Optimization technique in Portfolio creation

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1 Short Selling

2 Portfolio optimization



Short Selling $_{\circ \circ \circ}$

Portfolio optimization

Content

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2 Portfolio optimization

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An illustrative example with a long and a short position

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Graphical representation of the portfolio's effective frontier when applying short selling ...

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$$\sigma_p^2 = \sum_{i=1}^n \sum_{j=1}^n w_i * w_j * \sigma_{i,j}$$

Finding a portfolio with minimal risk

Two approaches can be applied to minimize the risk of portfolio:

1 Finding the absolute/Global minimum risk

Finding a portfolio with minimal risk

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- Finding the absolute/Global minimum risk
- 2 Finding the minimum risk at the desired return

Short Selling

 $\begin{array}{c} \text{Portfolio optimization} \\ \text{000} \bullet \text{000} \end{array}$

Minimum variance portfolio

• Solving the optimization problem in order to obtain weights



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The constraint:

$$\sum_{i=1}^{n} w_i = 1$$

Short Selling

Portfolio optimization 0000000

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$$rac{\partial L(ec{y})}{\partial y_i}$$
 for $i = 1, 2, \dots, n$

$$L(\vec{y}) = \sigma_{\rho}^2(\vec{w}) + \lambda_1(\sum_{i=1}^n w_i - 1)$$

• Applying partial derivation with respect to individual variables:

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- Apply matrix calculation ...

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Lagrangian function:

$$L(\vec{y}) = \sigma_{p}^{2}(\vec{w}) + \lambda_{1}(\sum_{i=1}^{n} w_{i} - 1) + \lambda_{2}(\sum_{i=1}^{n} w_{i} * \bar{r}_{i} - E(R_{p}))$$