



Aalto University
School of Science

MS-E2114 Investment Science

Lecture 7: Arbitrage pricing theory

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Overview

Single factor model

Multifactor models

Arbitrage Pricing Theory (APT)

Parameter estimation

Utility theory and risk aversion

Previous lecture

- ▶ The last lecture covered CAPM
- ▶ CAPM can be challenging to apply
 - ▶ The number of parameters is high
 - ⇒ For n assets, there are n expected returns, variances and $n(n - 1)/2$ covariances
 - ⇒ There are $2n + n(n - 1)/2$ parameters to be estimated
 - ⇒ Thus for $n = 500$ assets, there are 125750 parameters
 - ▶ Data may not be readily available
 - ▶ Time series may not be long enough
 - ▶ How old data is relevant and usable? (e.g., if the asset is company stock, does the appointment of a new CEO change asset characteristics?)
 - ▶ Covariances can depend on economic cycles and other external conditions

This lecture

- ▶ This lecture: Factor models and arbitrage pricing theory
- ▶ Investors may care about factors other than expected return and variance
 - ▶ E.g., preferences for geographical location, sectoral focus, ethical conduct of companies
 - ▶ Hence CAPM does not account for all factors
- ▶ Factor models help reduce the number of parameters to be estimated
- ▶ In addition, Von Neumann & Morgenstern utility theory visited briefly

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Single factor model

- ▶ Explain asset returns with an external factor
 - ▶ E.g. GDP (gross domestic product) or stock index
 - ▶ Return of asset i estimated as

$$r_i = a_i + b_i f + e_i,$$

where a_i and b_i are constants, f is the random explanatory variable and e_i is the random error term

- ▶ Assumptions:

- ▶ $\mathbb{E}[e_i] = 0$
- ▶ e_i is not correlated with f

$$\Rightarrow \mathbb{E}[(f - \bar{f})(e_i - \bar{e}_i)] = \mathbb{E}[(f - \bar{f})e_i] = 0$$

- ▶ Error terms of the assets are uncorrelated

$$\Rightarrow \mathbb{E}[(e_i - \bar{e}_i)(e_j - \bar{e}_j)] = \mathbb{E}[e_i e_j] = 0, i \neq j$$

- ▶ Variances of error terms are known (estimated)

$$\Rightarrow \mathbb{E}[e_i^2] = \sigma_{e_i}^2$$

Single factor model

- ▶ With these assumptions

$$\bar{r}_i = \mathbb{E}[r_i] = a_i + b_i \mathbb{E}[f] + \mathbb{E}[e_i]$$

$$\Rightarrow \bar{r}_i = a_i + b_i \bar{f}$$

$$\begin{aligned}\sigma_i^2 &= \text{Var}[r_i] = \mathbb{E}[(r_i - \bar{r}_i)^2] = \mathbb{E}[(a_i + b_i f + e_i - a_i - b_i \bar{f})^2] \\ &= \mathbb{E}[(b_i(f - \bar{f}) + e_i)^2] = \mathbb{E}[b_i^2(f - \bar{f})^2 + 2b_i(f - \bar{f})e_i + e_i^2]\end{aligned}$$

$$\Rightarrow \sigma_i^2 = b_i^2 \sigma_f^2 + \sigma_{e_i}^2$$

$$\begin{aligned}\sigma_{ij} &= \text{Cov}[r_i, r_j] = \mathbb{E}[(r_i - \bar{r}_i)(r_j - \bar{r}_j)] \\ &= \mathbb{E}[(b_i(f - \bar{f}) + e_i)(b_j(f - \bar{f}) + e_j)] \\ &= \mathbb{E}[b_i b_j (f - \bar{f})^2 + (b_j e_i + b_i e_j)(f - \bar{f}) + e_i e_j]\end{aligned}$$

$$\Rightarrow \sigma_{ij} = b_i b_j \sigma_f^2, i \neq j$$

Single factor model

- ▶ It follows that

$$\begin{aligned}\text{Cov}[r_i, f] &= \mathbb{E} [(r_i - \bar{r}_i)(f - \bar{f})] \\ &= \mathbb{E} [(b_i(f - \bar{f}) + e_i)(f - \bar{f})] \\ \Rightarrow \text{Cov}[r_i, f] &= b_i \sigma_f^2 \\ \Rightarrow b_i &= \frac{\text{Cov}[r_i, f]}{\sigma_f^2}\end{aligned}$$

