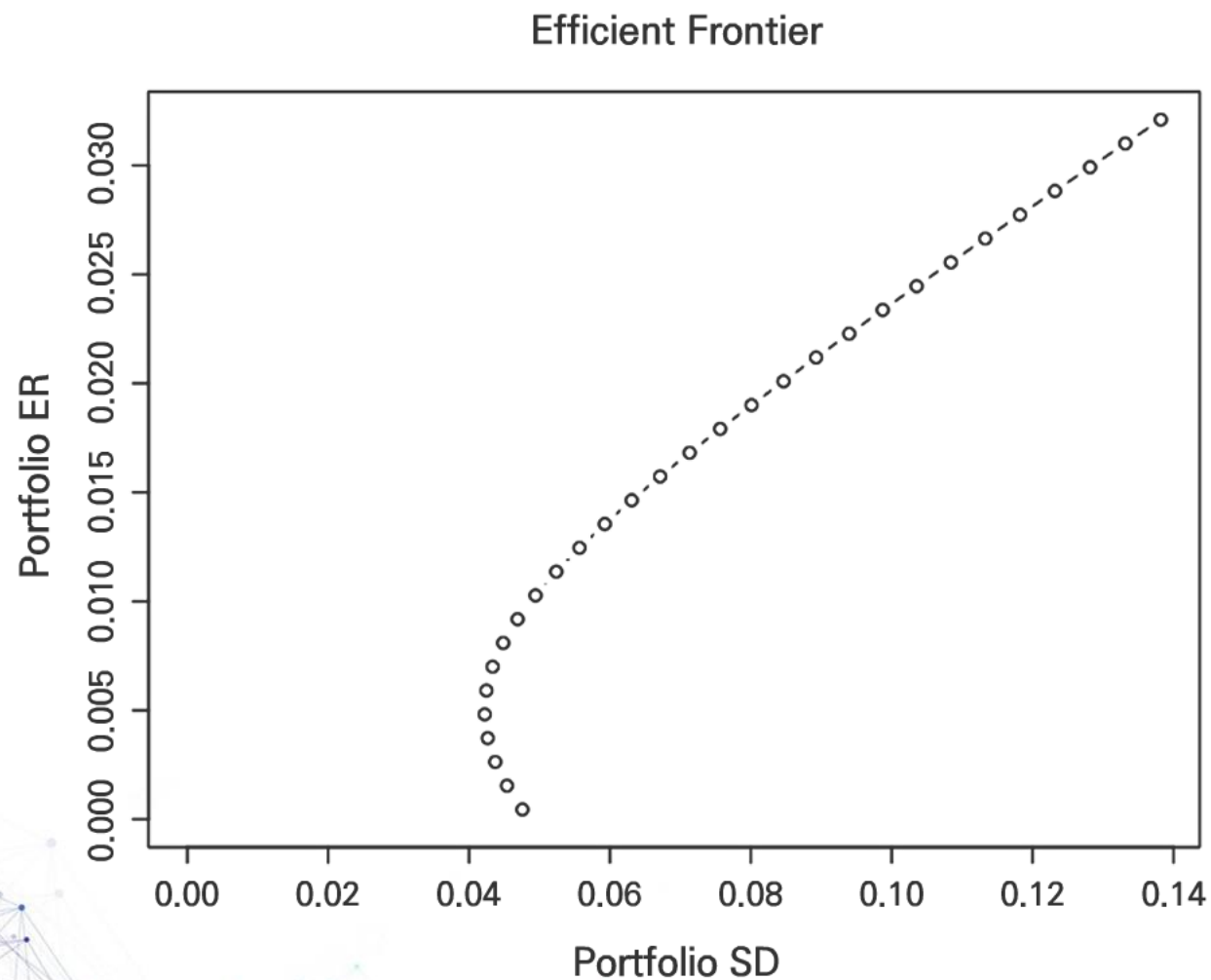
```
Call:
```

```
efficient.frontier(er = er, cov.mat = covmat, nport = 30, alpha.min = -2,
                  alpha.max = 1.5)
```

Frontier portfolios' expected returns and standard deviations

	port 1	port 2	port 3	port 4	port 5	port 6	port 7	port 8	port 9	port 10	port 11	port 12
ER	0.0321	0.0310	0.0299	0.0288	0.0277	0.0266	0.0256	0.0245	0.0234	0.0223	0.0212	0.0201
SD	0.1382	0.1332	0.1282	0.1232	0.1182	0.1133	0.1084	0.1035	0.0987	0.0940	0.0893	0.0846
	port 13	port 14	port 15	port 16	port 17	port 18	port 19	port 20	port 21	port 22	port 23	
ER	0.0190	0.0179	0.0168	0.0157	0.0146	0.0135	0.0125	0.0114	0.0103	0.0092	0.0081	
SD	0.0801	0.0757	0.0713	0.0671	0.0631	0.0593	0.0557	0.0524	0.0494	0.0469	0.0448	
	port 24	port 25	port 26	port 27	port 28	port 29	port 30					
ER	0.0070	0.0059	0.0048	0.0037	0.0026	0.0015	0.0005					
SD	0.0433	0.0425	0.0422	0.0427	0.0437	0.0454	0.0476					

## ◆ Efficient Frontier



## ◆ Adjustment with the Plot

```
par(family="AppleGothic") # Dealing with broken  
Korean fonts on Mac  
plot(ef, plot.assets=T)  
points(gmin.port$sd, gmin.port$er, col="blue")  
points(tan.port$sd, tan.port$er, col="red")  
sr.tan = (tan.port$er - rfree)/tan.port$sd  
abline(a=rfree, b=sr.tan)
```

## ◆ Adjustment with the Plot

