

## MF702 PROBLEM SET 6: SOLUTION

1. Consider  $M$  assets whose return is a  $M$ -dimensional random vector  $r$ . The expected value of  $r$  is a  $M$ -dimensional column vector  $\vec{r}$  and the covariance matrix is  $\Sigma$ .

(a) The Maximum Sharpe Ratio portfolio  $\Delta_{MSR}$  solves

$$\max_{\Delta} \frac{\Delta^\top \vec{r}}{\sqrt{\Delta^\top \Sigma \Delta}}, \quad \text{s.t. } \Delta^\top \vec{1} = 1.$$

Use the Lagrange multiplier method to show that  $\Delta_{MSR} = \frac{\Sigma^{-1} \vec{r}}{\vec{1}^\top \Sigma^{-1} \vec{r}}$ .

(b) The Global Minimum Variance portfolio  $\Delta_{GMV}$  solves

$$\min_{\Delta} \Delta^\top \Sigma \Delta, \quad \text{s.t. } \Delta^\top \vec{1} = 1.$$

Show that  $\Delta_{GMV} = \frac{\Sigma^{-1} \vec{1}}{\vec{1}^\top \Sigma^{-1} \vec{1}}$ .

**Solution:**

(a) Let us consider the unconstrained problem

$$\max_{\Delta, \lambda} \frac{\Delta^\top \vec{r}}{\sqrt{\Delta^\top \Sigma \Delta}} - \lambda (\Delta^\top \vec{1} - 1).$$

Take the first order condition, we obtain

$$\begin{aligned} \frac{\vec{r}}{(\Delta^\top \Sigma \Delta)^{\frac{1}{2}}} - \frac{\Delta^\top \vec{r} \Sigma \Delta}{(\Delta^\top \Sigma \Delta)^{\frac{3}{2}}} &= \lambda \vec{1}, \\ \Delta^\top \vec{1} &= 1. \end{aligned}$$

From the first equation above, left multiply by  $\Delta^\top$ , we observe the right-hand side is zero, which means  $\lambda = 0$ . Then the first equation yields

$$\vec{r} = \frac{\Delta^\top \vec{r}}{\Delta^\top \Sigma \Delta} \Sigma \Delta.$$

Note that  $\frac{\Delta^\top \vec{r}}{\Delta^\top \Sigma \Delta}$  is a constant, rename it as  $c$ . Then  $\vec{r} = c \Sigma \Delta$ , which implies

$$\Delta = \frac{1}{c} \Sigma^{-1} \vec{r}.$$

To satisfy  $\Delta^\top \vec{1} = 1$ , we need  $c = \vec{1}^\top \Sigma^{-1} \vec{r}$ . Therefore

$$\Delta = \frac{\Sigma^{-1} \vec{r}}{\vec{1}^\top \Sigma^{-1} \vec{r}}.$$

(b) Consider the unconstrained problem

$$\min_{\Delta, \lambda} \Delta^\top \Sigma \Delta + \lambda (\Delta^\top \vec{1} - 1).$$

Take the first order condition, we obtain

$$\begin{aligned} 2\Sigma\Delta &= -\lambda\vec{1}, \\ \Delta^\top\vec{1} &= 1. \end{aligned}$$

The first equation implies

$$\Delta = -\lambda\Sigma^{-1}\vec{1}.$$

In order to satisfy the second equation, we need  $-\lambda = ((\vec{1})^\top\Sigma^{-1}\vec{1})^{-1}$ . Therefore

$$\Delta = \frac{\Sigma^{-1}\vec{1}}{\vec{1}^\top\Sigma^{-1}\vec{1}}.$$

**2.** You can invest in three assets with expected returns and return covariance matrix given by

$$\vec{r} = \begin{pmatrix} 0.10 \\ 0.09 \\ 0.16 \end{pmatrix}, \quad \Sigma = \begin{pmatrix} 0.09 & -0.03 & 0.084 \\ -0.03 & 0.04 & 0.012 \\ 0.084 & 0.012 & 0.16 \end{pmatrix}.$$

You can also invest in the risk-free asset with return  $r_f = 0.02$ .