- ▶ A total of $3n + 2$ parameters to be estimated
 - ▶ $\bar{f}, \sigma_f^2, a_i, b_i,$ and $\sigma_{e_i}^2,$ for $i = 1, 2, \dots, n$
 - ▶ CAPM has $2n + n(n - 1)/2$ parameters

Estimating a_i and b_i

- ▶ Parameters a_i and b_i can be estimated from the time series of r_i and f
 - ▶ Estimates differ depending on the selected time span
 - ▶ Averaging and other statistical methods can be used to improve accuracy
- ▶ Standard statistical estimators

$$\hat{r}_i = \frac{1}{n} \sum_{k=1}^n r_i^k$$

$$\hat{\sigma}_i^2 = \frac{1}{n-1} \sum_{k=1}^n (r_i^k - \hat{r}_i)^2$$

$$\widehat{\text{Cov}}[r_i, f] = \frac{1}{n-1} \sum_{k=1}^n (r_i^k - \hat{r}_i) (f^k - \hat{f}),$$

where superscript k denotes the number of sample

Estimating a_i and b_i

- ▶ The model parameters can be calculated using the standard estimates

$$b_i = \frac{\widehat{\text{Cov}}[r_i, f]}{\hat{\sigma}_f^2}$$

$$a_i = \hat{r}_i - b_i \hat{f}$$

- ▶ The variance of error terms become

$$\begin{aligned}\sigma_i^2 &= b_i^2 \sigma_f^2 + \sigma_{e_i}^2 \\ \Rightarrow \hat{\sigma}_{e_i}^2 &= \hat{\sigma}_i^2 - b_i^2 \hat{\sigma}_f^2\end{aligned}$$

Portfolios in the single factor model

- ▶ Form a portfolio of n assets
 - ▶ Asset i has weight w_i
- ▶ Returns of the assets follow the factor model

$$r_i = a_i + b_i f + e_i$$

- ▶ Return of the portfolio

$$r = \sum_{w=1}^n w_i r_i = \sum_{i=1}^n r_i a_i + \left(\sum_{i=1}^n w_i b_i \right) f + \sum_{i=1}^n w_i e_i = a + b f + e,$$

where

$$a = \sum_{i=1}^n w_i a_i, \quad b = \sum_{i=1}^n w_i b_i, \quad e = \sum_{i=1}^n w_i e_i$$

Portfolios in the single factor model

- ▶ For the error term of the portfolio return, we have

$$\mathbb{E}[\mathbf{e}] = \mathbb{E}\left[\sum_{i=1}^n w_i \mathbf{e}_i\right] = \sum_{i=1}^n w_i \mathbb{E}[\mathbf{e}_i]$$

$$\Rightarrow \mathbb{E}[\mathbf{e}] = 0$$

$$\text{Cov}[f, \mathbf{e}] = \mathbb{E}\left[(f - \bar{f}) \sum_{i=1}^n w_i \mathbf{e}_i\right] = \sum_{i=1}^n w_i \mathbb{E}[(f - \bar{f}) \mathbf{e}_i]$$

$$\Rightarrow \text{Cov}[f, \mathbf{e}] = 0$$

$$\text{Var}[\mathbf{e}] = \mathbb{E}\left[\left(\sum_{i=1}^n w_i \mathbf{e}_i\right) \left(\sum_{j=1}^n w_j \mathbf{e}_j\right)\right] = \sum_{i=1}^n w_i^2 \mathbb{E}[\mathbf{e}_i^2]$$

$$\Rightarrow \text{Var}[\mathbf{e}] = \sum_{i=1}^n w_i^2 \sigma_{\mathbf{e}_i}^2$$

Portfolios in the single factor model

- ▶ Assume that assets have equal weights and the variance of error terms is $\sigma_{e_i}^2 = s^2$. Then the variance of the error term of the portfolio is

$$\sigma_e^2 = \text{Var}[e] = \sum_{i=1}^n w_i^2 \sigma_{e_i}^2 = \sum_{i=1}^n \frac{1}{n} s^2 = \frac{1}{n} s^2,$$

and the variance of the portfolio return is

$$\sigma^2 = \text{Var}[r] = b^2 \sigma_f^2 + \sigma_e^2,$$

where $\sigma_e^2 \rightarrow 0$ as $n \rightarrow \infty$

- ▶ Variance related to the error terms e_i can be diversified
- ▶ Variance related to terms $b_i f$ cannot be diversified

