

# Chapter 5

## Mean-Variance Analysis

**5.1.** Suppose there are two risky assets with means  $\mu_1 = 1.08$ ,  $\mu_2 = 1.16$ , standard deviations  $\sigma_1 = 0.25$ ,  $\sigma_2 = 0.35$ , and correlation  $\rho = 0.30$ . Calculate the GMV portfolio and locate it on Figure 5.1.

**Solution:** The GMV portfolio is

$$\pi = \frac{1}{\nu' \Sigma^{-1} \nu} \Sigma^{-1} \nu.$$

Substituting

$$\Sigma = \begin{pmatrix} 0.0625 & 0.02625 \\ 0.02625 & 0.1225 \end{pmatrix},$$

we obtain

$$\pi = \begin{pmatrix} 0.7264 \\ 0.2736 \end{pmatrix}.$$

Therefore, the mean and standard deviation of the GMV portfolio are  $\mu_{gmv} = \mu' \pi = 1.1019$  and  $\sigma_{gmv} = \sqrt{\pi' \Sigma \pi} = 0.2293$ . This plots as the point that is furthest to the left on the hyperbola in Figure 5.1.

**5.2.** Assume there is a risk-free asset. Consider an investor with quadratic utility  $-(\tilde{w} - \xi)^2/2$ , and no labor income.

(a) Explain why the result of Exercise 2.5 implies that the investor will choose a portfolio on the mean-variance frontier.

**Solution:** From Exercise 2.5, the optimal portfolio is

$$\phi = \frac{\kappa^2}{1 + \kappa^2}(\zeta - w_0 R_f) \Sigma^{-1}(\mu - R_f \iota).$$

This is proportional to  $\Sigma^{-1}(\mu - R_f \iota)$  and hence is on the mean-variance frontier.

(b) Under what circumstances will the investor choose a mean-variance efficient portfolio? Explain the economics of the condition you derive.

**Solution:** The frontier portfolios are scalar multiples of the vector  $\Sigma^{-1}(\mu - R_f \iota)$ . See (5.15). The positive scalar multiples are efficient (because they have  $\mu_{\text{targ}} > R_f$ ), and the negative scalar multiples are inefficient. Therefore, when  $\zeta > w_0 R_f$ , the optimal portfolio for the quadratic utility investor is on the efficient part of the frontier, and when  $\zeta < w_0 R_f$ , the optimal portfolio is on the inefficient part of the frontier.  $\zeta$  is the bliss level of wealth for the quadratic utility function. When  $\zeta < w_0 R_f$ , the investor can exceed the bliss level by simply holding the risk-free asset. Thus, higher returns can lower utility, so the investor holds an inefficient portfolio of risky assets.

(c) Re-derive the answer to Part (b) using the orthogonal projection characterization of the quadratic utility investor's optimal portfolio presented in Section ??.

**Solution:** Given that there is no labor income,  $\tilde{y}_p$  in (3.42) is zero. Also, given that there is a risk-free asset,  $\zeta_p = \zeta$  and  $E[\tilde{m}_p \zeta_p] = \zeta E[\tilde{m}_p] = \zeta / R_f$ . Therefore, (3.42) implies

$$\tilde{x} = \zeta - (\zeta / R_f - w_0) \tilde{R}_p.$$

The return  $\tilde{R}_p$  is on the inefficient part of the frontier, so the return producing  $\tilde{x}$  is on the efficient part of the frontier if and only if  $\zeta/R_f - w_0 > 0$ .

**5.3.** Suppose that the risk-free return is equal to the expected return of the GMV portfolio ( $R_f = B/C$ ). Show that there is no tangency portfolio.

Hint: Show there is no  $\delta$  and  $\lambda$  satisfying

$$\delta\Sigma^{-1}(\mu - R_f\iota) = \lambda\pi_{\text{mu}} + (1 - \lambda)\pi_{\text{gmv}}.$$

Recall that we are assuming  $\mu$  is not a scalar multiple of  $\iota$ .

