

**Linear Programming:
Parametric Simplex Method
and
The Efficient Frontier**

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1 Recall the Ingredients for Portfolio Optimization

Raw Data:

$R_j(t)$ = return on asset j
in time period t \implies

Derived Data:

$$r_j = \frac{1}{T} \sum_{t=1}^T R_j(t)$$
$$D_{tj} = R_j(t) - r_j.$$

Decision Variables:

x_j = fraction of portfolio
to invest in asset j

Decision Criteria:

$$r(x) = \sum_j r_j x_j$$
$$\text{risk}(x) = \frac{1}{T} \sum_{t=1}^T \left| \sum_j D_{tj} x_j \right|$$

2 Optimization Problem

Set a value for **risk aversion** parameter μ and maximize a combination of reward and negative-risk:

$$\begin{aligned} \text{maximize} \quad & \mu \sum_j r_j x_j - \frac{1}{T} \sum_{t=1}^T \left| \sum_j D_{tj} x_j \right| \\ \text{subject to} \quad & \sum_j x_j = 1 \\ & x_j \geq 0 \quad \text{for all investments } j \end{aligned}$$

Because of absolute values not a linear programming problem.

Easy to convert...

3 A Linear Programming Formulation

$$\text{maximize} \quad \mu \sum_j r_j x_j - \frac{1}{T} \sum_{t=1}^T y_t$$

$$\text{subject to} \quad -y_t \leq \sum_j D_{tj} x_j \leq y_t \quad \text{for all times } t$$

$$\sum_j x_j = 1$$

$$x_j \geq 0 \quad \text{for all investments } j$$

$$y_t \geq 0 \quad \text{for all times } t$$

4 Adding Slack Variables w_t^+ and w_t^-

$$\text{maximize } \mu \sum_j r_j x_j - \frac{1}{T} \sum_{t=1}^T y_t$$

$$\text{subject to } -y_t - \sum_j D_{tj} x_j + w_t^- = 0 \quad \text{for all times } t$$

$$-y_t + \sum_j D_{tj} x_j + w_t^+ = 0 \quad \text{for all times } t$$

$$\sum_j x_j = 1$$

$$x_j \geq 0 \quad \text{for all investments } j$$

$$y_t, w_t^-, w_t^+ \geq 0 \quad \text{for all times } t$$

5 The Solution for Large μ

Varying the risk bound $0 \leq \mu < \infty$ produces the **efficient frontier**.

Large values of μ favor reward whereas small values favor minimizing risk.

Beyond some finite threshold value for μ , the optimal solution will be a portfolio consisting of just one asset—the asset j^* with the largest average return:

$$r_{j^*} \geq r_j \quad \text{for all } j.$$

It's easy to identify basic vs. nonbasic variables:

- Variable x_{j^*} is basic whereas the remaining x_j 's are nonbasic.
- All of the y_t 's are basic.
- If $D_{tj^*} > 0$, then w_t^- is basic and w_t^+ is nonbasic. Otherwise, it is switched.

