

Robust Optimization:

What Works and What Does Not

Dong Shaw, Ph.D.

Paradigm Asset Management, LLC.

for

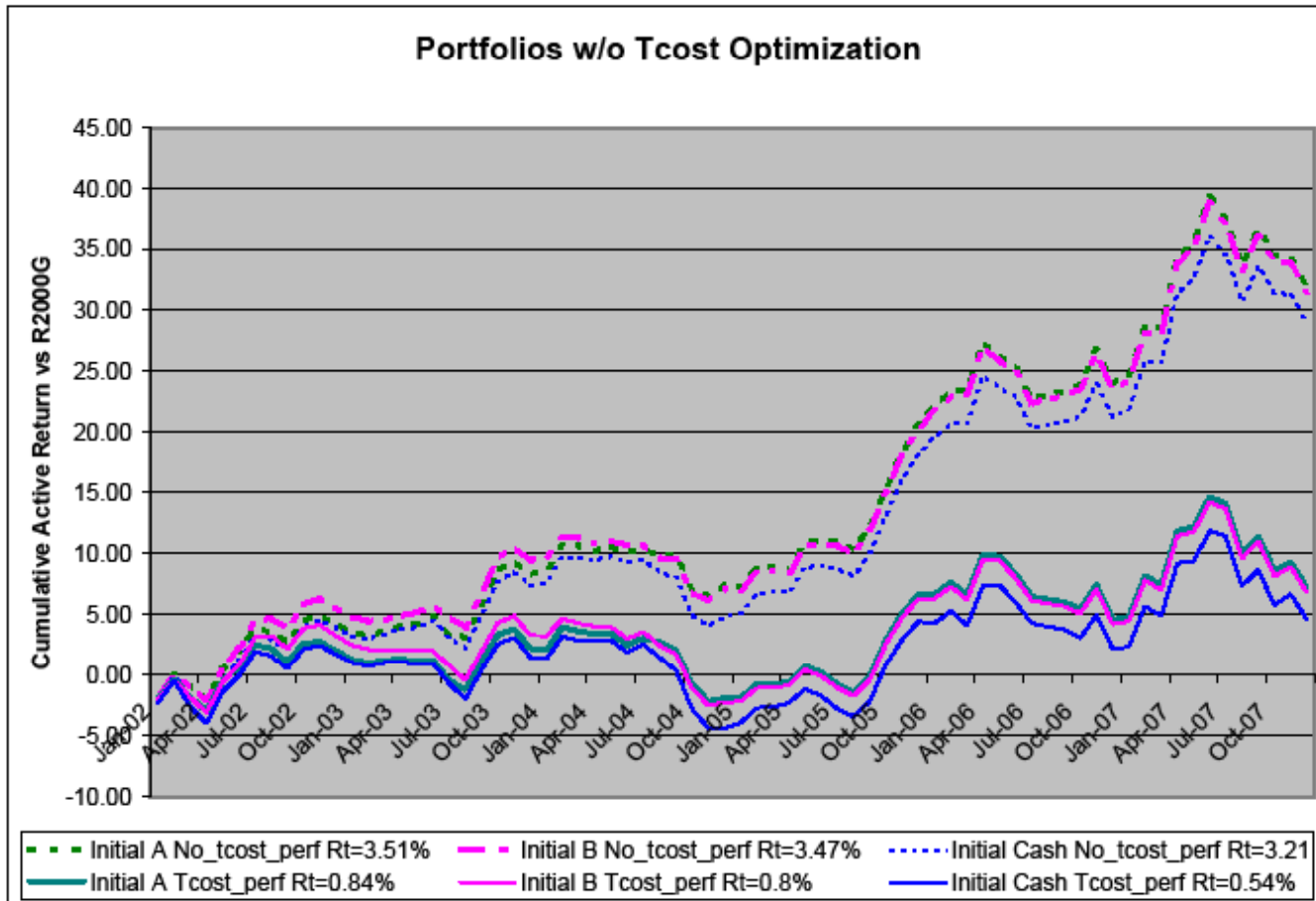
Northfield's Research Conference 2008

Overview

- **Why Optimized Portfolios Are Not Robust?**
 - **Ex-Post Performance and Multi-period Backtests**
- **Robust Optimization with SOCP (Second Order Cone Programming)**
 - **Equivalence to the Quadratic Penalty**
- **Two-Stage Optimization**
 - **Make a Portfolio Optimization Process (POP) Robust**
- **“Less is More”**
 - **One-period Static Optimization might be Over-Analyzed**
- **Closing Remarks**

