Financial Markets and Investments
Formulas

Basic Concepts: $\qquad R_i$ – random return of asset i ; $\qquad \bar{R}_i = \mathbb{E}(R_i)$;

$$\sigma_i^2 = \mathrm{Var}(R_i) = \mathbb{E}[(R_i - \bar{R}_i)^2] = \mathbb{E}(R_i^2) - [\mathbb{E}(R_i)]^2 \; ; \quad \Rightarrow \quad \sigma_i = \sqrt{\mathrm{Var}(R_i)} \; ;$$

$$\sigma_{ij} = \mathrm{Cov}(R_i, R_j) = \mathbb{E}[(R_i - \bar{R}_i)(R_j - \bar{R}_j)] \; ; \qquad\qquad \Rightarrow \rho_{ij} = \frac{\sigma_{ij}}{\sigma_i \sigma_j} \; .$$

Portfolios P (n risky assets): x_i – proportion of initial wealth W_0 invested in asset i.

$$\bar{R}_P = \sum_{i=1}^{n} x_i \bar{R}_i \qquad\qquad \sigma_P^2 = \sum_{i=1}^{n}\sum_{j=1}^{n} x_i x_j \sigma_{ij} = \sum_{i=1}^{n} x_i^2 \sigma_i^2 + \sum_{i=1}^{n}\sum_{j=1,j\neq i}^{n} x_i x_j \sigma_{ij}$$

Homogeneous portfolios H :
$$\sigma_H^2 = \frac{1}{n}\overbrace{\left[\sum_{i=1}^{n}\left(\frac{\sigma_i^2}{n}\right)\right]}^{\overline{\sigma_i^2}} + \frac{n-1}{n}\overbrace{\left[\sum_{i=1}^{n}\sum_{j=1,j\neq i}^{n}\frac{\sigma_{ij}}{n(n-1)}\right]}^{\overline{\sigma_{ij}}}$$

Mean-Variance Theory (MVT) $\qquad\qquad (n = 2) \Rightarrow \qquad x_1^{MV} = \dfrac{\sigma_2^2 - \sigma_{12}}{\sigma_1^2 + \sigma_2^2 - 2\sigma_{12}}$

Vector notation (n risky assets):

$$\bar{R} = \begin{pmatrix} \bar{R}_1 \\ \bar{R}_2 \\ \vdots \\ \bar{R}_n \end{pmatrix} \quad V = \begin{pmatrix} \sigma_1^2 & \sigma_{12} & \cdots & \sigma_{1n} \\ \sigma_{21} & \sigma_2^2 & \cdots & \sigma_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \sigma_{n1} & \sigma_{n2} & \cdots & \sigma_n^2 \end{pmatrix} \quad X = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \quad \mathbb{1} = \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix}$$

- Portfolio ($X_P'\mathbb{1} = 1$): $\quad \bar{R}_P = X_P'\bar{R} \qquad \sigma_P^2 = X_P'VX_P \qquad \mathrm{Cov}(R_{P1}, R_{P2}) = X_{P1}'VX_{P2}$

- Hyperbola: $\qquad\qquad\qquad\qquad \sigma_P^2 = \dfrac{A\bar{R}_P^2 - 2B\bar{R}_P + C}{AC - B^2} \qquad \text{where} \qquad \begin{aligned} A &= \mathbb{1}'V^{-1}\mathbb{1} \\ B &= \mathbb{1}'V^{-1}\bar{R} \\ C &= \bar{R}'V^{-1}\bar{R} \end{aligned}$

- Minimum Variance Portfolio: $\qquad X_{MV} = \dfrac{1}{A}V^{-1}\mathbb{1}$

- Tangent Portfolio
    - Shortselling allowed: $\qquad Z = V^{-1}\left[\bar{R} - R_f\mathbb{1}\right]$
        - unlimited: $\qquad\qquad \Rightarrow \quad x_i^T = \dfrac{z_i}{\sum_{i=1}^{n} z_i} \qquad X_T = \dfrac{Z}{Z'\mathbb{1}}$
        - restricted (Lintner): $\qquad \Rightarrow \quad x_i^T = \dfrac{z_i}{\sum_{i=1}^{n} |z_i|} \qquad X_T = \dfrac{Z}{|Z|'\mathbb{1}}$
        - real-life restrictions: $\quad n = 2 \Rightarrow$ trivial , $\quad n \geq 3 \Rightarrow$ *numerical solution.*
    - Shortselling forbidden: $\qquad n = 2 \Rightarrow$ trivial , $\quad n \geq 3 \Rightarrow$ *numerical solution.*

MVT – return generating models:

- Constant Correlation Model: $\rho_{ij} = \rho \quad \forall i, j$

$$\text{Ranking} = \frac{\bar{R}_i - R_f}{\sigma_i} \ ; \ C_k = \frac{\rho \sum_{i=1}^{k} \left(\frac{\bar{R}_i - R_f}{\sigma_i} \right)}{1 - \rho + k\rho}(\text{cut-off}) \ ; \ Z_i = \frac{1}{(1-\rho)\sigma_i} \left(\frac{\bar{R}_i - R_f}{\sigma_i} - C^* \right)$$

- Single Index Model: $R_i = \alpha_i + \beta_i R_M + e_i \ ; \quad \sigma_i^2 = \beta_i^2 \sigma_M^2 + \sigma_{e_i}^2 \ ; \quad \sigma_{ij} = \beta_i \beta_j \sigma_M^2$.

    Blume's adjust.: $\beta_{2i} = a + b\beta_{1i}$ \quad Vasicek's adjust.: $\beta_{2i} = \frac{\sigma_{\beta_{1i}}^2}{\sigma_{\beta_{1i}}^2 + \sigma_{\bar{\beta}_1}^2} \bar{\beta}_1 + \frac{\sigma_{\bar{\beta}_1}^2}{\sigma_{\beta_{1i}}^2 + \sigma_{\bar{\beta}_1}^2} \beta_{1i}$

$$\text{Ranking} = \frac{\bar{R}_i - R_f}{\beta_i} \ ; \ C_k = \frac{\sigma_m^2 \sum_{i=1}^{k} \frac{(\bar{R}_i - R_f)\beta_i}{\sigma_{e_i}^2}}{1 + \sigma_m^2 \sum_{i=1}^{k} \frac{\beta_i^2}{\sigma_{e_i}^2}}(\text{cut-off}) \ ; \ Z_i = \frac{\beta_i}{\sigma_{e_i}^2} \left(\frac{\bar{R}_i - R_f}{\beta_i} - C^* \right)$$

- Multi-index Model: $R_i = a_i + \sum_{k=1}^{K} b_{ik} I_k + c_i; \ \sigma_i^2 = \sum_{k=1}^{K} b_{ik}^2 \sigma_{I_k}^2 + \sigma_{c_i}^2; \ \sigma_{ij} = \sum_{k=1}^{K} b_{ik} b_{jk} \sigma_{I_k}^2$

Investor Preferences - Expected Utility Theory

- For the utility function $U(W)$

$$\text{ARA: } A(W) = -\frac{U''(W)}{U'(W)} \ ; \qquad \text{RRA: } R(W) = W A(W)$$

    2nd ord. Taylor approx. : $U(W) \approx U(W_0) + U'(W_0)(W - W_0) + \frac{1}{2} U''(W_0)(W - W_0)^2$

- Risk tolerance function (RTF) with domain $(\sigma, \bar{R})$ is $f(\sigma, \bar{R}) = \mathbb{E}[U(W)]$.
    2nd ord. Taylor approx. : $f(\sigma, \bar{R}) \approx \bar{R} - \frac{1}{2} RRA(W_0) \left(\bar{R}^2 + \sigma^2 \right)$

- The indifference curves of the RTF are : $f(\sigma, \bar{R}) = K$, for constant levels K.

Investor Preferences - other \quad Geometric average: $\bar{R}_i^G = \prod_{m=1}^{M} (1 + R_{ij})^{p_{ij}} - 1$;

"Safety First": (Roy) $\min \Pr(R_p < R_L)$, \quad (Kataoka) $\max R_L$, \qquad (Telser) $\max \bar{R}_p$,
$$s.a. \ \Pr(R_p < R_L) \leq \alpha \quad s.a. \ \Pr(R_p \leq R_L) \leq \alpha$$

Equilibrium Models

<u>CAPM:</u> \quad $\bar{R}_i^{eq} = a + \beta_i b \quad \Rightarrow \quad \bar{R}_i^{eq} = R_f + \beta_i \left(\bar{R}_M - R_f \right)$; \quad $\beta_i = \frac{\sigma_{iM}}{\sigma_M^2}$

<u>APT:</u> \qquad $\bar{R}_i^{eq} = \lambda_0 + \sum_{j=1}^{J} b_{ij} \lambda_j \quad \Rightarrow \bar{R}_i^{eq} = R_f + b_{i1} \left(\bar{I}_1 - R_f \right) + \sum_{j=1}^{J} b_{ij} \left(\bar{I}_j - R_f \right)$

Performance indicators:

Sharpe: $SR = \frac{\bar{R}_p - R_f}{\sigma_p}$; \quad Treynor: $TY = \frac{\bar{R}_p - R_f}{\beta_p}$; \quad Jensen: $J = R_p - \left(R_f + \beta_p [\bar{R}_M - R_f] \right)$