Single factor model and CAPM

- ▶ CAPM is a special case of the single factor model
 - ▶ Factor = return of the market portfolio

$$r_i - r_f = \alpha_i + \beta_i(r_M - r_f) + e_i$$

- ▶ This forms the **characteristic line** (set $e_i = 0$)
- ▶ Taking the expectation gives

$$\Rightarrow \bar{r}_i - r_f = \alpha_i + \beta_i(\bar{r}_M - r_f)$$

- ▶ In CAPM, the intercept $\alpha_i = 0$
 - ▶ See Lecture 6, Jensen Index
- ▶ Covariance of $r_i - r_f$ with r_M is

$$\sigma_{iM} = \text{Cov}[r_i - r_f, r_M] = \text{Cov}[\alpha_i + \beta_i(r_M - r_f) + e_i, r_M] = \beta_i \sigma_M^2$$
$$\Rightarrow \beta_i = \frac{\sigma_{iM}}{\sigma_M^2}$$

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Multifactor models

- ▶ The return can be explained with more than one factor
- ▶ With two factors

$$r_i = a_i + b_{1i}f_1 + b_{2i}f_2 + e_i,$$

where a_i is the intercept and b_{1i}, b_{2i} are factor loadings

- ▶ Assumptions
 - ▶ Expected error $\mathbb{E}[e_i] = 0$
 - ▶ Error terms are uncorrelated with factors and each other
 - ▶ **Factors can correlate with each other**

Multifactor models

- ▶ Expected return in the two factor model

$$\bar{r}_i = \mathbb{E}[r_i] = a_i + b_{1i}\bar{f}_1 + b_{2i}\bar{f}_2$$

- ▶ Covariance

$$\begin{aligned} \text{Cov}[r_i, r_j] &= \mathbb{E} \left[(b_{1i}(f_1 - \bar{f}_1) + b_{2i}(f_2 - \bar{f}_2) + e_i) \right. \\ &\quad \left. (b_{1j}(f_1 - \bar{f}_1) + b_{2j}(f_2 - \bar{f}_2) + e_j) \right] \\ &= \begin{cases} b_{1i}b_{1j}\sigma_{f_1}^2 + (b_{1i}b_{2j} + b_{2i}b_{1j})\sigma_{f_1, f_2} + b_{2i}b_{2j}\sigma_{f_2}^2, & i \neq j \\ b_{1i}^2\sigma_{f_1}^2 + 2b_{1i}b_{2j}\sigma_{f_1, f_2} + b_{2i}^2\sigma_{f_2}^2 + \sigma_{e_i}^2, & i = j \end{cases} \end{aligned}$$

Estimating the factor model parameters

- ▶ Loadings b_{1i} , b_{2i} can be estimated from the covariance matrix

$$\begin{aligned}\text{Cov}[r_i, f_1] &= \mathbb{E} [(b_{1i}(f_1 - \bar{f}_1) + b_{2i}(f_2 - \bar{f}_2) + e_i) (f_1 - \bar{f}_1)] \\ &= b_{1i}\sigma_{f_1}^2 + b_{2i}\sigma_{f_1, f_2}\end{aligned}$$

$$\text{Cov}[r_i, f_2] = b_{2i}\sigma_{f_2}^2 + b_{1i}\sigma_{f_1, f_2}$$

- ▶ Solve these equations for b_{1i} and b_{2i}
- ▶ The use of multiple factors is warranted if a single factor model has a large error term variance
 - ▶ If the error term variance is nearly as high as the variance of returns, the factor model does not explain much
 - ▶ Too many factors leads to overfitting \Rightarrow Poor predictions power
- ▶ Factor models are easier to estimate than CAPM
 - ▶ There are fewer parameters
 - ▶ New factors can be introduced as necessary

Selection of factors

- ▶ No unambiguous answer - depends on what the key factors are believed to be
- ▶ External factors, such as
 - ▶ Gross National Product (GNP)
 - ▶ Consumer price indices
 - ▶ Unemployment rate
- ▶ Extracted factors, such as
 - ▶ Market portfolio return
 - ▶ Average return of companies in one industry
 - ▶ Days since the last market peak
- ▶ Firm characteristics, such as
 - ▶ Price-earnings ratio
 - ▶ Dividend payout ratio
 - ▶ Earnings forecast