**Solution:** The mean-variance frontier considering only the risky assets is the set  $\lambda\pi_\mu + (1 - \lambda)\pi_\iota$  for some  $\lambda$ , and the mean-variance frontier including the risk-free asset is the set  $\delta\Sigma^{-1}(\mu - R_f\iota)$  for some  $\delta$ . For the frontiers to intersect, we must have

$$\delta\Sigma^{-1}(\mu - R_f\iota) = \lambda\pi_\mu + (1 - \lambda)\pi_\iota.$$

This is equivalent to

$$\left(\delta - \frac{\lambda}{\iota'\Sigma^{-1}\mu}\right)\Sigma^{-1}\mu = \left(\delta R_f + \frac{1 - \lambda}{\iota'\Sigma^{-1}\iota}\right)\Sigma^{-1}\iota,$$

and premultiplying by  $\Sigma$  gives

$$\left(\delta - \frac{\lambda}{\iota'\Sigma^{-1}\mu}\right)\mu = \left(\delta R_f + \frac{1 - \lambda}{\iota'\Sigma^{-1}\iota}\right)\iota.$$

Because  $\mu$  is not proportional to  $\iota$ , this equation can hold only if

$$\delta - \frac{\lambda}{\iota'\Sigma^{-1}\mu} = \delta R_f + \frac{1 - \lambda}{\iota'\Sigma^{-1}\iota} = 0.$$

This implies

$$\frac{\lambda}{\iota'\Sigma^{-1}\mu}R_f + \frac{1 - \lambda}{\iota'\Sigma^{-1}\iota} = 0,$$

and substituting  $R_f = B/C = \iota' \Sigma^{-1} \mu / \iota' \Sigma^{-1} \iota$  yields

$$\frac{1}{\iota' \Sigma^{-1} \iota} = 0,$$

which is impossible.

**5.4.** Show that  $\mathbb{E}[\tilde{R}^2] \geq \mathbb{E}[\tilde{R}_p^2]$  for every return  $\tilde{R}$  (thus,  $\tilde{R}_p$  is the minimum second-moment return).

The returns having a given second moment  $a$  are the returns satisfying  $\mathbb{E}[\tilde{R}^2] = a$ , which is equivalent to

$$\text{var}(\tilde{R}) + \mathbb{E}[\tilde{R}]^2 = a;$$

thus, they plot on the circle  $x^2 + y^2 = a$  in (standard deviation, mean) space. Use the fact that  $\tilde{R}_p$  is the minimum second-moment return to illustrate graphically that  $\tilde{R}_p$  must be on the inefficient part of the frontier, with and without a risk-free asset (assuming  $\mathbb{E}[\tilde{R}_p] > 0$  in the absence of a risk-free asset).

**Solution:** Using Facts 1, 2 and 8,

$$\mathbb{E}[\tilde{R}^2] = \mathbb{E}[(\tilde{R}_p + b\tilde{e}_p + \tilde{\varepsilon})^2] = \mathbb{E}[\tilde{R}_p^2] + b^2 \mathbb{E}[\tilde{e}_p^2] + \mathbb{E}[\tilde{\varepsilon}^2] \geq \mathbb{E}[\tilde{R}_p^2].$$

With a risk-free asset, the cone intersects the vertical axis at  $R_f > 0$ , and the point on the cone closest to the origin is on the lower part. In the absence of a risk-free asset, the assumption  $\mathbb{E}[\tilde{R}_p] > 0$  implies that global minimum variance portfolio has a positive expected return (use the definition of  $b_m$  and Facts 16 and 17 — which imply  $1 - \mathbb{E}[\tilde{e}_p] > 0$  — to deduce this). Thus, the point on the hyperbola closest to the origin must be on the lower part of the hyperbola.

**5.5.** Write any return  $\tilde{R}$  as  $\tilde{R}_p + (\tilde{R} - \tilde{R}_p)$  and use the fact that  $1 - \tilde{e}_p$  is orthogonal to excess returns—because  $\tilde{e}_p$  represents the expectation operator on the space of excess returns—to show that

$$\tilde{x} \stackrel{\text{def}}{=} \frac{1}{\mathbb{E}[\tilde{R}_p]} (1 - \tilde{e}_p)$$

is an SDF. When there is a risk-free asset,  $\tilde{x}$ , being spanned by a constant and an excess return, is in the span of the returns and hence must equal  $\tilde{m}_p$ . Use this fact to demonstrate (??).

