

MUNI
ECON

SUPTECH WORKSHOP III

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Background Session I – Introduction to modern portfolio theory

A primer on microeconomics

Utility and choice

Preference relation

$$a \succ b \quad a \sim b \quad a \prec b$$

Rationality assumptions:

- Every investor possesses a **complete** preference relation.
- The preference relation satisfies the property of **transitivity**.
- The preference relation is **continuous**.

Utility

Previous are sufficient to guarantee the existence of a continuous function $u : \mathbb{R}^N \rightarrow \mathbb{R}$ such that, for any consumption bundles a and b ,

$$a \succeq b \Leftrightarrow u(a) \geq u(b)$$

This real-valued function u is called a **utility function**.

Risk aversion

Consider an investor with wealth Y and a fair-game lottery $L = (h, -h, 0.5)$ with $h > 0$.

We say an investor is **risk averse** iff

$$Y \succ Y + L$$

This implies the utility function to be strictly concave:

$$E[U(Y)] > E[U(Y + L)]$$

$$U(Y) > \frac{1}{2}U(Y + h) + \frac{1}{2}U(Y - h)$$

Thus, $U''(Y) < 0$ and we have decreasing marginal utility of wealth.

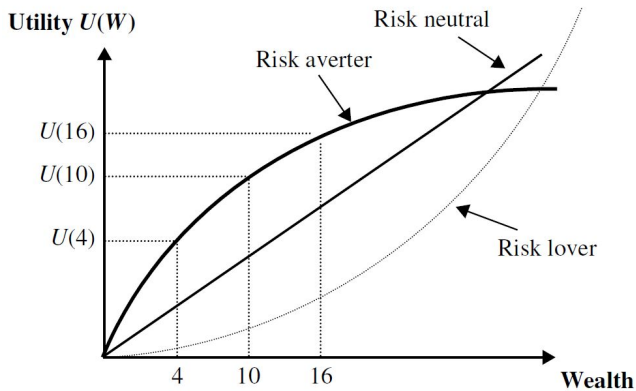


Figure 2 Utility functions

Linear algebra basics

Linear algebra basics

A vector $\mathbf{x} \in \mathbb{R}^n$ and a matrix $\mathbf{A} \in \mathbb{R}^n \times \mathbb{R}^m$ for $n, m \in \mathbb{N}$.

$$\mathbf{1}_n = (1, 1, \dots, 1) \quad \mathbf{I}_n = \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & \cdots & 1 \end{pmatrix}$$

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 0 \\ 4 & 5 & 6 \end{pmatrix}, \quad \mathbf{A}^T = \begin{pmatrix} 1 & 0 & 4 \\ 2 & 1 & 5 \\ 3 & 0 & 6 \end{pmatrix}$$

$$i \left(\text{---} \right) \left(\begin{array}{c} | \\ | \\ | \end{array} \right) = \left(\text{---} \begin{array}{c} | \\ | \\ | \end{array} \right)$$

$$\begin{pmatrix} 2 & 1 & 0 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 1 & 0 \\ 2 & 3 \end{pmatrix} = ?$$

Some examples

$$\mathbf{r} = (r_1, r_2, \dots, r_n)$$

$$E(\mathbf{r}) = (E(r_1), E(r_2), \dots, E(r_n))$$

$$\mathbf{w}^T \mathbf{r} = (w_1, w_2, \dots, w_n) \begin{pmatrix} r_1 \\ r_2 \\ \vdots \\ r_n \end{pmatrix} = w_1 r_1 + w_2 r_2 + \dots + w_n r_n = \sum_{i=1}^n w_i r_i$$

$$\mathbf{1}_n^T \mathbf{w} = (1, 1, \dots, 1) \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix} = \sum_{i=1}^n w_i$$

More examples

Let $\mathbf{w} \in \mathbb{R}^n$ and $\Sigma \in \mathbb{R}^n \times \mathbb{R}^n$.