- (a) What is the market portfolio ( $\Delta_{MSR}$ ) in this setting? Compute the optimal weights invested in each risky security for this portfolio.
- (b) Find the minimum-volatility portfolio which achieves an expected return of  $\bar{r} = 0.20$ .

**Solution:**

(a) The market portfolio is the Maximum Sharpe Ratio portfolio with respect to the excess return, and takes the form

$$\Delta_{MSR,e} = \frac{\Sigma^{-1}\vec{r}_e}{\vec{1}^\top\Sigma^{-1}\vec{r}_e}.$$

Using the given values, a direct calculation gives

$$\Delta_{MSR,e} = \begin{pmatrix} 0.5603 \\ 0.6886 \\ -0.2489 \end{pmatrix}.$$

(b) Using the above values, the excess return of the market portfolio is

$$\Delta_{MSR,e}^\top\vec{r}_e = 0.058178,$$

while the target excess return is  $\bar{r}_e = \bar{r} - r_f = 0.18$ . Thus,

$$\alpha^*(\bar{r}) = \frac{\bar{r}_e}{\Delta_{MSR,e}^\top\vec{r}_e} = 3.09395.$$

Therefore the minimum volatility portfolio which achieves the target return is

$$\Delta^*(\bar{r}) = \alpha^*(\bar{r})\Delta_{MSR,e} = 3.0935 \begin{pmatrix} 0.5603 \\ 0.6886 \\ -0.2489 \end{pmatrix} = \begin{pmatrix} 1.7335 \\ 2.1306 \\ -0.7702 \end{pmatrix}.$$

The residual fraction  $1 - \alpha^*(\bar{r}) = -0.209395$  is put into the ZCB. Thus we are highly leveraged in the market portfolio, with a significant short position in the ZCB.

**3. (ETF construction)** Often we wish to track an index  $I$ , but for practical reasons (such as wishing to avoid transactions costs) we cannot invest in all the securities which comprise  $I$ . In this setting, we wish to construct a portfolio that mimics the index most closely, in the sense of minimizing the tracking error, while still maintaining a target level of return.

More precisely, suppose we may invest in a subset of the index, comprising of  $M$  stocks. The stocks have (random) returns  $r_1, \dots, r_M$  with expected values  $\vec{r} = (\bar{r}_1, \dots, \bar{r}_M)$  and covariance matrix  $\Sigma = \{\Sigma_{mk}\}_{m,k=1}^M$  where  $\Sigma_{mk} = Cov(r_m, r_k)$ . On the other hand, the index  $I$  has random return  $r_I$  with expected value  $\bar{r}_I$  and variance  $\sigma_I^2$ . We are able to estimate the covariance of the index and stock returns, obtaining the vector  $\Upsilon = \{\Upsilon_m\}_{m=1}^M$ ,  $\Upsilon_m = Cov(r_m, r_I)$ . Lastly, we assume the investor may not allocate money to the ZCB.

We wish to find the portfolio weights  $\Delta = (\Delta_1, \dots, \Delta_M)$  whose random return  $r(\Delta)$  minimize

$$Var(r(\Delta) - r_I), \tag{1}$$

still yielding an expected return of  $\bar{r}(\Delta) = \bar{r}$ , where  $\bar{r}$  is the target return.

(a) Consider the (related) problem of trying to find the portfolio which is most closely correlated with the index, but has no opinion about the expected return: i.e.

$$\max_{\Delta} Corr(r(\Delta), r_I), \quad \text{s.t. } \Delta^\top \vec{1} = 1.$$

Explicitly identify the optimal portfolio  $\Delta_{Corr,I}$ .

(b) Now, come back to the problem in (1):

$$Var(r(\Delta) - r_I), \quad \text{s.t. } \bar{r}(\Delta) = \bar{r}, \Delta^\top \vec{1} = 1.$$

Show that the optimal portfolio admits the decomposition

$$\Delta^* = \alpha_{C,I} \Delta_{Corr,I} + \alpha_M \Delta_{MSR} + \alpha_G \Delta_{GMV}.$$

Explicitly identify the constants  $\alpha_{C,I}$ ,  $\alpha_M$ , and  $\alpha_G$ .