**Solution:** We have

$$\begin{aligned}\mathbb{E}[\tilde{x}\tilde{R}] &= \frac{1}{\mathbb{E}[\tilde{R}_p]}\mathbb{E}[(1-\tilde{e}_p)\tilde{R}_p] + \frac{1}{\mathbb{E}[\tilde{R}_p]}\mathbb{E}[(1-\tilde{e}_p)(\tilde{R}-\tilde{R}_p)] \\ &= \frac{1}{\mathbb{E}[\tilde{R}_p]}\mathbb{E}[(1-\tilde{e}_p)\tilde{R}_p] \\ &= 1,\end{aligned}$$

using  $\mathbb{E}[\tilde{R}-\tilde{R}_p] = \mathbb{E}[\tilde{e}_p(\tilde{R}-\tilde{R}_p)]$  for the second equality and Fact 8 for the third. Thus,  $\tilde{x}$  is an SDF. This implies

$$\frac{1}{R_f} = \mathbb{E}[\tilde{x}] = \frac{1 - \mathbb{E}[\tilde{e}_p]}{\mathbb{E}[\tilde{R}_p]}.$$

Moreover,  $\tilde{x} = \tilde{m}_p$  implies

$$\tilde{R}_p = \frac{\tilde{x}}{\mathbb{E}[\tilde{x}^2]},$$

and

$$\mathbb{E}[\tilde{x}^2] = \frac{1}{\mathbb{E}[\tilde{R}_p]^2}(1 - 2\mathbb{E}[\tilde{e}_p] + \mathbb{E}[\tilde{e}_p^2]) = \frac{1 - \mathbb{E}[\tilde{e}_p]}{\mathbb{E}[\tilde{R}_p]^2} = \frac{1}{R_f \mathbb{E}[\tilde{R}_p]},$$

using Fact 16 for the second equality. Thus,

$$\tilde{R}_p = R_f \mathbb{E}[\tilde{R}_p] \left( \frac{1}{\mathbb{E}[\tilde{R}_p]}(1 - \tilde{e}_p) \right) = R_f(1 - \tilde{e}_p).$$

**5.6.** Establish the properties claimed for the risk-free return proxies:

(a) Show that  $\text{var}(\tilde{R}) \geq \text{var}(\tilde{R}_p + b_m \tilde{e}_p)$  for every return  $\tilde{R}$ .

**Solution:** By Fact 15, the minimum variance return is  $\tilde{R}_p + b\tilde{e}_p$  for some  $b$ . Using Fact 8, we have

$$\text{var}(\tilde{R}_p + b\tilde{e}_p) = \text{var}(\tilde{R}_p) - 2b\mathbb{E}[\tilde{R}_p]\mathbb{E}[\tilde{e}_p] + \text{var}(\tilde{e}_p),$$

and by Fact 17, this equals

$$\text{var}(\tilde{R}_p) + \left( b^2(1 - \mathbb{E}[\tilde{e}_p]) - 2b\mathbb{E}[\tilde{R}_p] \right) \mathbb{E}[\tilde{e}_p].$$

By Fact 16,  $\mathbb{E}[\tilde{e}_p] > 0$ , so the minimum variance return is found by minimizing  $(b^2(1 - \mathbb{E}[\tilde{e}_p]) - 2b\mathbb{E}[\tilde{R}_p])$  in  $b$ , with solution  $b = b_m$ .

(b) Show that  $\text{cov}(\tilde{R}_p, \tilde{R}_p + b_z \tilde{e}_p) = 0$ .

**Solution:** Using Fact 8, we have  $\text{cov}(\tilde{R}_p, \tilde{R}_p + b_z \tilde{e}_p) = \text{var}(\tilde{R}_p) - b_z \mathbb{E}[\tilde{R}_p] \mathbb{E}[\tilde{e}_p] = 0$ .

(c) Prove (??), showing that  $\tilde{R}_p + b_c \tilde{e}_p$  represents the constant  $b_c$  times the expectation operator on the space of returns.

**Solution:** Using Fact 11 and the definition of  $b_c$ , we have

$$b_c \mathbb{E}[\tilde{R}] = b_c \mathbb{E}[\tilde{R}_p + b \tilde{e}_p + \tilde{\varepsilon}] = \mathbb{E}[\tilde{R}_p^2] + b b_c \mathbb{E}[\tilde{e}_p].$$

From Facts 2, 8, 11, and 16,

$$\begin{aligned} \mathbb{E}[\tilde{R}(\tilde{R}_p + b_c \tilde{e}_p)] &= \mathbb{E}[(\tilde{R}_p + b \tilde{e}_p + \tilde{\varepsilon})(\tilde{R}_p + b_c \tilde{e}_p)] \\ &= \mathbb{E}[\tilde{R}_p^2] + b b_c \mathbb{E}[\tilde{e}_p^2] \\ &= \mathbb{E}[\tilde{R}_p^2] + b b_c \mathbb{E}[\tilde{e}_p]. \end{aligned}$$

Thus,

$$b_c \mathbb{E}[\tilde{R}] = \mathbb{E}[\tilde{R}(\tilde{R}_p + b_c \tilde{e}_p)].$$

**5.7.** If all returns are joint normally distributed, then  $\tilde{R}_p$ ,  $\tilde{e}_p$  and  $\tilde{\varepsilon}$  are joint normally distributed in the orthogonal decomposition  $\tilde{R} = \tilde{R}_p + b \tilde{e}_p + \tilde{\varepsilon}$  of any return  $\tilde{R}$  (because  $\tilde{R}_p$  is a return and  $\tilde{e}_p$  and  $\tilde{\varepsilon}$  are excess returns). Assuming all returns are joint normally distributed, use the orthogonal decomposition to compute the optimal return for a CARA investor.