$$\mathbf{w}^T \Sigma \mathbf{w} = (w_1, w_2, \dots, w_n) \begin{pmatrix} \text{cov}(r_1, r_1) & \text{cov}(r_1, r_2) & \cdots & \text{cov}(r_1, r_n) \\ \text{cov}(r_2, r_1) & \text{cov}(r_2, r_2) & \cdots & \text{cov}(r_2, r_n) \\ \vdots & \vdots & \ddots & \vdots \\ \text{cov}(r_n, r_1) & \text{cov}(r_n, r_2) & \cdots & \text{cov}(r_n, r_n) \end{pmatrix} \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix}$$

$$\mathbf{w}^T \Sigma \mathbf{w} = \sum_{i=1}^n \sum_{j=1}^n w_i w_j \text{cov}(r_i, r_j)$$

Expected value

$$E(X) = \sum p_i x_i$$

Properties:

- **E1** $E(a) = a$
- **E2** $E(aX) = aE(X)$
- **E3** $E(X + Y) = E(X) + E(Y)$

Covariance

$$\text{cov}(X, Y) = E[(X - E(X))(Y - E(Y))]$$

Properties:

- **C1** $\text{cov}(a, X) = 0$
- **C2** $\text{cov}(X, Y) = \text{cov}(Y, X)$
- **C3** $\text{cov}(a + bX, Y) = b\text{cov}(X, Y)$
- **C4** $\text{cov}(X + Y, Z) = \text{cov}(X, Z) + \text{cov}(Y, Z)$

Variance

$$\text{var}(X) = E[(X - E(X))^2] = \text{cov}(X, X)$$

Properties:

- **D1** $\text{var}(a) = 0$
- **D2** $\text{var}(a + bX) = b^2\text{var}(X)$
- **D3** $\text{var}(X + Y) = \text{var}(X) + \text{var}(Y) + 2\text{cov}(X, Y)$

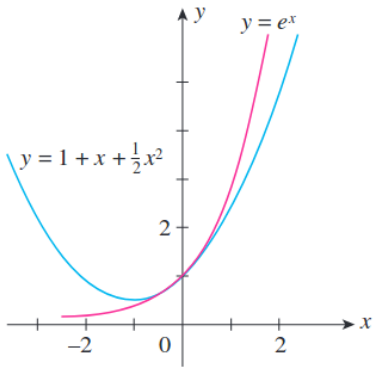
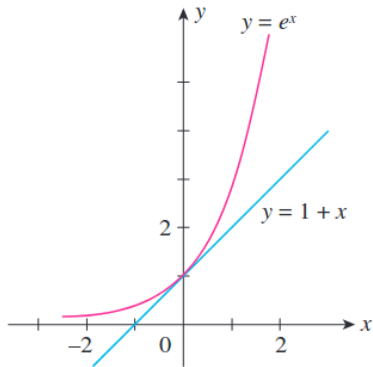
Correlation

$$\rho(X, Y) = \frac{\text{cov}(X, Y)}{\sqrt{\text{var}(X)\text{var}(Y)}}$$

Mean-variance portfolio theory

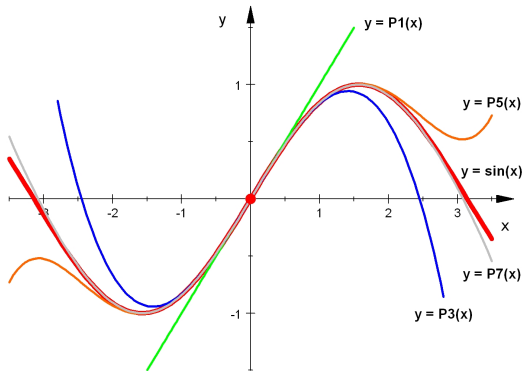
Mean-variance preferences

The general n -variate problem is difficult, often simplified to M-V. Taylor expansion of $U(\tilde{Y}_1)$ around $E(\tilde{Y}_1)$:



Mean-variance preferences

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Mean-variance preferences

The general n -variate problem is difficult, often simplified to M-V. Taylor expansion of $U(\tilde{Y}_1)$ around $E(\tilde{Y}_1)$:

$$\begin{aligned}
 U(\tilde{Y}_1) &= U(E[\tilde{Y}_1]) + \\
 &\quad U'(E[\tilde{Y}_1]) \cdot (\tilde{Y}_1 - E[\tilde{Y}_1]) + \\
 &\quad \frac{1}{2}U''(E[\tilde{Y}_1]) \cdot (\tilde{Y}_1 - E[\tilde{Y}_1])^2 + \varepsilon
 \end{aligned}$$

thus for the **expected utility**

$$E[U(\tilde{Y}_1)] = U(E[\tilde{Y}_1]) + \frac{1}{2}U''(E[\tilde{Y}_1]) \cdot Var[\tilde{Y}_1] + E[\varepsilon]$$

The case with two assets

$$r_1, r_2 \quad \sigma_1^2, \sigma_2^2 \quad \rho_{1,2} = \text{COV}(r_1, r_2) \quad w_1, w_2$$

$$r_p = w_1 r_1 + w_2 r_2$$

$$E(r_p) = ??$$

$$\sigma_p = ??$$

The case with two assets

$$r_1, r_2 \quad \sigma_1^2, \sigma_2^2 \quad \rho_{1,2} = \text{COV}(r_1, r_2) \quad w_1, w_2$$

$$r_p = w_1 r_1 + w_2 r_2$$

$$E(r_p) = w_1 E(r_1) + w_2 E(r_2)$$

$$\sigma_p^2 = \text{var}(r_p) = w_1^2 \sigma_1^2 + w_2^2 \sigma_2^2 + 2w_1 w_2 \text{COV}(r_1, r_2)$$

Portfolio – initial setup, n assets

Let $\mathbf{w} \in \mathbb{R}^n$ and $\Sigma \in \mathbb{R}^n \times \mathbb{R}^n$.

$$\mathbf{w}^T \Sigma \mathbf{w} = (w_1, w_2, \dots, w_n) \begin{pmatrix} \text{cov}(r_1, r_1) & \text{cov}(r_1, r_2) & \cdots & \text{cov}(r_1, r_n) \\ \text{cov}(r_2, r_1) & \text{cov}(r_2, r_2) & \cdots & \text{cov}(r_2, r_n) \\ \vdots & \vdots & \ddots & \vdots \\ \text{cov}(r_n, r_1) & \text{cov}(r_n, r_2) & \cdots & \text{cov}(r_n, r_n) \end{pmatrix} \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix}$$

$$\mathbf{w}^T \Sigma \mathbf{w} = \sum_{i=1}^n \sum_{j=1}^n w_i w_j \text{cov}(r_i, r_j)$$

The diversification/insurance principle

$$\begin{aligned}\sigma_p^2 &= \sum_{i=1}^n \sum_{j=1}^n w_i w_j \text{COV}(r_i, r_j) \\ &= \sum_{i=1}^n w_i^2 \text{var}(r_i) + 2 \sum_{i=1}^n \sum_{j>i} w_i w_j \text{COV}(r_i, r_j)\end{aligned}$$

For mutually uncorrelated r_i with equal variance $\sigma_i = \sigma$, and an equal weights strategy $w_i = 1/n$ we have

$$\sigma_p^2 = \sum_{i=1}^n \frac{1}{n^2} \sigma^2 = \frac{n}{n^2} \sigma^2 = \frac{\sigma^2}{n}$$

And for the limiting case $\lim_{n \rightarrow \infty} \sigma_p^2 = \lim_{n \rightarrow \infty} \sigma^2/n = 0$

