

Math 5760/6890: Introduction to Mathematical Finance

The 2-security Markowitz Efficient Frontier

See Petters and Dong 2016, Section 3.2

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A quick recap

For Markowitz 2-security portfolio optimization:

- Return rates \mathbf{R} for the two securities are random variables
- Assume first- and second-order statistics of these are available: $\boldsymbol{\mu} = \mathbb{E}\mathbf{R}$ and $\text{Cov}(\mathbf{R})$
- The portfolio is defined by the weights \mathbf{w}
- The expected return rate of the portfolio is $\mu_P = \langle \boldsymbol{\mu}, \mathbf{w} \rangle$
- The squared risk of the portfolio is $\sigma_P^2 = \text{Var} \langle \mathbf{R}, \mathbf{w} \rangle = \mathbf{w}^T \text{Cov}(\mathbf{R}) \mathbf{w}$

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We typically seek a risk-optimal portfolio for a given target mean μ_P :

$$\min_{\mathbf{w}} \mathbf{w}^T \mathbf{A} \mathbf{w} \text{ subject to } \langle \mathbf{w}, \mathbf{1} \rangle = 1, \text{ and } \langle \mathbf{w}, \boldsymbol{\mu} \rangle = \mu_P. \quad , \quad \mathbf{A} = \text{Cov} \mathbf{R}$$

- We can formulate an optimization that minimizes risk at a given expected return rate.
- There is typically only one feasible portfolio, hence it's optimal.
- The resulting optimized risk can be *much* lower than the individual security risks.
- The optimal-risk portfolio is *not* necessarily efficient: there can be portfolios having the same risk but higher expected return!

- We can formulate an optimization that minimizes risk at a given expected return rate.
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Today: efficient portfolios and the efficient “frontier”.

Example

Consider the problem of constructing a (Markowitz) 2-security portfolio. The individual securities have expected return rates and covariance given by,

$$\mathbb{E}\mathbf{R}(T) = \begin{pmatrix} 2 \\ 6 \end{pmatrix}, \quad \text{Cov}\mathbf{R}(T) = \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix}.$$

We've computed the risk-optimal portfolio as a function of μ_P :

$$\mathbf{w} = \begin{pmatrix} \frac{6-\mu_P}{4} \\ \frac{\mu_P-2}{4} \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 3 \\ -1 \end{pmatrix} + \frac{\mu_P}{4} \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$\frac{\mu_P-2}{4}$

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And we've used this to determine the optimal risk σ_P for a fixed μ_P :

$$\begin{aligned} \sigma_P^2 &= \frac{3}{8}\mu_P^2 - 3\mu_P + \frac{13}{2} \\ &= \frac{3}{8}(\mu_P - 4)^2 + \frac{1}{2} \end{aligned}$$

$$\sigma_P^2 = \frac{3}{8}\mu_P^2 - 3\mu_P + \frac{13}{2} = \frac{3}{8}(\mu_P - 4)^2 + \frac{1}{2}$$

This relationship can be used to identify the set of all possible risk-optimal portfolios as a graph of valid (σ_P, μ_P) pairs.

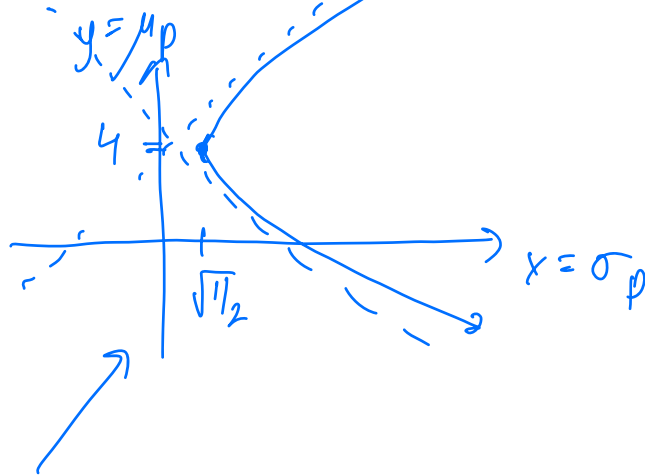
$$\rightarrow \sigma_P^2 - \frac{3}{8}(\mu_P - 4)^2 = \frac{1}{2}$$

$$\frac{\sigma_P^2}{\frac{1}{2}} - \frac{(\mu_P - 4)^2}{\frac{8}{3} \cdot \frac{1}{2}} = 1 \rightarrow \frac{\sigma_P^2}{\frac{1}{2}} - \frac{(\mu_P - 4)^2}{\frac{4}{3}} = 1$$

$$(x, y) = (\sigma_P, \mu_P)$$

$$\frac{x^2}{\frac{1}{2}} - \frac{(y - 4)^2}{\frac{4}{3}} = 1$$

→ relationship defining a graph of a hyperbola (!!)



