

CASE STUDY: Portfolio Optimization with Exponential, Logarithmic, and Linear-Quadratic Utilities (exp_eut, log_eut, pm2_pen, avg_g)

Background

Utility functions are quite popular in various financial applications. This case study compares portfolio optimization problems with Exponential, Logarithmic, and Linear-Quadratic utility functions. The rate of return dataset for a portfolio is provided for benchmarking purposes by EpiRisk Research company via Drs. Roger Wets and Michael Tian. The EpiRisk Research relies on letting the manager of a fixed-income portfolio solve a sequence of so-called tacking (optimization) models, described below, to shape the returns' distribution. The shape of the distribution is adjusted by selecting the coefficients of the appraisal (\sim utility) function.

Notations

I = number of instruments in the portfolio; $i = \{1, \dots, I\}$ index of instruments in the portfolio;

J = number of scenarios; $j = \{1, \dots, J\}$ index of scenarios;

x_i = portion of wealth invested in instrument i , $i = \{1, \dots, I\}$;

$x = (x_1, \dots, x_I)$ = vector of decision variables;

l_i = lower bound on the portion of wealth invested in instrument i , $i = \{1, \dots, I\}$;

u_i = upper bound on the portion of wealth invested in instrument i , $i = \{1, \dots, I\}$;

p_i^0 = price of instrument i at initial time 0;

$p^0 = (p_1^0, \dots, p_I^0)$ = price-vector at initial time 0;

p_{ji}^1 = price of instrument i for scenario j at time 1;

$\theta_{ji} = (p_{ji}^1 - p_i^0) / p_i^0$ = rate of return of instrument i for scenario j ;

$\tilde{\theta}_{ji} = p_{ji}^1 / p_i^0 = \theta_{ji} + 1$ = return of instrument i for scenario j ;

θ_i = random value having J equally probable scenarios of rates of return, $\{\theta_{1i}, \dots, \theta_{ji}\}$, $i = 1, \dots, I$;

$\tilde{\theta}_i$ = random value having J equally probable scenarios of returns, $\{\tilde{\theta}_{1i}, \dots, \tilde{\theta}_{ji}\}$, $i = 1, \dots, I$;

$\theta = (\theta_1, \dots, \theta_I)$ = random vector of rates of return;

$\tilde{\theta} = (\tilde{\theta}_1, \dots, \tilde{\theta}_I)$ = random return vector;

$\theta_j = (\theta_{j1}, \dots, \theta_{jI})$ = vector of rates of return for scenario j ;

$\tilde{\theta}_j = (\tilde{\theta}_{j1}, \dots, \tilde{\theta}_{jI})$ = return vector for scenario j ;

$R(\theta_j, x) = \sum_{i=1}^I \theta_{ji} x_i$ = rate of return of the portfolio for scenario j ;

$R(\tilde{\theta}_j, x) = \sum_{i=1}^I \tilde{\theta}_{ji} x_i = \sum_{i=1}^I (\theta_{ji} + 1) x_i = R(\theta_j, x) + \sum_{i=1}^I x_i$ = return of the portfolio for scenario j ;

$$R(\tilde{\theta}_j, x) = R(\theta_j, x) + 1, \text{ if } \sum_{i=1}^I x_i = 1;$$

$$avg_g(R(\theta, x)) = E[R(\theta, x)] = \text{Average Gain function};$$

$pm2_pen(-R(\theta, x), w) = E(\max\{0, -R(\theta, x) - w\})^2$ = Partial Moment Two Penalty function, where w is a threshold value;

$$exp_eut(R(\theta, x)) = E[-e^{-aR(\theta, x)}] = \text{Exponential utility function, where } a > 0;$$

$$log_eut(R(\tilde{\theta}, x)) = E[\ln(R(\tilde{\theta}, x))] = \text{Logarithmic utility function.}$$

$E\{F(R(\theta, x))\}$ = Linear-Quadratic utility function, where

$F(R)$ = an appraisal (= monitoring) function, a linear-quadratic function modelling the decision maker attitude towards risk.

$F(R)$ is defined as follows:

$$F(R) = \begin{cases} -qR - \frac{1}{2}[u(r-s) - q(r+s)], & \text{when } R < s \\ [(u+q)R^2 - 2(qr+us)R - \tilde{k}] / [2(r-s)], & \text{when } R \in [s, r]; \\ u(R-r), & \text{when } R \geq r \end{cases}$$

r = the hurdle rate;

s = the sub-hurdle rate, $s < r$;

u, q = constants, $q < -u < 0$;

$\tilde{k} = (ur^2 - qr^2 - 2usr) = \text{constant};$

The function $F(R)$ can be represented as follows:

$$F(R) = F_1(R) + F_2(R) + F_3(R);$$

$$F_1(R) = u(R-r), -\infty < R < \infty;$$

$$F_2(R) = \begin{cases} 0, & \text{when } R > r \\ \frac{u+q}{2(r+s)}(R-r)^2, & \text{when } R \leq r \end{cases};$$

$$F_3(R) = \begin{cases} 0, & \text{when } R > s \\ -\frac{u+q}{2(r+s)}(R-r)^2, & \text{when } R \leq s \end{cases}$$

Optimization Problem 1

Maximizing Exponential Utility

$$\max \left[\exp_eut \left(R(\theta, x) \right) \right] \quad (\text{CS.1})$$

subject to

budget constraint

$$\sum_{i=1}^I x_i = 1, \quad (\text{CS.2})$$

bounds on positions

$$l_i \leq x_i \leq u_i, \quad i = 1, 2, \dots, I. \quad (\text{CS.3})$$

Optimization Problem 2

Maximizing Linear-Quadratic Utility

$$\begin{aligned} \max E \{ F(R(\theta, x)) \} = \\ = \max \{ u \cdot \text{avg_g}(R(\theta, x)) + \\ + \frac{u+q}{2(r-s)} \text{pm2_pen}(-R(\theta, x), -r) - \\ - \frac{u+q}{2(r-s)} \text{pm2_pen}(-R(\theta, x), -s) \} \end{aligned} \quad (\text{CS.4})$$

subject to

budget constraint

$$\sum_{i=1}^I x_i = 1, \quad (\text{CS.5})$$

bounds on positions

$$l_i \leq x_i \leq u_i, \quad i = 1, 2, \dots, I. \quad (\text{CS.6})$$

Optimization Problem 3

Maximizing Logarithmic Utility

$$\max \left[\log_eut \left(R(\tilde{\theta}, x) \right) \right] = \max \left[\log_eut \left(R(\theta, x) + 1 \right) \right] \quad (\text{CS.7})$$

subject to

budget constraint

$$\sum_{i=1}^I x_i = 1, \quad (\text{CS.8})$$

bounds on positions

$$l_i \leq x_i \leq u_i, \quad i = 1, 2, \dots, I. \quad (\text{CS.9})$$