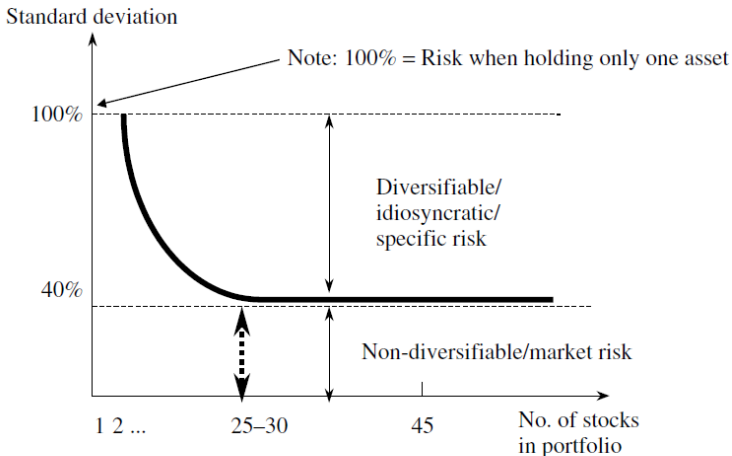


Figure 1 Random selection of stocks

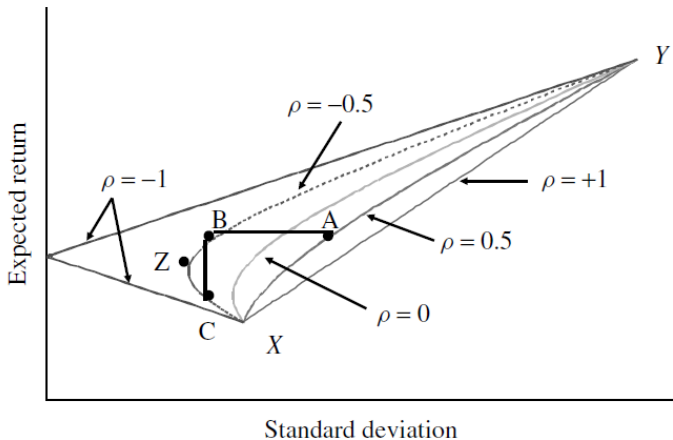


Figure 3 Efficient frontier and correlation

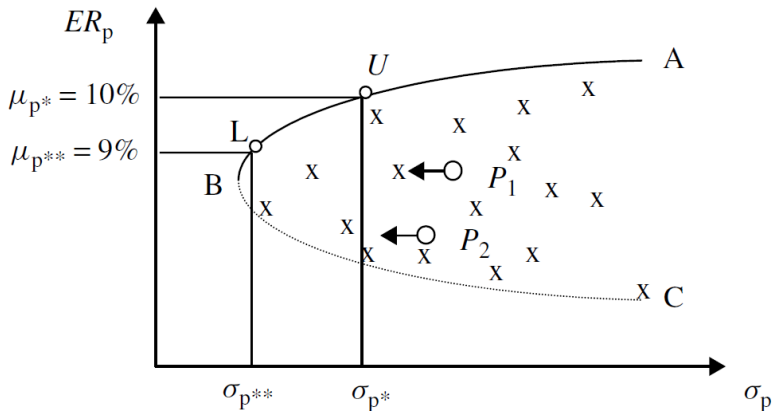


Figure 4 Efficient and inefficient portfolios

Risk-free asset: transformation line

Assume a risk-free asset with $E(r) = r$ and $\sigma_r = 0$.

Now we mix a risky asset $(r_m, E(r_m), \sigma_m)$ with the risk free asset. For the portfolio with w_1 invested into the risky asset, we get

$$E(r_p) = w_m E(r_m) + (1 - w_m)r$$

$$\sigma_p = w_m \sigma_m$$

$$E(r_p) = \frac{\sigma_p}{\sigma_m} E(r_m) + \left(1 - \frac{\sigma_p}{\sigma_m}\right) r$$