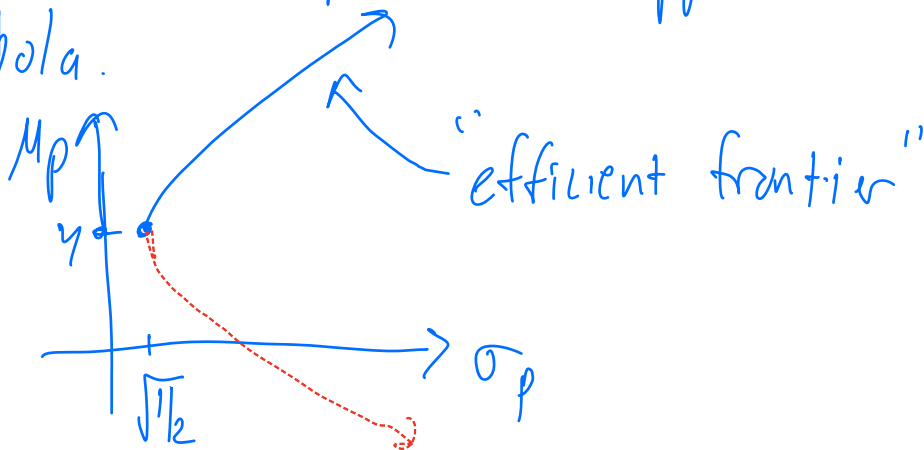
$$\frac{y-4}{4/3} = \pm \frac{x}{\sqrt{1/2}} \quad (\text{slant asymptotes})$$

Graph of all risk-optimal portfolios.
Which ones are efficient?

Non-
V

Efficient: if decrease σ_p , I can increase μ_p
if I increase μ_p , I can decrease σ_p .

Set of efficient portfolios: upper half of hyperbola.



The particular aspects we've explored in the previous example are generic:

- The set of risk-optimal portfolios is defined by the graph of a hyperbola in the (σ_P, μ_P) plane.
- The upper half of this graph are efficient portfolios – the *efficient frontier*.

$$\sigma_p^2 = \frac{2}{8} (\mu_p - 4)^2 + \frac{1}{2}$$

$$\mu_p = 4, \sigma_p = \sqrt{\frac{1}{2}} \rightarrow \text{global risk-min. portfolio.}$$

The particular aspects we've explored in the previous example are generic:

- The set of risk-optimal portfolios is defined by the graph of a hyperbola in the (σ_P, μ_P) plane.
- The upper half of this graph are efficient portfolios – the *efficient frontier*.
- There is a *global* risk-optimal portfolio, which corresponds to a particular expected return rate.
- In the (general) 2-security model, any feasible portfolio is risk-optimal.¹

¹This is not true if $\mu_1 = \mu_2$.

The general 2-security case

The previous aspects hold in the general 2-security case, assuming $\mu_1 = \mu_2$ and a positive-definite covariance.

Example

Consider the general 2-security setup:

$$\boldsymbol{\mu} = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}, \quad \mathbf{A} = \text{Cov}(\mathbf{R}),$$

where \mathbf{A} is positive-definite and $\mu_1 \neq \mu_2$. Show that risk-optimal portfolios are given by the locus of points (σ_P, μ_P) satisfying,

$$\frac{\sigma_P^2}{a^2} - \frac{(\mu_P - \mu_G)^2}{b^2} = 1.$$

where a, b, μ_G are explicit constants: μ_G is the expected return rate of the global risk-optimal portfolio.