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Arbitrage Pricing Theory (APT)

- ▶ APT is a pricing theory (complementary to CAPM)
- ▶ Key assumptions
 - ▶ Investors prefer higher returns
 - ▶ There is a large number of assets
 - ▶ The investors need not optimize with respect to mean-variance (unlike what CAPM assumes)
- ▶ Idea: There is a factor price for each factor
 - ▶ Price of asset follows from linearity of pricing
 - ▶ Factor prices derived based on the assumption of no arbitrage
- ▶ A single factor with two assets and no error term

$$r_i = a_i + b_i f,$$

$$r_j = a_j + b_j f$$

Arbitrage Pricing Theory (APT)

- ▶ Invest in assets i (weight $w_i = w$) and j ($w_j = 1 - w$) that follow single factor model
- ▶ Portfolio return

$$\begin{aligned}r &= w(a_i + b_i f) + (1 - w)(a_j + b_j f) \\ &= wa_i + (1 - w)a_j + (wb_i + (1 - w)b_j) f\end{aligned}$$

- ▶ Select the weight w so that the coefficient of factor f is 0

$$\begin{aligned}wb_i + (1 - w)b_j &= 0 \\ \Rightarrow w &= \frac{b_j}{b_j - b_i}\end{aligned}$$

Arbitrage Pricing Theory (APT)

- ▶ The portfolio with coefficient 0 for factor f is risk free (no variance), hence its return must be $r_f = \lambda_0$

$$\begin{aligned}r &= \frac{b_j}{b_j - b_i} a_i + \left(1 - \frac{b_j}{b_j - b_i}\right) a_j \\&= \frac{b_j}{b_j - b_i} a_i - \frac{b_i}{b_j - b_i} a_j = \lambda_0 \\&\Rightarrow b_j a_i - b_i a_j = \lambda_0 (b_j - b_i) \\&\Rightarrow b_j (a_i - \lambda_0) = b_i (a_j - \lambda_0) \\&\Rightarrow \frac{a_i - \lambda_0}{b_i} = \frac{a_j - \lambda_0}{b_j}\end{aligned}$$

Arbitrage Pricing Theory (APT)

- ▶ Thus, for every asset i , ratio $(a_i - \lambda_0)/b_i$ must be equal to some constant c

$$\Rightarrow \frac{a_i - \lambda_0}{b_i} = c$$

$$\Leftrightarrow a_i = \lambda_0 + b_i c$$

- ▶ Thus

$$\begin{aligned}\bar{r}_i &= a_i + b_i \bar{f} = \lambda_0 + b_i c + b_i \bar{f} \\ &= \lambda_0 + b_i (c + \bar{f}) = \lambda_0 + b_i \lambda_1,\end{aligned}$$

which can be generalized to several factors

Arbitrage Pricing Theory (APT)

Definition

(**Simple APT**) Suppose that there are n assets whose rates of return are governed by $m < n$ factors according to the equation

$$r_i = a_i + \sum_{j=1}^m b_{ij} f_j$$

for $i = 1, 2, \dots, n$. Then there are constants $\lambda_0, \lambda_1, \dots, \lambda_m$ such that

$$\bar{r}_i = \lambda_0 + \sum_{j=1}^m b_{ij} \lambda_j$$

for $j = 1, 2, \dots, m$.

- ▶ λ_j =price of risk associated with factor j (i.e., factor price)
- ▶ b_{ij} = factor loading j of asset i

Well-diversified portfolio

- ▶ Adding the error term e_i to the simple APT gives

$$r_i = a_i + \sum_{j=1}^m b_{ij}f_j + e_i$$

- ▶ Variance of error terms (slide 12)

$$\sigma_e^2 = \sum_{i=1}^n w_i^2 \sigma_{e_i}^2$$

- ▶ Assume that all variances are bounded, that is

$$\sigma_{e_i}^2 \leq S^2$$

for some S , and assume that all assets have similar weights (i.e., we have $w_i \leq W/n$ for some $W \approx 1$)

