### Math 5760/6890: Introduction to Mathematical Finance

### The Mutual Fund Theorem

See Petters and Dong 2016, Section 3.5

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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

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

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

where  $C_{\mathcal{N}}(\underline{\mathcal{R}})$ 

- $A = Cov(R_P)$  is assumed positive-definite
- $\mu = \mathbb{E} \boldsymbol{R}$  is assumed <u>not</u> parallel to 1.
- $\mu_P$  is a target expected return rate

# Practical concerns

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

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

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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):

$$\begin{aligned} \operatorname{Cov}(aX, bY) &= ab\operatorname{Cov}(X, Y) \\ \sqrt{\operatorname{Var}(aX)} &= |a|\sqrt{\operatorname{Var}X} \\ & \downarrow \\ \rho(aX, bY) &\coloneqq \frac{\operatorname{Cov}(aX, bY)}{\sqrt{\operatorname{Var}(aX)\operatorname{Var}(bY)}} = \frac{ab}{|ab|} \frac{\operatorname{Cov}(X, Y)}{\sqrt{\operatorname{Var}(X)\operatorname{Var}(Y)}} = \operatorname{sign}(ab)\rho(X, Y). \end{aligned}$$

A couple of small digressions:

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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  $\mu_1$ .
- Hence, the set of all feasible portfolios collapses to a line on a  $\mu_P$  vs  $\sigma_P$  plot.
- The (unique) global variance-minimizing portfolio is the only rational choice.

# Mutual Funds

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.

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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.)

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- Even if you successfully constructed a large-N portfolio, you'd have to implement it: you'd have to purchase/continuously restructure investment in N securities.

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  $\mu_P$  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.

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

$$\boldsymbol{w}=\boldsymbol{v}_0+\mu_P\boldsymbol{v}_1,$$

for some target  $\mu_P$ , with  $oldsymbol{v}_0$  and  $oldsymbol{v}_1$  some computable vectors.

Fund A: 
$$W_A = V_0 + M_A Y_1$$
 ( $M_A$  fixed)  
Fund B:  $W_B = V_0 + M_B Y_1$  ( $M_B$  fixed)  $M_A \neq M_B$   
then for  $A \in \mathbb{R}$ :  $A W_A + (I - \lambda) W_B = V_0 + [\lambda M_A + (I - \lambda) A_B] V_1$ 

=> Can choose & s.t. A pra+(1-X) AB is any R

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

# The Mutual Fund Theorem

We have proven the following:

#### Theorem

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

$$\boldsymbol{w} = \alpha \boldsymbol{w}_{\alpha}^{\prime} + (1 - \alpha) \boldsymbol{w}_{\beta}^{\prime},$$

where  $w_{\alpha}$  and  $w_{\beta}$  are two other distinct efficient portfolios (e.g., mutual funds).

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

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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.

l.e.:

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

D11-S08(b)



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.