Portfolio constraints: $w_1 + w_2 = 1 \rightarrow \mu_1 w_1 + \mu_1 w_2 = \mu_1 \quad (1)$

$\mu_1 w_1 + \mu_2 w_2 = \mu_P \quad (2)$

$$(2) - (1): w_2 (\mu_2 - \mu_1) = \mu_p - \mu_1$$

$$w_2 = \frac{\mu_p - \mu_1}{\mu_2 - \mu_1}$$

$$w_1 = 1 - w_2 = \frac{\mu_2 - \mu_p}{\mu_2 - \mu_1}$$

$$\underline{w} = \begin{pmatrix} w_1 \\ w_2 \end{pmatrix} = \frac{1}{\mu_2 - \mu_1} \begin{pmatrix} \mu_2 - \mu_p \\ \mu_p - \mu_1 \end{pmatrix}$$

$$= \underbrace{\frac{1}{\mu_2 - \mu_1} \begin{pmatrix} \mu_2 \\ -\mu_1 \end{pmatrix}}_{\underline{v}_0} + \underbrace{\frac{\mu_p}{\mu_2 - \mu_1} \begin{pmatrix} -1 \\ 1 \end{pmatrix}}_{\mu_p \underline{v}_1}$$

$$\sigma_p^2 = \underline{w}^T \underline{A} \underline{w} = (\underline{v}_0 + \mu_p \underline{v}_1)^T \underline{A} (\underline{v}_0 + \mu_p \underline{v}_1)$$

$$= \underline{v}_0^T \underline{A} \underline{v}_0 + 2 \mu_p \underline{v}_1^T \underline{A} \underline{v}_0 + \mu_p^2 \underline{v}_1^T \underline{A} \underline{v}_1$$

$$= \underline{v}_1^T \underline{A} \underline{v}_1 \left[\mu_p^2 + 2 \mu_p \frac{\underline{v}_1^T \underline{A} \underline{v}_0}{\underline{v}_1^T \underline{A} \underline{v}_1} \right] + \underline{v}_0^T \underline{A} \underline{v}_0$$

$$= \underline{v}_1^T \underline{A} \underline{v}_1 \left[\mu_p + \frac{\underline{v}_1^T \underline{A} \underline{v}_0}{\underline{v}_1^T \underline{A} \underline{v}_1} \right]^2 - \frac{(\underline{v}_1^T \underline{A} \underline{v}_0)^2}{\underline{v}_1^T \underline{A} \underline{v}_1} + \underline{v}_0^T \underline{A} \underline{v}_0$$

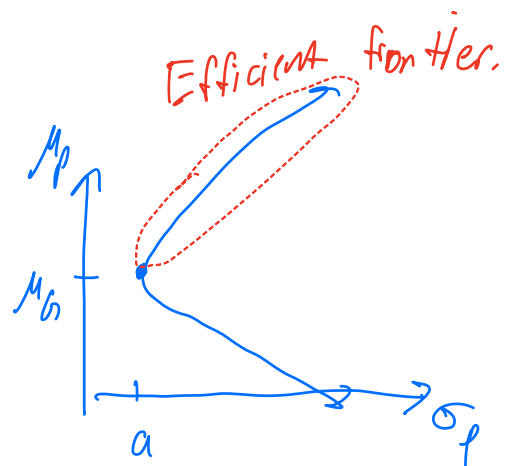
$$\sigma_p^2 = \frac{\left[\mu_p + \frac{v_1^T A v_0}{v_1^T A v_1} \right]^2}{\left(\frac{1}{v_1^T A v_1} \right)} = v_0^T A v_0 - \frac{(v_1^T A v_0)^2}{v_1^T A v_1}$$

$$a^2 = v_0^T A v_0 - \frac{(v_1^T A v_0)^2}{v_1^T A v_1}$$

$$b^2 = \frac{v_0^T A v_0}{v_1^T A v_1} - \frac{(v_1^T A v_0)^2}{(v_1^T A v_1)^2}$$

$$\Rightarrow \frac{\sigma_p^2}{a^2} - \frac{(\mu_p - \mu_G)^2}{b^2} = 1$$

$$\mu_G = \frac{-v_1^T A v_0}{v_1^T A v_1}$$



What about short selling?

L07-S08

For general $\mathbb{E}\mathbf{R}$, $\text{Cov}(\mathbf{R})$, μ_P , the risk-optimal portfolio could be highly leveraged (even if it's efficient).

