

Math 5760/6890: Introduction to Mathematical Finance

2-Security Markowitz Efficient Frontier

See Petters and Dong 2016, Section 3.2

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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 :

$$\underline{A} = \text{Cov}(\underline{R})$$

$$\min_{\mathbf{w}} \mathbf{w}^T \mathbf{A} \mathbf{w} \text{ subject to } \langle \mathbf{w}, \mathbf{1} \rangle = 1, \text{ and}$$

$$\langle \mathbf{w}, \boldsymbol{\mu} \rangle = \mu_P.$$

- 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!

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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 :

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

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

$$\sigma_P^2 = \frac{3}{8}\mu_P^2 - 3\mu_P + \frac{13}{2}$$

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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.

$$\sigma_P^2 - \frac{3}{8}\mu_P^2 + 3\mu_P = \frac{13}{2}$$

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

$$\sigma_P^2 - \frac{3}{8}(\mu_P^2 - 8\mu_P + 16) = \frac{13}{2} - 6$$

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

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

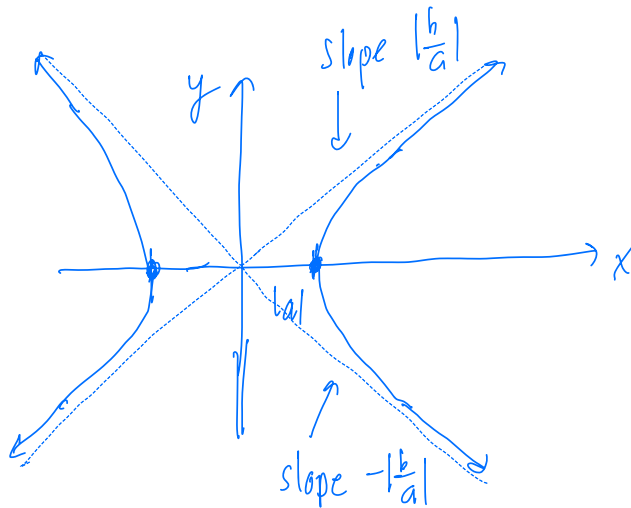
$$\frac{\sigma_P^2}{1/2} - \frac{(\mu_P - 4)^2}{4/3} = 1$$

$$\text{" } \frac{x^2}{a^2} - \frac{y^2}{b^2} = 1 \text{ "}$$

$$y=0 \rightarrow x = \pm \sqrt{a^2} = \pm a$$

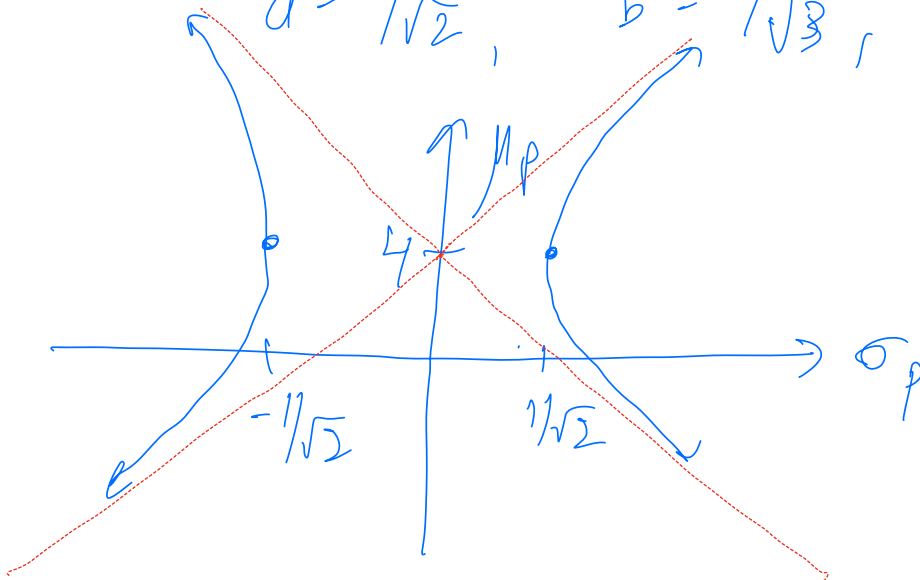
$$x, y \text{ very large} : \frac{x^2}{a^2} - \frac{y^2}{b^2} = 1 \approx \frac{x^2}{a^2} = \frac{y^2}{b^2}$$