$$= r + \frac{E(r_m) - r}{\sigma_m} \sigma_p$$

$$= \delta_0 + \delta_m \sigma_p$$

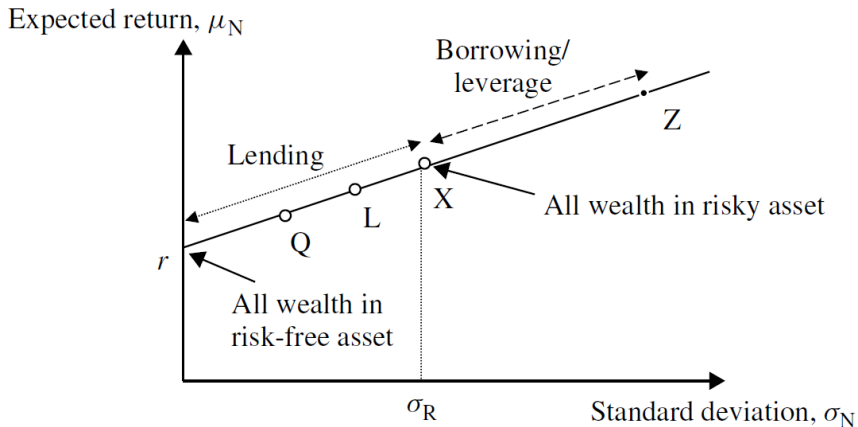
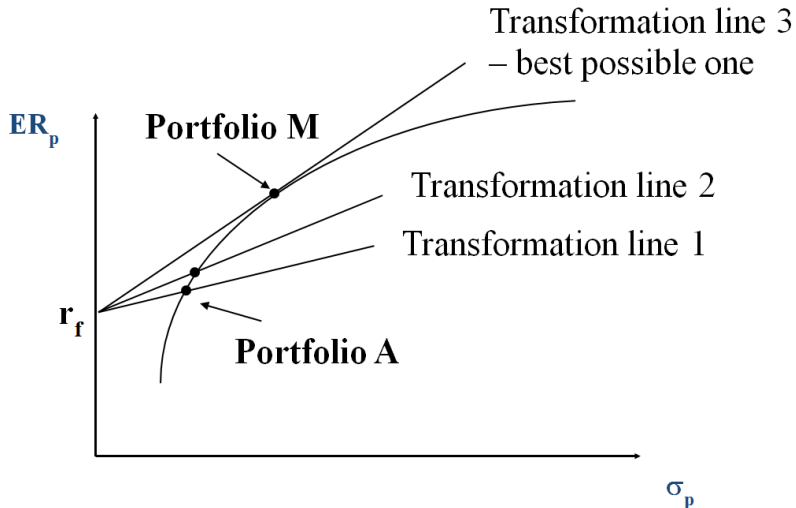


Figure 5 Transformation line



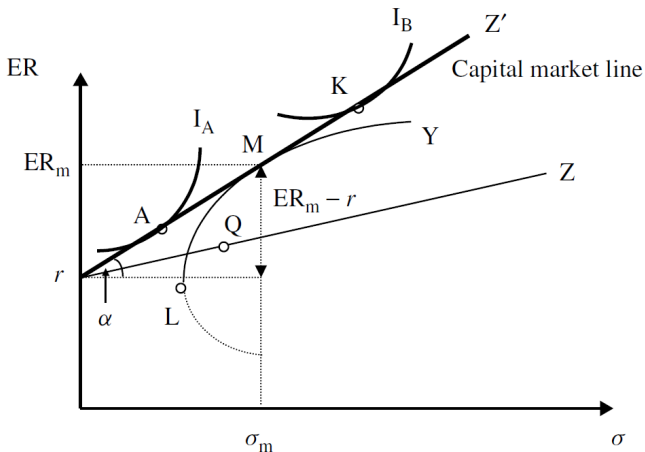


Figure 6 Portfolio choice

Separation principle

The investor makes two separate decisions:

- without any recourse to the individual's preferences, the investor determines the point of tangency, the **market portfolio**.
- the investor then determines how he will **combine** the market portfolio of risky assets with the riskless asset

The slope of the CML is called **market price of risk**.

$$\frac{dE(r_p)}{d\sigma_p} = \frac{E(r_m) - r}{\sigma_m}$$

Capital Asset Pricing Model – CAPM

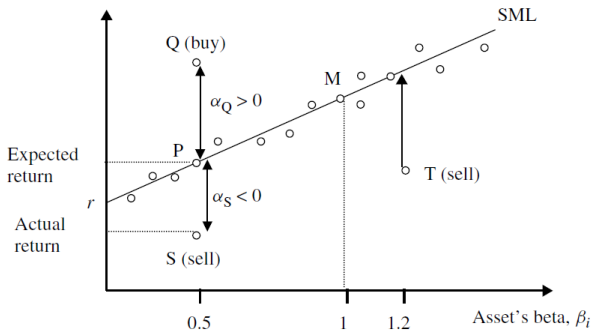
$$E(r_i) = r + \beta[E(r_m) - r]$$

$$\beta = \frac{\text{COV}(r_i, r_m)}{\sigma_m^2}$$

CAPM assumptions

- Assumption 1 :
 - Investors agree in their forecasts of expected returns, standard deviation and correlations
 - Therefore all investors optimally hold risky assets in the same relative proportions
- Assumption 2 :
 - Investors generally behave optimally. In equilibrium prices of securities adjust so that when investors are holding their optimal portfolio, aggregate demand equals its supply.