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In practice, we may want to disallow short selling to avoid such situations.

The corresponding optimization problem is an augmentation of what we've already seen:

$$\begin{aligned} \min_{\mathbf{w}} \mathbf{w}^T \mathbf{A} \mathbf{w} \quad \text{subject to} \quad & \langle \mathbf{w}, \mathbf{1} \rangle = 1, \text{ and} \\ & \langle \mathbf{w}, \boldsymbol{\mu} \rangle = \mu_P, \text{ and} \\ & w_i \geq 0, \quad i = 1, \dots, N. \end{aligned}$$

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In general, this is a harder optimization problem to solve.

But in the 2-security case, the solution is fairly transparent.

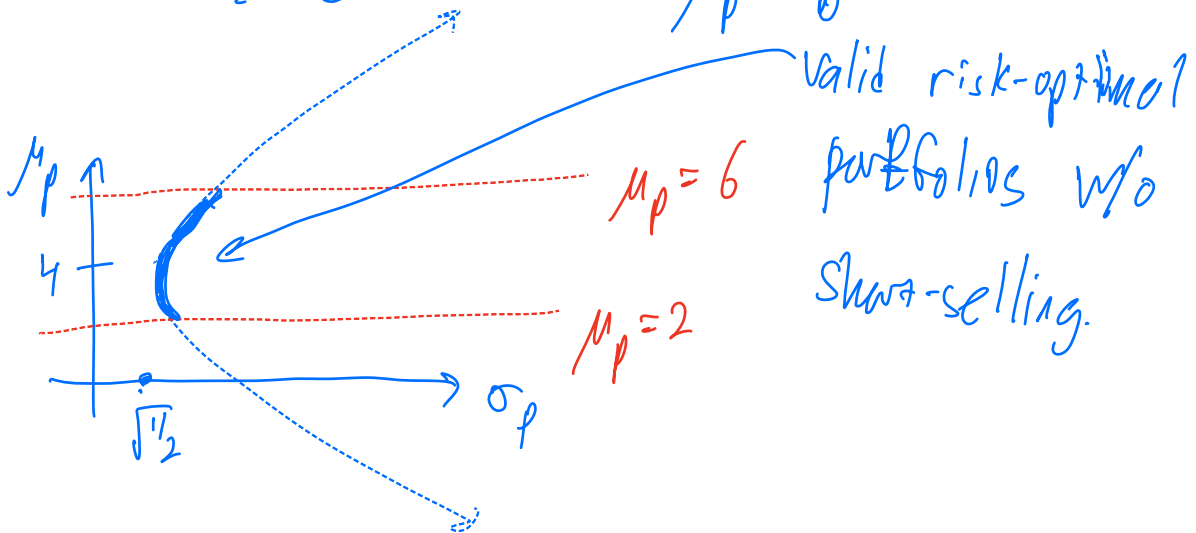
First
Previous example

$$\underline{\mu} = \begin{pmatrix} 2 \\ 6 \end{pmatrix} \quad \underline{A} = \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix}$$

$$\underline{w} = \begin{pmatrix} \frac{6 - \mu_p}{4} \\ \frac{\mu_p - 2}{4} \end{pmatrix}$$

$$\sigma_p^2 = \frac{3}{8} (\mu_p - 4)^2 + \frac{1}{2}$$

$$w_1, w_2 \geq 0 \implies 2 \leq \mu_p \leq 6$$



In the 2-security case:

- Generally, no optimization is needed for 2-security portfolios: the weight constraints uniquely identify portfolios
- Many risk-optimal portfolios are *not* efficient
- The set of risk-optimal portfolios can be explicitly parameterized + plotted
- There is a global variance-minimizing portfolio – investors might not want this portfolio.

Many of the lessons we've learned from 2-security portfolios will hold in the N -security case:


- Risk-optimal portfolios can be computed and plotted
- The efficient frontier is visually identifiable
- There is a globally risk-optimal portfolio

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There are some differences that make things more complicated:

- Optimization is generally required
- There are (many) feasible portfolios that are not risk-optimal
- Explicit pen+paper computations become harder

 Petters, Arlie O. and Xiaoying Dong (2016). *An Introduction to Mathematical Finance with Applications: Understanding and Building Financial Intuition*. Springer. ISBN: 978-1-4939-3783-7.