$$|y| = \frac{b}{a} |x|$$

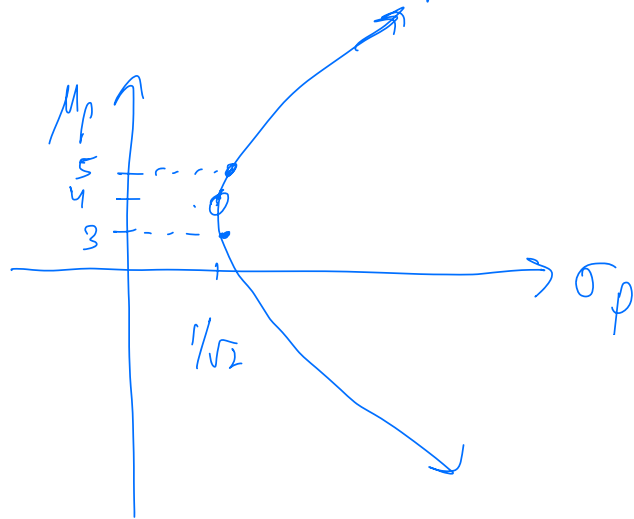


$$\text{Back to } \frac{\sigma_p^2}{1/2} - \frac{(\mu_p - 4)^2}{4/3} = 1$$

$$a = 1/\sqrt{2}, \quad b = 2/\sqrt{3}, \quad \frac{b}{a} = \frac{2}{\sqrt{3}} \sqrt{2}$$



No negative variances: $\sigma_p < 0$ irrelevant.



o: "global" variance-minimizing portfolio
 $\mu_p = 4$

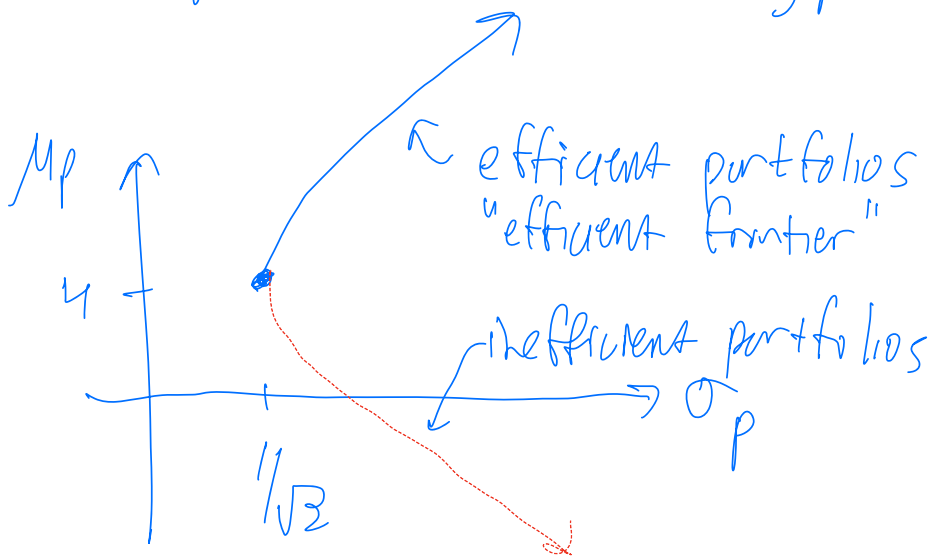
$$\underbrace{\underline{w} @ \mu_p = 4}_{\underline{w} = \begin{pmatrix} 1/2 \\ 1/2 \end{pmatrix}} \quad \sigma_p = 1/\sqrt{2}$$

Which portfolios are efficient?

Efficient portfolio: μ_p increases $\Rightarrow \sigma_p$ increases
 σ_p decreases $\Rightarrow \mu_p$ decrease

"In math": $\frac{d\sigma_p}{d\mu_p} > 0$ $\frac{d\mu_p}{d\sigma_p} > 0$

⇒ Efficient portfolios are the "upper half" of the 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*.