Required return = SML and actual return = \circ



Securities that lie above (below) the SML have a positive (negative) 'alpha', indicating a positive (negative) 'abnormal return', after correcting for 'beta risk'.

Figure 8 Security market line, SML

Systematic risk

$$\begin{aligned}\sigma_p &= \sum_{i=1}^n \sum_{j=1}^n w_i w_j \text{cov}(r_i, r_j) \\ &= \sum_{i=1}^n w_i^2 \text{var}(r_i) + 2 \sum_{i=1}^n \sum_{j>i} w_i w_j \text{cov}(r_i, r_j)\end{aligned}$$

For constant covariance $\text{cov}(r_i, r_j) = \text{cov}, i \neq j$ with equal variance $\sigma_i^2 = \sigma^2$, and an equal weights strategy $w_i = 1/n$ we have

$$\sigma_p^2 = \frac{n}{n^2} \sigma^2 + \frac{n(n-1)}{n^2} \text{cov} = \frac{1}{n} \sigma^2 + \left(1 - \frac{1}{n}\right) \text{cov}$$

And for the limiting case $\lim_{n \rightarrow \infty} \sigma_p^2 = \text{cov}$

Properties of Betas

Betas represent an asset's systematic (market or non-diversifiable) risk.

Beta of the market portfolio : $\beta_m = 1$

Beta of the risk-free asset: $\beta_r = 0$

Portfolio beta: $\beta_p = \sum w_i \beta_i$

Applications of betas: market timing (bull/bear markets), portfolio construction, performance measures, risk management.

The Efficient Frontier: Markowitz Model

$$\min \frac{1}{2} \sigma_p^2 = \frac{1}{2} \mathbf{w}^T \mathbf{\Omega} \mathbf{w}$$

$$\mathbf{w}^T E(\mathbf{r}) = E(r_p)$$

$$\mathbf{w}^T \mathbf{1} = 1$$

The Two-Fund Theorem: if w_1 and w_2 represent efficient portfolios, then $\alpha w_1 + (1 - \alpha)w_2$ is also an efficient portfolio for any $\alpha \in \mathbb{R}$.

Borrowing and Lending: Market Portfolio

$$\max \frac{E(r_p) - r}{\sigma_p}$$

$$\mathbf{w}^T E(\mathbf{r}) = E(r_p)$$

$$\mathbf{w}^T \mathbf{1} = 1$$

$$\sigma_p = \mathbf{w}^T \Omega \mathbf{w}$$

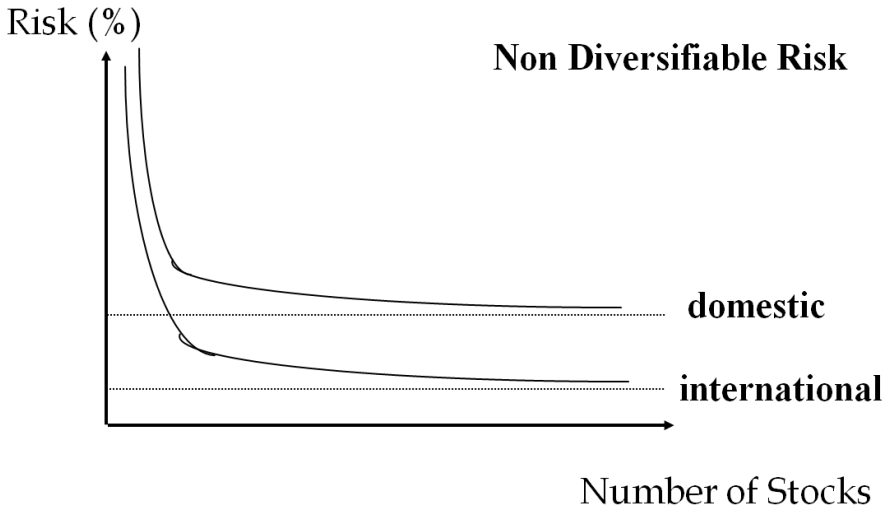
International Diversification

International investments:

- Can you enhance your risk return profile ?
 - US investors seem to overweight US stocks
 - Other investors prefer their home country (**Home country bias**)

International diversification is easy (and 'cheap')

- Improvements in technology (the internet)
- 'Customer friendly' products : Mutual funds, investment trusts, index funds



Benefits and Costs of Intl. Investments

Benefits:

- Interdependence of domestic and international stock markets
- Interdependence between the foreign stock returns and exchange rate

Costs:

- Equity risk: could be more (or less than domestic market)
- Exchange rate risk
- Political risk
- Information risk

Performance Measures / Risk Adjusted Rate of Return

■ Sharpe ratio:

$$SR_i = (E(r_i) - r) / \sigma_i$$

Risk is measured by the standard deviation (total risk of security). Aim: maximize. (CML)

■ Treynor ratio:

$$TR_i = (E(r_i) - r) / \beta_i$$

Risk is measured by beta (market risk only). Aim: maximize. (SML)

Estimating the Betas

Time series regression:

$$r_{i,t} - r_t = \alpha_i + \beta_i(r_{m,t} - r_t) + \varepsilon_{i,t}$$

Arbitrage pricing theory

$$R_{i,t} = a_i + \sum_{j=1}^k b_{i,j} F_{j,t} + \varepsilon_{i,t}$$

Note that the factors are common to all stocks, ie, they are *pervasive* risk factors. The parameters $\beta_{i,j}$, called *factor loadings*, measure the sensitivity of security i to factor j . The random variable $\varepsilon_{i,t}$ is the residual, the part of return not explained by the common factors ($\text{var}[\varepsilon_{i,j}]$ is idiosyncratic risk).

Exact factor pricing, single factor

Assume a single factor is responsible for return dynamics:

$$r_j = a_j + \beta_j F$$

Construct a portfolio consisting of a risk-free asset and the factor with weights $(w_f, w_F)^T = (1 - \beta_j, \beta_j)^T$. The portfolio return is then

$$r_p = w_f r_f + w_F F = (1 - \beta_j) r_f + \beta_j F$$

The slopes are the same, and by *no arbitrage condition*, so must be the intercepts, $a_j = (1 - \beta_j) r_f$. Thus, $r_j = r_f + \beta_j (F - r_f)$ and for the expectation

$$E(r_j) = r_f + \beta_j (E(F) - r_f)$$

When the risk factor is the market, we replace F with R_m and get CAPM.

Fama and French (1993) 3-factor model

$$E(r_j) - r_f = \beta_j[E(r_m) - r_f] + \beta_{js}E[SMB] + \beta_{jh}E[HML]$$

Small Value	Big Value
Small Neutral	Big Neutral
Small Growth	Big Growth

Thresholds: size (median), book to market equity (30th & 70th percentile)

Size premium: $SMB = (SmallValue + SmallNeutral + SmallGrowth)/3 - (BigValue + BigNeutral + BigGrowth)/3$,

Value premium:

$(SmallValue + BigValue)/2 - (SmallGrowth + BigGrowth)/2$.

<Ken French data library>

Empirical testing of CAPM

Time-series tests:

$$E(R_{i,t}) - r_t = \alpha_i + \beta_i[E(R_{m,t}) - r_t]$$

Cross-section: **Fama-MacBeth** (1973) rolling regression For any single t , run

$$\mathbf{R}_t = \alpha_t^T \mathbf{e} + \gamma_t \boldsymbol{\beta} + \theta_t \mathbf{Z}_t + \boldsymbol{\varepsilon}$$

If CAPM holds, then for any t , $\alpha_t = \theta_t = 0$ and $\gamma_t > 0$. Thus, it is easy to test for $E(\alpha_t) = 0$, $E(\theta_t) = 0$ and $E(\gamma_t) > 0$ from a sample from $t = 1, 2, \dots, T$.