**Solution:** When returns are normally distributed, a CARA investor chooses the return  $\tilde{R}$  that maximizes

$$\mathbb{E}[\tilde{R}] - \frac{1}{2}\alpha w_0 \text{var}(\tilde{R}).$$

Given  $\tilde{R} = \tilde{R}_p + b\tilde{e}_p + \tilde{\varepsilon}$  and Facts 11 and 15, the objective function is

$$\mathbb{E}[\tilde{R} + b\tilde{e}_p] - \frac{1}{2}\alpha w_0 [\text{var}(\tilde{R}_p + b\tilde{e}_p) + \text{var}(\tilde{\varepsilon})],$$

so it is optimal to choose  $\tilde{\varepsilon} = 0$ . The investor chooses  $b$  to maximize

$$b\mathbb{E}[\tilde{e}_p] - \frac{1}{2}\alpha w_0 [2b \text{cov}(\tilde{R}_p, \tilde{e}_p) + b^2 \text{var}(\tilde{e}_p)],$$

and the optimum satisfies

$$\mathbb{E}[\tilde{e}_p] - \alpha w_0 \text{cov}(\tilde{R}_p, \tilde{e}_p) - \alpha w_0 \text{var}(\tilde{e}_p)b = 0,$$

implying

$$b = \frac{\mathbb{E}[\tilde{e}_p]}{\alpha w_0 \text{var}(\tilde{e}_p)} - \frac{\text{cov}(\tilde{R}_p, \tilde{e}_p)}{\text{var}(\tilde{e}_p)}.$$

Using Facts 8 and 17, we can simplify this further to

$$b = \frac{1 + \alpha w_0 \mathbb{E}[\tilde{R}_p]}{\alpha w_0 (1 - \mathbb{E}[\tilde{e}_p])}.$$

**5.8.** Assume there is a risk-free asset.

(a) Using the formula (3.45) for  $\tilde{m}_p$ , compute  $\lambda$  such that

$$\tilde{R}_p = \lambda \pi'_{\text{tang}} \tilde{\mathbf{R}} + (1 - \lambda) R_f.$$

**Solution:** We have

$$\tilde{m}_p = \frac{1}{R_f} + \left( \iota - \frac{1}{R_f} \mu \right)' \Sigma^{-1} (\tilde{\mathbf{R}} - \mu).$$

Hence

$$\text{var}(\tilde{m}_p) = \frac{\kappa^2}{R_f^2},$$

where  $\kappa^2 = (R_f \boldsymbol{\iota} - \boldsymbol{\mu})' \boldsymbol{\Sigma}^{-1} (R_f \boldsymbol{\iota} - \boldsymbol{\mu})$  is the squared maximum Sharpe ratio. Because  $\mathbb{E}[\tilde{m}_p] = 1/R_f$ , this implies

$$\mathbb{E}[\tilde{m}_p^2] = \frac{1 + \kappa^2}{R_f^2}.$$

Therefore, by the definition  $\tilde{R}_p = \tilde{m}_p / \mathbb{E}[\tilde{m}_p^2]$ , we have

$$\tilde{R}_p = \frac{R_f}{1 + \kappa^2} + \frac{R_f}{1 + \kappa^2} (R_f \boldsymbol{\iota} - \boldsymbol{\mu})' \boldsymbol{\Sigma}^{-1} (\tilde{\mathbf{R}} - \boldsymbol{\mu}),$$

in the notation of Section 5.2. Setting

$$\lambda = -\frac{R_f(B - R_f C)}{1 + \kappa^2},$$

we have

$$1 - \lambda = \frac{1 + \kappa^2 + R_f B - R_f^2 C}{1 + \kappa^2} = \frac{1 + A - R_f B}{1 + \kappa^2},$$

because  $\kappa^2 = A - 2R_f B + R_f^2 C$ . Thus,

$$\tilde{R}_p = \lambda \pi'_{\text{tang}} \tilde{\mathbf{R}} + (1 - \lambda) R_f.$$

(b) Show that  $\lambda$  in part (a) is negative when  $R_f < B/C$  and positive when  $R_f > B/C$ . Note: This shows that  $\tilde{R}_p$  is on the inefficient part of the frontier, because the portfolio generating  $\tilde{R}_p$  is short the tangency portfolio when the tangency portfolio is efficient and long the tangency portfolio when it is inefficient.

