

Exercises

Portfolio Optimization: Theory and Application Chapter 14 – Robust Portfolios

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Exercise 14.1: Sensitivity of naive portfolios

- Choose a portfolio optimization formulation.
- Collect T observations of the returns of N assets, form the data matrix $\mathbf{X} \in \mathbb{R}^{T \times N}$, and estimate the mean vector and covariance matrix.
- Sample B times new data matrices $\mathbf{X}^{(b)} \in \mathbb{R}^{T \times N}$, $b = 1, \dots, B$, from a Gaussian distribution with the previous mean vector and covariance matrix as true parameters of the distribution.
- For each data sample, solve the portfolio optimization obtaining the solutions $\mathbf{w}^{(b)}$, $b = 1, \dots, B$.
- Compare the different portfolios $\mathbf{w}^{(b)}$:
 - Use barplots to visually compare the allocations of the different portfolios.
 - Plot a histogram of the objective value achieved by the different portfolios (evaluated under the original mean vector and covariance matrix used to draw the sampled data).

Exercise 14.2: Sensitivity of robust worst-case portfolios

Repeat Exercise 14.1 but using a robust worst-case version of the portfolio formulation. Compare its sensitivity with that of the naive portfolio.

Exercise 14.3: Sensitivity of resampled portfolios

Repeat Exercise 14.1 but using a bagged version of the portfolio formulation. Compare its sensitivity with that of the naive portfolio.

Exercise 14.4: Worst-case mean vector under an ellipsoidal uncertainty set

Consider the ellipsoidal uncertainty set for $\boldsymbol{\mu}$:

$$\mathcal{U}_{\boldsymbol{\mu}} = \left\{ \boldsymbol{\mu} = \hat{\boldsymbol{\mu}} + \kappa \mathbf{S}^{1/2} \mathbf{u} \mid \|\mathbf{u}\|_2 \leq 1 \right\},$$

where $\mathbf{S}^{1/2}$ is the symmetric square-root matrix of the shape \mathbf{S} and κ determines the size of the ellipsoid.

Derive the worst-case value of $\mathbf{w}^\top \boldsymbol{\mu}$.

Exercise 14.5: Worst-case mean vector under a box uncertainty region

Consider the box uncertainty region for $\boldsymbol{\mu}$:

$$\mathcal{U}_{\boldsymbol{\mu}} = \{ \boldsymbol{\mu} \mid -\boldsymbol{\delta} \leq \boldsymbol{\mu} - \hat{\boldsymbol{\mu}} \leq \boldsymbol{\delta} \},$$

where $\boldsymbol{\delta}$ is the half-width of the box in all dimensions.

Derive the worst-case value of $\mathbf{w}^\top \boldsymbol{\mu}$.

Exercise 14.6: Worst-case covariance matrix under a data spherical uncertainty region

Consider the spherical uncertainty region for the data matrix $\mathbf{X} \in \mathbb{R}^{T \times N}$ (containing T observations of the N assets),

$$\mathcal{U}_{\mathbf{X}} = \left\{ \mathbf{X} \mid \|\mathbf{X} - \hat{\mathbf{X}}\|_F \leq \epsilon \right\}.$$

Derive the worst-case value of $\mathbf{w}^\top \boldsymbol{\Sigma} \mathbf{w}$ under a sample covariance estimation $\hat{\boldsymbol{\Sigma}} = \frac{1}{T} \hat{\mathbf{X}}^\top \hat{\mathbf{X}}$.

Exercise 14.7: Robust worst-case mean–variance portfolio

- Formulate a minimax version of the robust worst-case mean–variance portfolio for some choice of uncertainty sets for the mean vector and covariance matrix.
- Rewrite the problem in convex form by a brute-force sampling of the uncertainty sets and solve with a solver.
- Rewrite the problem in convex form by properly dealing with the worst-case mean vector and covariance matrix (either deriving the closed form or via Lagrange duality), and solve with a solver.
- Compare both solutions (try a different number of samples in the brute-force sampling approach).

Exercise 14.8: Convexity of robust mean–variance portfolio under ellipsoidal covariance matrix

- Formulate in minimax form the robust mean–variance portfolio with robustness in the variance under an ellipsoidal uncertainty set for the covariance matrix.
- Using the Lagrange dual problem version of the worst-case covariance matrix, rewrite the robust mean–variance portfolio as a regular (not minimax) optimization problem.

- c. Is this optimization problem convex? If not, can you rewrite it in convex form as a semidefinite program?

Exercise 14.9: Robust worst-case maximum Sharpe ratio portfolio

Write down the following portfolio formulations:

- a. Naive formulation of the maximum Sharpe ratio portfolio in convex form.
- b. Robust worst-case formulation of the maximum Sharpe ratio portfolio under general uncertainty sets for the mean vector and covariance matrix.
- c. Choose some specific uncertainty regions and rewrite the robust worst-case formulation of the maximum Sharpe ratio in convex form.

Exercise 14.10: Performance of resampled portfolios

- a. Choose a portfolio optimization formulation.
- b. Perform a backtest of
 - the naive portfolio
 - the bagged portfolio
 - the subset resampled portfolio
 - the subset bagged portfolio.
- c. Compare the performance and the computational cost.