

Math 5760/6890: Introduction to Mathematical Finance Mutual Fund Theorem

See Petters and Dong 2016, Section 3.5

Thursday's class (Sept. 28 online on Zoom)
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HW #5: Due Thursday (Oct 5)



Given N securities, Markowitz Portfolio Analysis is a one-period model that prescribes efficient portfolios

- as those for which one cannot attain higher reward without higher risk
- as those for which one cannot reduce risk without also reducing reward

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The definition of reward and risk are first- and second-order statistics (mean-variance analysis) of the portfolio return rate. The risk-optimal (not necessarily efficient) formulation is

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

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

where $\text{Cov}(\mathbf{R})$

- $\mathbf{A} = \text{Cov}(\mathbf{R})$ is assumed positive-definite
- $\boldsymbol{\mu} = \mathbb{E}\mathbf{R}$ is assumed not parallel to $\mathbf{1}$.
- μ_P is a target expected return rate

$$\boldsymbol{\mu} = (\mu_1, \mu_2, \dots, \mu_N)^T$$

Practical concerns

Our setup assumes statistics such as $\mathbb{E}\mathbf{R}$ and $\text{Cov}(\mathbf{R})$ are available.

These statistics are typically computed using historical data. E.g.,

- <https://www.portfoliovisualizer.com>
- <https://www.buyupside.com/index.html>
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Typically, cross-covariance entries are reported as *correlations*. The main reason for this is that correlations aren't sensitive to units of time (scaling):

$$\text{Cov}(aX, bY) = ab\text{Cov}(X, Y)$$

$$\sqrt{\text{Var}(aX)} = |a|\sqrt{\text{Var}X}$$

\Downarrow

$$\rho(aX, bY) := \frac{\text{Cov}(aX, bY)}{\sqrt{\text{Var}(aX)} \sqrt{\text{Var}(bY)}} = \frac{ab}{|ab|} \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X)} \sqrt{\text{Var}(Y)}} = \text{sign}(ab) \rho(X, Y).$$

\uparrow
 $[-1, 1]$

$[-1, 1]$

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(And if they existed, investors would quickly take advantage of the arbitrage, removing it from the market.)

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However, if it were to happen, then one can show:

- ▶ The expected return rate for any Markowitz portfolio can only be the value μ_1 .
- ▶ Hence, the set of all feasible portfolios collapses to a line on a μ_P vs σ_P plot.
- ▶ The (unique) global variance-minimizing portfolio is the only rational choice.

A simple example of portfolio engineering in practice: mutual funds.

A mutual fund is an investment fund in which assets from various investors are collectively pooled to purchase securities according to a portfolio design.

I.e., a mutual fund is a security in which an investor can effectively purchase shares of a portfolio (not of individual securities).

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Generally speaking, there are actively-managed, and passively-managed funds.

- Passively managed funds purchase stocks according to some preset rule, such as according to a stock market index (“index funds”).
This requires relatively little hands-on work, and so typically have lower fees.
- Actively managed funds have financial professionals who attempt to engineer portfolios to beat preset rules/market index returns.
Since this requires employees/managers to engineer portfolios, these typically come with higher fees.

(Most actively managed funds are built with procedures that are substantially more technical than simple mean-variance analysis.)

As we've seen, anyone seeking to build an N -security portfolio has some tasks ahead of them:

- Collecting + analyzing stock data to compute expected returns and correlations
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Hence, in practice one requires nontrivial investment + infrastructure to construct N -security portfolios.

This is essentially why actively managed funds charge fees – and so it is reasonable to assume that actively managed funds produce efficient portfolios.

The little guy

This whole reality produces a quandry for you and me:

- It's best to leave construction of N -security portfolios to professionals, since most of us don't have the time + resources to do this ourselves.
- The available professionally-managed funds might not have an expected return rate μ_P that we prefer.

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There is a reasonable solution here, but let's clarify the setup. We'll assume:

- We are in an N -security market, and all (actively-managed) mutual funds invest in the same N securities.
- Actively-managed mutual funds produce efficient portfolios. \rightarrow portfolios lie on efficient frontier.

With this setup, then a(ny) mutual fund will produce a portfolio given by,

$$w = v_0 + \mu_P v_1,$$

for some target μ_P , with v_0 and v_1 some computable vectors.

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Using this, we can construct other efficient portfolios with arbitrary values of $\mathbb{E}R_P$.

Assume I have access to portfolios α, β with target expected return rates μ_α, μ_β .

$\underline{w}_\alpha = \underline{v}_0 + \mu_\alpha \underline{v}_1$
 $\underline{w}_\beta = \underline{v}_0 + \mu_\beta \underline{v}_1$

} What if I want a portfolio w/
 return rate $\mu = \mathbb{E}R_p$?

Choose γ , and construct a portfolio as

$$\begin{aligned}
 \underline{w} &= \gamma \underline{w}_\alpha + (1-\gamma) \underline{w}_\beta = \gamma \underline{v}_0 + \gamma \mu_\alpha \underline{v}_1 + (1-\gamma) \underline{v}_0 + \\
 &\quad (1-\gamma) \mu_\beta \underline{v}_1 \\
 &= \underline{v}_0 + \underbrace{(\gamma \mu_\alpha + (1-\gamma) \mu_\beta)}_{\mu} \underline{v}_1
 \end{aligned}$$

Choose γ to make
this equal μ .

$$\Rightarrow \mu = \gamma \mu_\alpha + (1-\gamma) \mu_\beta$$

$$\gamma = \frac{\mu - \mu_\beta}{\mu_\alpha - \mu_\beta} \quad (\mu_\alpha \neq \mu_\beta)$$

The Mutual Fund Theorem

We have proven the following:

Theorem (Mutual Fund Thm)

Consider an N -security market. If w_α and w_β are two distinct efficient portfolios (e.g., mutual funds), then any other efficient portfolio w can be written as,

$$w = \alpha w_\alpha + (1 - \alpha) w_\beta,$$

where w_α and w_β are two other distinct efficient portfolios (e.g., mutual funds).

With $R_P, R_\alpha,$ and R_β the return rates corresponding to portfolios $w, w_\alpha,$ and $w_\beta,$ respectively, then if $\mathbb{E}R_P$ is between $\mathbb{E}R_\alpha$ and $\mathbb{E}R_\beta$, then both coefficients α and $1 - \alpha$ are non-negative.

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Theorem

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The punchline: if we have the ability to purchase from two distinct efficient mutual funds (N -security portfolios), then we can easily construct a 2-security portfolio with a target mean that has the risk characteristics of the whole set of N securities.

I.e.:

(2 efficient N -security portfolios) \implies (All efficient N -security portfolios)

("Efficient" can be replaced by "risk-optimal")

 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.