**Solution:**

$$\lambda = -\frac{R_f(B - R_f C)}{1 + \kappa^2} < 0$$

when  $B > R_f C$  and positive when  $B < R_f C$ .

**5.9.** Consider the problem of choosing a portfolio  $\pi$  of risky assets, a proportion  $\phi_b \geq 0$  to borrow and a proportion  $\phi_\ell \geq 0$  to lend to maximize the expected return  $\pi' \mu + \phi_\ell R_\ell - \phi_b R_b$  subject to the constraints  $(1/2)\pi' \Sigma \pi \leq k$  and  $\iota' \pi + \phi_\ell - \phi_b = 1$ . Assume  $B/C > R_b > R_\ell$ , where  $B$  and  $C$  are defined in (??). Define

$$\begin{aligned}\pi_b &= \frac{1}{\iota' \Sigma^{-1}(\mu - R_b \iota)} \Sigma^{-1}(\mu - R_b \iota), \\ \pi_\ell &= \frac{1}{\iota' \Sigma^{-1}(\mu - R_\ell \iota)} \Sigma^{-1}(\mu - R_\ell \iota).\end{aligned}$$

Using the Kuhn-Tucker conditions, show that the solution is either (i)  $\pi = (1 - \phi_\ell)\pi_\ell$  for  $0 \leq \phi_\ell \leq 1$ , (ii)  $\pi = \lambda\pi_\ell + (1 - \lambda)\pi_b$  for  $0 \leq \lambda \leq 1$ , or (iii)  $\pi = (1 + \phi_b)\pi_b$  for  $\phi_b \geq 0$ .

**Solution:** The Kuhn-Tucker conditions are

$$\mu - \delta \Sigma \pi - \gamma \iota = 0,$$

$$R_\ell - \gamma + \eta_\ell = 0,$$

$$-R_b + \gamma + \eta_b = 0,$$

$$\phi_\ell, \phi_b, \eta_\ell, \eta_b, \delta \geq 0,$$

$$\frac{1}{2} \pi' \Sigma \pi \leq k,$$

$$\iota' \pi + \phi_\ell - \phi_b = 1,$$

$$\eta_\ell \phi_\ell = \eta_b \phi_b = \delta \left( \frac{1}{2} \pi' \Sigma \pi - k \right) = 0.$$

There are three possibilities to consider: (i)  $\phi_\ell > 0$ , (ii),  $\phi_b > 0$ , (iii)  $\phi_\ell = \phi_b = 0$ .

(i) If  $\phi_\ell > 0$ , then  $\eta_\ell = 0$ ,  $\gamma = R_\ell$ , and

$$\pi = \frac{1}{\delta} \Sigma^{-1}(\mu - R_\ell \iota).$$

Also,  $\gamma = R_\ell$  implies  $\eta_b = R_b - R_\ell > 0$ . Hence,  $\phi_b = 0$ , and  $\iota' \pi = 1 - \phi_\ell$ . This implies  $\pi = (1 - \phi_\ell)\pi_\ell$ .

(ii) If  $\phi_b > 0$ , then  $\eta_b = 0$ ,  $\gamma = -R_b$ , and

$$\pi = \frac{1}{\delta} \Sigma^{-1} (\mu - R_b \iota).$$

Also,  $\gamma = -R_b$  implies  $\eta_\ell = R_b - R_\ell > 0$ , so  $\phi_\ell = 0$ . This implies  $\iota' \pi = 1 + \phi_b$ . Hence,  $\pi = (1 + \phi_b) \pi_b$ .

(iii) If  $\phi_\ell = \phi_b = 0$ , then

$$\pi = \frac{1}{\delta} \Sigma^{-1} (\mu - \gamma \iota),$$

where  $\gamma = R_\ell + \eta_\ell \geq R_\ell$  and  $\gamma = R_b - \eta_b \leq R_b$ . Thus,  $\gamma = \lambda R_\ell + (1 - \lambda) R_b$  for some  $0 \leq \lambda \leq 1$ .

From  $\iota' \pi = 1$ , it follows that  $\pi = \lambda \pi_\ell + (1 - \lambda) \pi_b$ .