- ▶ This means that the portfolio is well diversified

Well-diversified portfolio

- ▶ With the assumptions of similar and bounded weights, we have

$$\sigma_e^2 = \sum_{i=1}^n w_i^2 \sigma_{e_i}^2 \leq \sum_{i=1}^n \frac{W^2}{n^2} S^2 = \frac{1}{n} W^2 S^2$$

$$\Rightarrow \lim_{n \rightarrow \infty} \sigma_e^2 = 0$$

- ▶ Hence, a well-diversified portfolio has practically no non-diversifiable risk
- ▶ As a result, the return of well diversified portfolios are fully explained by the factor model (because the error terms tend to cancel out)

$$r = a + \sum_{j=1}^m b_j f_j,$$

General APT

- ▶ By simple APT, the expectation of the return of well a diversified portfolio is

$$\bar{r} = \lambda_0 + \sum_{j=1}^m b_j \lambda_j,$$

- ▶ Because different well-diversified portfolios can be formed with weights that differ on only a small number of basic assets, it follow that these individual assets must also satisfy

$$\bar{r}_i = \lambda_0 + \sum_{j=1}^m b_{ij} \lambda_j,$$

- ▶ This pricing equation is referred to as the general APT (this argumentation is not technically rigorous, but the above argument is the key idea)

APT and CAPM

- ▶ Consider the two factor model

$$r_i = a_i + b_{1i}f_1 + b_{2i}f_2 + e_i$$

$$\begin{aligned}\text{Cov}[r_M, r_i] &= \mathbb{E}[(r_M - \bar{r}_M)(b_{1i}(f_1 - \bar{f}_1) + b_{2i}(f_2 - \bar{f}_2) + e_i)] \\ &= b_{1i} \text{Cov}[r_M, f_1] + b_{2i} \text{Cov}[r_M, f_2] + \text{Cov}[r_M, e_i]\end{aligned}$$

- ▶ Because in APT n is large (market portfolio well diversified), we have $\text{Cov}[r_M, e_i] \approx 0$
- ▶ Dividing by σ_M^2 gives the beta of an asset

$$\begin{aligned}\beta_i &= b_{1i} \frac{\text{Cov}[f_1, r_M]}{\sigma_M^2} + b_{2i} \frac{\text{Cov}[f_2, r_M]}{\sigma_M^2} \\ &= b_{1i}\beta_{f_1} + b_{2i}\beta_{f_2}\end{aligned}$$

- ▶ Difference between CAPM and APT
 - ▶ In CAPM, the β of an asset is the sum of the factors' betas
 - ▶ Factors have β :s and assets have loading factors only (i.e., they correlate with market through the factors)

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Parameter estimation

- ▶ Annual return r_y formed from monthly returns r_1, r_2, \dots, r_{12}

$$1 + r_y = (1 + r_1)(1 + r_2) \cdots (1 + r_{12})$$

- ▶ Assume that monthly returns are small:

$$1 + r_y \approx 1 + r_1 + r_2 + \cdots + r_{12}$$

$$\Rightarrow r_y = r_1 + r_2 + \cdots + r_{12}$$

- ▶ Monthly returns are equally distributed and uncorrelated

$$\bar{r}_y = 12\bar{r}$$

$$\sigma_y^2 = \mathbb{E} \left[\left(\sum_{i=1}^{12} (r_i - \bar{r}) \right)^2 \right] = \mathbb{E} \left[\sum_{i=1}^{12} (r_i - \bar{r})^2 \right] = 12\sigma^2$$

Parameter estimation

- ▶ Even periods of other length are possible
- ▶ If there are p periods in a year, then

$$\bar{r}_p = p\bar{r}_y$$

$$\sigma_p = \sigma_y\sqrt{p}$$

- ▶ When p becomes smaller, the ratio of standard deviation and expected value increases
 - ⇒ Finding short term estimators becomes more difficult
 - ▶ If the yearly parameters are $\mathbb{E}[r_y] = 12\%$ and $\sigma_y = 15\%$, the monthly parameters $p = 1/12$ are $\mathbb{E}[r] = 1\%$ and $\sigma = 1/\sqrt{12} \cdot 15\% = 4.33\%$
 - ▶ The one month return is within the interval $1 \pm 4.33\%$ with a 68% probability, which is a rather wide confidence interval
- ▶ Thus, single period expected returns are hard to estimate reliably even if the time series are long

Parameter estimation

- ▶ Let the time series consist of n months with monthly expected return $\mathbb{E}[r]$ and standard deviation σ
- ▶ Unbiased estimator

$$\hat{r} = \frac{1}{n} \sum_{i=1}^n r_i$$

$$\Rightarrow \mathbb{E}[\hat{r}] = \frac{1}{n} \sum_{i=1}^n \mathbb{E}[r_i] = \frac{1}{n} \sum_{i=1}^n \bar{r} = \bar{r}$$