6 The Optimal Dictionary for Large μ

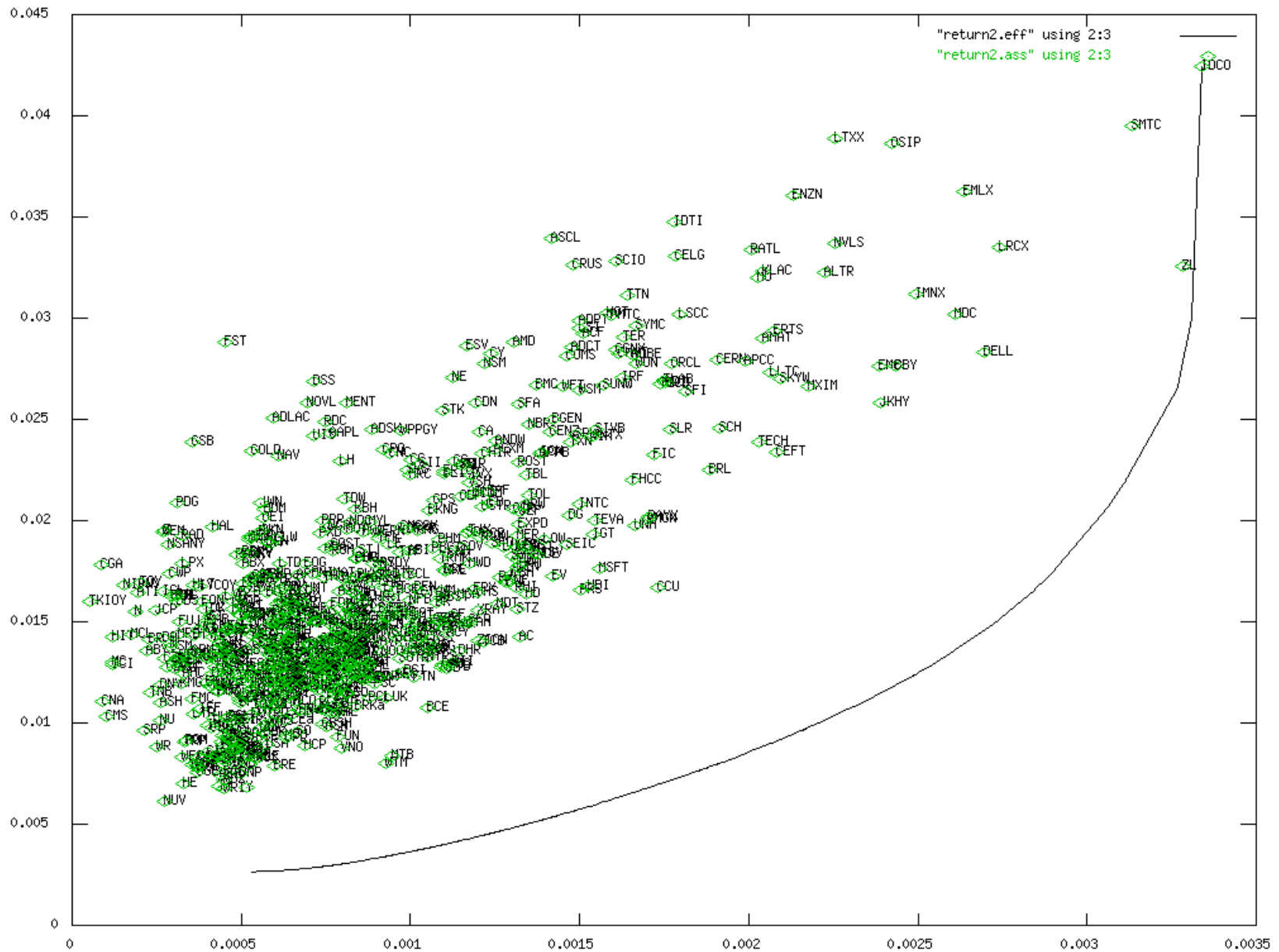
Let

$$T^+ = \{t : D_{tj^*} > 0\}, \quad T^- = \{t : D_{tj^*} < 0\}, \quad \text{and} \quad \epsilon_t = \begin{cases} 1, & \text{for } t \in T^+ \\ -1, & \text{for } t \in T^- \end{cases}$$

It's tedious, but here's the optimal dictionary:

$$\begin{aligned} \zeta &= \frac{1}{T} \sum_{t=1}^T \epsilon_t D_{tj^*} & - \frac{1}{T} \sum_{j \neq j^*} \sum_{t=1}^T \epsilon_t (D_{tj} - D_{tj^*}) x_j & - \frac{1}{T} \sum_{t \in T^-} w_t^- & - \frac{1}{T} \sum_{t \in T^+} w_t^+ \\ &+ \mu r_{j^*} & + \mu \sum_{j \neq j^*} (r_j - r_{j^*}) x_j & & \\ \hline y_t &= -D_{tj^*} & - \sum_{j \neq j^*} (D_{tj} - D_{tj^*}) x_j & + w_t^- & t \in T^- \\ w_t^- &= 2D_{tj^*} & + 2 \sum_{j \neq j^*} (D_{tj} - D_{tj^*}) x_j & + w_t^+ & t \in T^+ \\ y_t &= D_{tj^*} & + \sum_{j \neq j^*} (D_{tj} - D_{tj^*}) x_j & + w_t^+ & t \in T^+ \\ w_t^+ &= -2D_{tj^*} & - 2 \sum_{j \neq j^*} (D_{tj} - D_{tj^*}) x_j & + w_t^- & t \in T^- \\ x_{j^*} &= 1 & - \sum_{j \neq j^*} x_j & & \end{aligned}$$

7 Efficient Frontier



Click [here](#) for an expanded browser view.

8 Computing the Efficient Frontier

Using a reasonably efficient code for the parametric self-dual simplex method (simpo), it took **22,000** pivots and **1.5 hours** to solve for **one point** on the efficient frontier.

Customizing this same code to solve parametrically for every point on the efficient frontier, it took **20,500** pivots and **57 minutes** to compute **every point** on the frontier.

The efficient frontier consists of **1308** distinct portfolios. Click **here** for a complete list (**warning: the file is 2.5 MBytes**).