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 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.

weight vectors are unique¹ $w_1 + w_2 = 1$
 $\mu_1 w_1 + \mu_2 w_2 = \mu_P$

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

$$\underline{v}_0 = \frac{1}{\mu_2 - \mu_1} \begin{pmatrix} \mu_2 \\ -\mu_1 \end{pmatrix}$$

$$\underline{v}_1 = \frac{1}{\mu_2 - \mu_1} \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$$\sigma_p^2 = \underline{w}^T \underline{A} \underline{w}$$

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

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

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

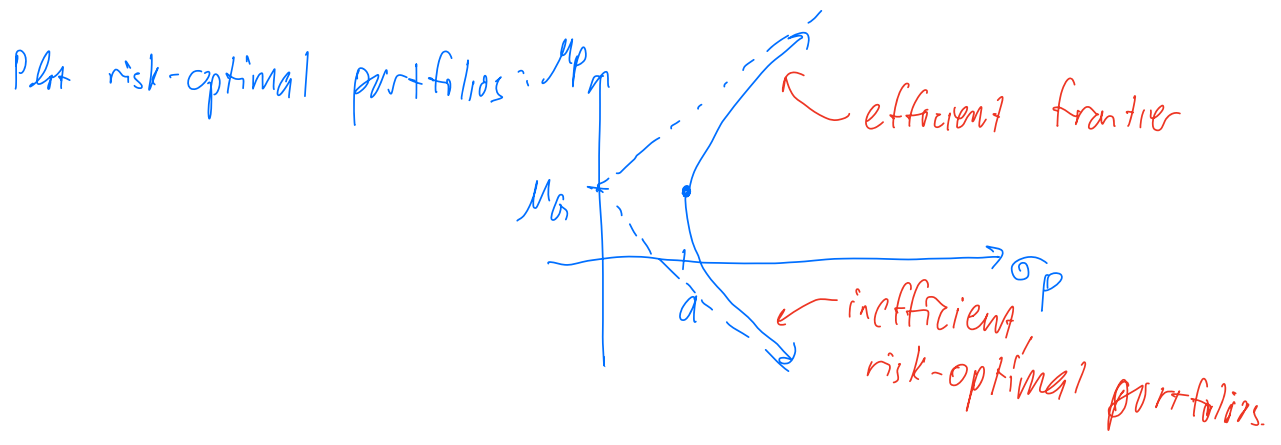
$$= \underline{v}_1^T \underline{A} \underline{v}_1 \left(\mu_p^2 + \frac{2 \underline{v}_1^T \underline{A} \underline{v}_0}{\underline{v}_1^T \underline{A} \underline{v}_1} \mu_p + \left(\frac{\underline{v}_1^T \underline{A} \underline{v}_0}{\underline{v}_1^T \underline{A} \underline{v}_1} \right)^2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

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

$$\Rightarrow \frac{\sigma_p^2}{a^2} - \frac{(\mu_p - \mu_0)^2}{b^2} = 1, \quad \mu_0 = \frac{-\underline{v}_1^T \underline{A} \underline{v}_0}{\underline{v}_1^T \underline{A} \underline{v}_1}$$

$$a^2 = \underline{v}_0^T \underline{A} \underline{v}_0 - \frac{(\underline{v}_1^T \underline{A} \underline{v}_0)^2}{\underline{v}_1^T \underline{A} \underline{v}_1} (> 0)$$

$$b^2 = \frac{a^2}{\underline{v}_1^T \underline{A} \underline{v}_1}$$



What about short selling?

D09-S08(a)

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

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

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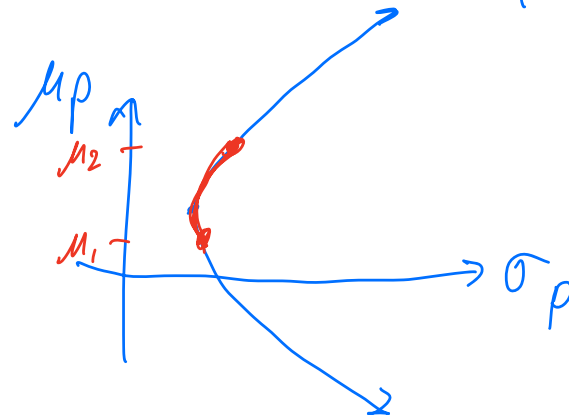
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(2-security case)

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

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



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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- Risk-optimal portfolios can be computed and plotted
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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.