- ▶ Variance and standard deviation of the estimator are

$$\sigma_{\hat{r}}^2 = \mathbb{E}[(\hat{r} - \bar{r})^2] = \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n (r_i - \bar{r})\right]^2 = \frac{1}{n} \sigma^2$$

$$\Rightarrow \sigma_{\hat{r}} = \frac{1}{\sqrt{n}} \sigma$$

Parameter estimation

- ▶ Standard deviation $\sigma_{\hat{r}}$ of estimator \hat{r} decreases slowly with n , because \sqrt{n} is in its denominator
 - ▶ Let monthly $\mathbb{E}[r] = 1\%$ and $\sigma = 4.33\%$ and consider $n = 12$ monthly time series

$$\sigma_{\hat{r}} = \frac{1}{\sqrt{12}} 4.33\% = 1.25\%$$

- ▶ Should we want to estimate the standard deviation which is within 10% of the expected returns ($0.1 \cdot 1\% = 0.10\%$), then we would need a time series of 156 years and 3 months

$$\sigma_{\hat{r}} = \frac{1}{\sqrt{12}} 4.33\% = 0.10\%$$
$$\Rightarrow n = 1875 = 12 \cdot 156.25$$

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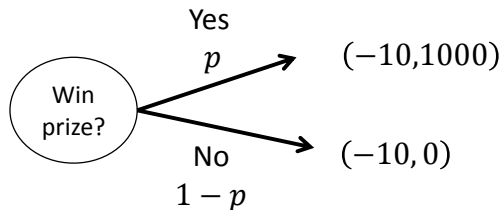
Utility theory and risk aversion

Investor's risk preferences

- ▶ We have discussed the formation of efficient portfolios
 - ▶ Mean-variance portfolio theory
 - ▶ CAPM
- ▶ Which one out of the many efficient portfolios should an investor select?
- ▶ Specifically, portfolios are characterized by the expected return and risk (standard deviation), that is, by tuple (r, σ)
 - ⇒ Which combination of these parameters should be selected?

Uncertain cash flows

- ▶ Cash flows typically uncertain
 - ▶ E.g. price of stock at a future date
- ▶ Modeled with random cash flows
 - ▶ $x \sim D$, where D is the distribution of the random vector x
 - ▶ E.g., consider a lottery ticket which costs 10 € and gives a 1000 € prize with probability p



Expected utility theory

(Von Neumann & Morgenstern 1944)

- ▶ Rational investors' preferences under uncertainty are consistent with a utility function $U : R \rightarrow R$
 - ▶ Wealth level W_1 preferred to wealth level W_2 if and only if

$$\mathbb{E}[U(W_1)] > \mathbb{E}[U(W_2)]$$

- ▶ Utility functions are
 - ▶ Non-decreasing and continuous
 - ▶ Unique up to positive affine transformations
 - ⇒ $U(W)$ and $V(W)$ yield same preferences if and only if

$$U(W) = aV(W) + b,$$

where $a > 0$

Utility theory example

- ▶ Investor invests in either
 - ▶ A: bank account for a profit of 6 k€
 - ▶ B: Stock that yields a profit of
 - ▶ 10 k€ (probability 0.2)
 - ▶ 5 k€ (probability 0.4)
 - ▶ 1 k€ (probability 0.4)
- ▶ Her utility function is $U(x) = \sqrt{x}$, where unit of x is k€

$$\mathbb{E}[U(A)] = U(6) = 2.45$$

$$\mathbb{E}[U(B)] = 0.2U(10) + 0.4U(5) + 0.4U(1) = 1.93$$

⇒ A is preferred to B, because

$$\mathbb{E}[U(A)] = 2.45 > 1.93 = \mathbb{E}[U(B)]$$

Widely used utility functions

- ▶ Linear

$$U(W) = W$$

- ▶ Exponential ($a > 0$)

$$U(W) = -e^{-aW}$$

- ▶ Logarithmic

$$U(W) = \ln W$$

- ▶ Power ($b \leq 1, b \neq 0$)

