有效前沿:给定预期回报率,使不含无风险资产组合方差最小

Theorem: As long as the covariance matrix of returns is non-singular, there is a mean-variance frontier

定理:只要收益率的协方差矩阵是非奇异的,就存在均值方差前沿 Notations:

$$\alpha = \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix}, \qquad R = E(\tilde{R}) = \begin{pmatrix} \mu_1 \\ \mu_2 \\ \vdots \\ \mu_n \end{pmatrix}, \quad \Omega = cov(\tilde{R}, \tilde{R}'), \quad I = \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix}_{(n,1)}$$

Model Construction:

假定有n个风险资产,给定投资组合的期望收益率为 μ_p 。

$$\min_{\alpha} \alpha' \Omega \alpha$$

s.t.

$$\begin{cases} \alpha' R = \mu_p \\ \alpha' I = 1 \end{cases}$$

Model Solving:

$$\mathcal{L} = \frac{1}{2}\alpha'\Omega\alpha + \lambda(\mu_p - \alpha'R) + \gamma(1 - \alpha'I)$$

$$\frac{\partial \mathcal{L}}{\partial \alpha} = \Omega \alpha - \lambda R - \gamma I = 0$$
$$\Rightarrow \alpha = \lambda \Omega^{-1} R + \gamma \Omega^{-1} I$$

$$\begin{split} \frac{\partial \mathcal{L}}{\partial \lambda} &= \mu_p - \alpha' R = 0 \\ &\Rightarrow \alpha' R = \mu_p \Rightarrow R' \alpha = \mu_p \\ &\Rightarrow R' (\lambda \Omega^{-1} R + \gamma \Omega^{-1} I) = \mu_p \\ &\Rightarrow \lambda R' \Omega^{-1} R + \gamma R' \Omega^{-1} I = \mu_n \end{split}$$

$$\frac{\partial \mathcal{L}}{\partial \gamma} = 1 - \alpha' \mathbf{I} = 0$$

$$\Rightarrow \alpha' \mathbf{I} = 1 \Rightarrow \mathbf{I}' \alpha = 1$$

$$\Rightarrow \mathbf{I}' (\lambda \Omega^{-1} R + \gamma \Omega^{-1} \mathbf{I}) = 1$$

$$\Rightarrow \lambda \mathbf{I}' \Omega^{-1} R + \gamma \mathbf{I}' \Omega^{-1} \mathbf{I} = 1$$

Define:

$$A = R'\Omega^{-1}R$$
, $B = I'\Omega^{-1}R$, $C = I'\Omega^{-1}I$

We can get:

$$\begin{cases} \frac{\partial \mathcal{L}}{\partial \alpha} = \lambda A + \gamma B = \mu_p \\ \frac{\partial \mathcal{L}}{\partial \alpha} = \lambda A B + \gamma B^2 = B \mu_p \\ A \frac{\partial \mathcal{L}}{\partial \gamma} = \lambda A B + \gamma A C = A \end{cases} \Rightarrow \begin{cases} A = \frac{\mu_p C - B}{AC - B^2} \\ A \frac{\partial \mathcal{L}}{\partial \gamma} = \lambda A B + \gamma A C = A \end{cases} \Rightarrow \begin{cases} \lambda = \frac{\mu_p C - B}{AC - B^2} \\ \gamma = \frac{A - B \mu_p}{AC - B^2} \end{cases}$$

$$\alpha_p = \alpha = \lambda \Omega^{-1} R + \gamma \Omega^{-1} I = \frac{\mu_p C - B}{AC - B^2} \Omega^{-1} R + \frac{A - B \mu_p}{AC - B^2} \Omega^{-1} I$$

由于 A、B、C 都可以通过历史数据获取,因此给定很多个投资组合的期望收益率为 μ_p ,可以求出与此一一对应的最佳投资组合配置 α_p ,然后求得投资组合的标准差,即可绘制有效前沿。但很多时候,我们好难把握期望收益率 μ_p 的范围,导致有效前沿曲线不优雅。下面继续简化最佳投资组合配置 α_p 。

$$\alpha_p = \frac{B(\mu_p C - B)}{AC - B^2} \frac{\Omega^{-1}R}{B} + \frac{C(A - B\mu_p)}{AC - B^2} \frac{\Omega^{-1}I}{C} = x \frac{\Omega^{-1}R}{B} + (1 - x) \frac{\Omega^{-1}I}{C}$$

通过改变 x 的值,就可以求到最佳投资组合配置 α_p ,继而求出投资组合的期望回报以及标准差。

在最佳投资组合配置 α_p 下,

$$\Omega \alpha_p = \frac{\mu_p C - B}{AC - B^2} R + \frac{A - B\mu_p}{AC - B^2} I$$