**Solution:**

(a) Note that

$$\text{Corr}(r(\Delta), r_I) = \frac{\text{Cov}(r(\Delta), r_I)}{\sigma_I \sqrt{\text{Var}(r(\Delta))}} = \frac{\Delta^\top \Upsilon}{\sigma_I \sqrt{\Delta^\top \Sigma \Delta}},$$

where the equalities follow since  $\text{Cov}(r(\Delta), r_I) = \sum_{m=1}^M \Delta_m \text{Cov}(r_m, r_I)$ , and because  $\text{Var}(r(\Delta)) = \Delta^\top \Sigma \Delta$ . Thus, as  $\sigma_I$  plays no role, we see that the problem of finding the maximal correlated portfolio is, up to the notational difference of having  $\Upsilon$  instead of  $\vec{r}$ , the same as Problem 1 (a). Therefore, we already know the answer is

$$\Delta_{\text{corr}, I} = \frac{\Sigma^{-1} \Upsilon}{\vec{1}^\top \Sigma^{-1} \Upsilon}.$$

(b) First, note that

$$\text{Var}(r(\Delta) - r_I) = \text{Var}(r(\Delta)) + \sigma_I^2 - 2\text{Cov}(r(\Delta), r_I) = \Delta^\top \Sigma \Delta + \sigma_I^2 - 2\Delta^\top \Upsilon.$$

Thus, using Lagrange multipliers, and noting that the term  $\sigma_I^2$  does not depend on  $\Delta$ , the unconstrained problem is

$$\min_{\Delta, \lambda_1, \lambda_2} \Delta^\top \Sigma \Delta - 2\Delta^\top \Upsilon - \lambda_1(\Delta^\top \vec{r} - \bar{r}) - \lambda_2(\Delta^\top \vec{1} - 1).$$

The first order conditions with respect to  $\Delta$  is

$$\begin{aligned} \Delta^* &= \Sigma^{-1} \Upsilon + \frac{\lambda_1}{2} \Sigma^{-1} \vec{r} + \frac{\lambda_2}{2} \Sigma^{-1} \vec{1} \\ &= \alpha_{C,I} \Delta_{\text{corr}, I} + \alpha_M \Delta_{\text{MSR}} + \alpha_G \Delta_{\text{GMV}}, \end{aligned}$$

where

$$\alpha_{C,I} = \vec{1}^\top \Sigma^{-1} \Upsilon, \quad \alpha_M = \frac{\lambda_1}{2} \vec{1}^\top \Sigma^{-1} \vec{r}, \quad \alpha_G = \frac{\lambda_2}{2} \vec{1}^\top \Sigma^{-1} \vec{1}.$$

As  $\alpha_{C,I}$  is explicitly known, it remains to find  $\alpha_M, \alpha_G$ , which enforce the target return and fully invested restrictions:

$$1 = \alpha_{C,I} + \alpha_M + \alpha_G,$$

$$\bar{r} = \alpha_{C,I} \bar{r}_{\text{corr}, I} + \alpha_M \bar{r}_{\text{MSR}} + \alpha_G \bar{r}_{\text{GMV}}.$$

From above equations, we can solve for

$$\begin{aligned} \alpha_M &= \frac{\bar{r} - \bar{r}_{\text{GMV}} - \alpha_{C,I}(\bar{r}_{\text{corr}, I} - \bar{r}_{\text{GMV}})}{\bar{r}_{\text{MSR}} - \bar{r}_{\text{GMV}}}, \\ \alpha_G &= \frac{\bar{r}_{\text{MSR}} - \bar{r} - \alpha_{C,I}(\bar{r}_{\text{MSR}} - \bar{r}_{\text{corr}, I})}{\bar{r}_{\text{MSR}} - \bar{r}_{\text{GMV}}}. \end{aligned}$$