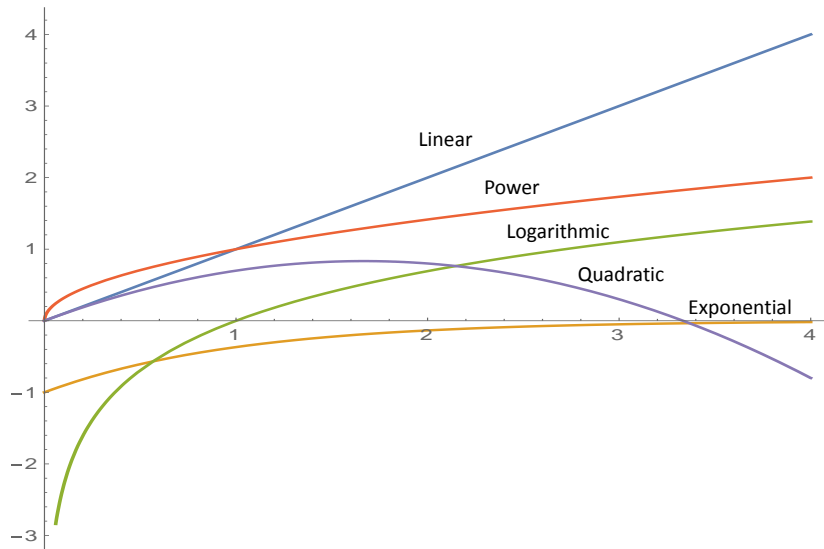
$$U(W) = bW^b$$

- ▶ Quadratic

$$U(W) = W - bW^2$$

(increasing for $W < 1/(2b)$)

Widely used utility functions



Risk aversion

- ▶ Investor with utility function U is:
 - ▶ Risk neutral, if U is linear
 - ▶ Risk averse, if U is concave

$$U(\lambda x + (1 - \lambda)y) \geq \lambda U(x) + (1 - \lambda)U(y)$$

for all $0 \leq \lambda \leq 1$

- ▶ The certainty equivalent of a random wealth X is the certain wealth C for which

$$U(C) = \mathbb{E} [U(x)]$$

- ▶ E.g, for a 50% chance to win 100 € and 50% chance of winning nothing, the certainty equivalent could be $C = 40\text{€}$
- ▶ Certainty equivalent can be calculated as

$$C = U^{-1} (\mathbb{E} [U(x)])$$

Risk aversion coefficient

- ▶ Arrow-Pratt risk aversion coefficient

$$a(x) = -\frac{U''(x)}{U'(x)}$$

- ▶ Measures degree of risk aversion with the normalized concavity
- ▶ For the exponential utility function, risk aversion coefficient is constant

$$U(x) = -e^{-bx}$$
$$\Rightarrow a(x) = b$$

- ▶ For logarithmic utility function, risk aversion decreases with wealth

$$U(x) = \ln x$$
$$\Rightarrow a(x) = \frac{1}{x}$$

Elicitation of utility functions

- ▶ The utility function may help the investor choose investments that suit her
- ▶ Elicitation methods
 - ▶ Ask for certainty equivalents to get the value of U for different x
 - ▶ Select the functional form of utility function, fix some parameters to 1, proceed by carrying out more utility assessments
 - ▶ Questionnaires (Luenberger p. 238)

Utility function and the mean-variance criterion

- ▶ Risk aversion is related to the mean-variance criterion
- ▶ Example: Assume quadratic utility

$$U(x) = ax - \frac{1}{2}bx^2, \quad \text{where } a > 0, b \geq 0$$

- ▶ This is increasing for $x \leq a/b$
- ▶ Assume that the initial wealth level is 0 (the result can be extended for positive wealth levels)
- ▶ Portfolio with random wealth y has

$$\mathbb{E}[U(y)] = \mathbb{E}\left[ay - \frac{1}{2}by^2\right] = a\mathbb{E}[y] - \frac{1}{2}b\mathbb{E}[y^2]$$

$$\Rightarrow \mathbb{E}[U(y)] = a\mathbb{E}[y] - \frac{1}{2}b\mathbb{E}[y]^2 - \frac{1}{2}b\text{Var}[y]$$

- ⇒ If one seeks to attain a given expected return $\mathbb{E}[y]$, the portfolio is selected by minimizing variance subject to this requirement

Overview

Single factor model

Multifactor models

Arbitrage Pricing Theory (APT)

Parameter estimation

Utility theory and risk aversion