投资组合方差:

$$var = \alpha_{p}'\Omega\alpha_{p} = \alpha'\left(\frac{\mu_{p}C - B}{AC - B^{2}}R + \frac{A - B\mu_{p}}{AC - B^{2}}I\right) = \frac{\mu_{p}C - B}{AC - B^{2}}\alpha'R + \frac{A - B\mu_{p}}{AC - B^{2}}\alpha'I$$

$$= \frac{\mu_{p}C - B}{AC - B^{2}}\mu_{p} + \frac{A - B\mu_{p}}{AC - B^{2}} \cdot 1 = \frac{C\mu_{p}^{2} - B\mu_{p}}{AC - B^{2}} + \frac{A - B\mu_{p}}{AC - B^{2}}$$

$$= \frac{C\mu_{p}^{2} - 2B\mu_{p} + A}{AC - B^{2}} = \frac{C\left(\mu_{p} - \frac{B}{C}\right)^{2} - \frac{B^{2}}{C} + A}{AC - B^{2}}$$

$$= \frac{C\left(\mu_{p} - \frac{B}{C}\right)^{2} + \frac{AC - B^{2}}{C}}{AC - B^{2}} = \frac{C\left(\mu_{p} - \frac{B}{C}\right)^{2} + \frac{1}{C}}{AC - B^{2}}$$

上式中,当 $\mu_p = \frac{B}{c}$ 时,投资组合方差最小,最小方差为 $\frac{1}{c}$ 。

$$\alpha_{minvar} = \frac{\mu_{p}C - B}{AC - B^{2}} \Omega^{-1}R + \frac{A - B\mu_{p}}{AC - B^{2}} \Omega^{-1}I = \frac{\frac{B}{C}C - B}{AC - B^{2}} \Omega^{-1}R + \frac{A - B\frac{B}{C}}{AC - B^{2}} \Omega^{-1}I$$

$$= 0 + \frac{\frac{1}{C}(AC - B^{2})}{AC - B^{2}} \Omega^{-1}I = \frac{\Omega^{-1}I}{C} = \frac{\Omega^{-1}I}{I'\Omega^{-1}I}$$

资本市场线(Capital Market Line):给定预期回报率,使含无风险资产组合方差最小(夏普比率最大)

Model Construction:

假定有n个风险资产以及 1 个无风险资产(无风险收益率为 R_f)组成的投资组合,给定投资组合的期望收益率为 μ_n 。

$$\min_{\alpha} \alpha' \Omega \alpha$$

s.t.

$$\alpha' R + (1 - \alpha' I) R_f = \mu_p$$

Model Solving:

$$\mathcal{L} = \frac{1}{2}\alpha'\Omega\alpha + \delta\left(\mu_p - \alpha'R - (1-\alpha'\mathsf{I})R_f\right)$$

$$\frac{\partial \mathcal{L}}{\partial \alpha} = \Omega \alpha - \delta (R - R_f I) = 0$$
$$\Rightarrow \alpha = \delta \Omega^{-1} (R - R_f I)$$

$$\begin{split} \frac{\partial \mathcal{L}}{\partial \delta} &= \mu_p - \alpha' R - (1 - \alpha' I) R_f = 0 \\ &\Rightarrow \alpha' R + (1 - \alpha' I) R_f = \mu_p \\ &\Rightarrow R' \alpha + (1 - I' \alpha) R_f = \mu_p \\ &\Rightarrow R' \delta \Omega^{-1} (R - R_f I) + [1 - I' \delta \Omega^{-1} (R - R_f I)] R_f = \mu_p \\ &\Rightarrow \delta R' \Omega^{-1} (R - R_f I) + R_f - \delta R_f I' \Omega^{-1} (R - R_f I) = \mu_p \\ &\Rightarrow \delta R' \Omega^{-1} (R - R_f I) - \delta R_f I' \Omega^{-1} (R - R_f I) = \mu_p - R_f \\ &\Rightarrow \delta [(R' - R_f I') \Omega^{-1} (R - R_f I)] = \mu_p - R_f \\ &\Rightarrow \delta [(R - R_f I)' \Omega^{-1} (R - R_f I)] = \mu_p - R_f \\ &\Rightarrow \delta = \frac{\mu_p - R_f}{[(R - R_f I)' \Omega^{-1} (R - R_f I)]} \end{split}$$

$$\alpha_p = \alpha = \delta \Omega^{-1} (R - R_f I) = \frac{\mu_p - R_f}{[(R - R_f I)' \Omega^{-1} (R - R_f I)]} \Omega^{-1} (R - R_f I)$$