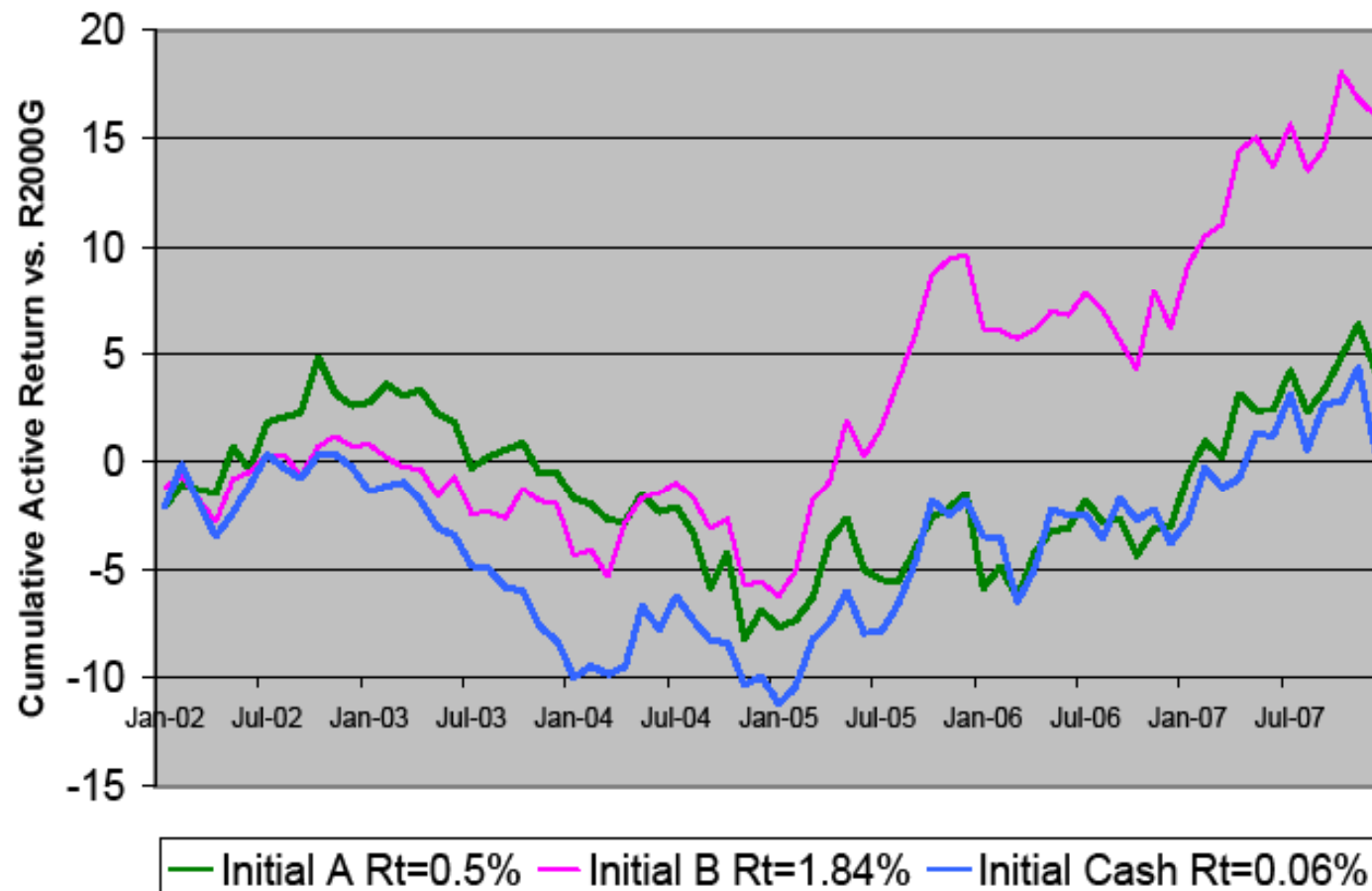
An Example

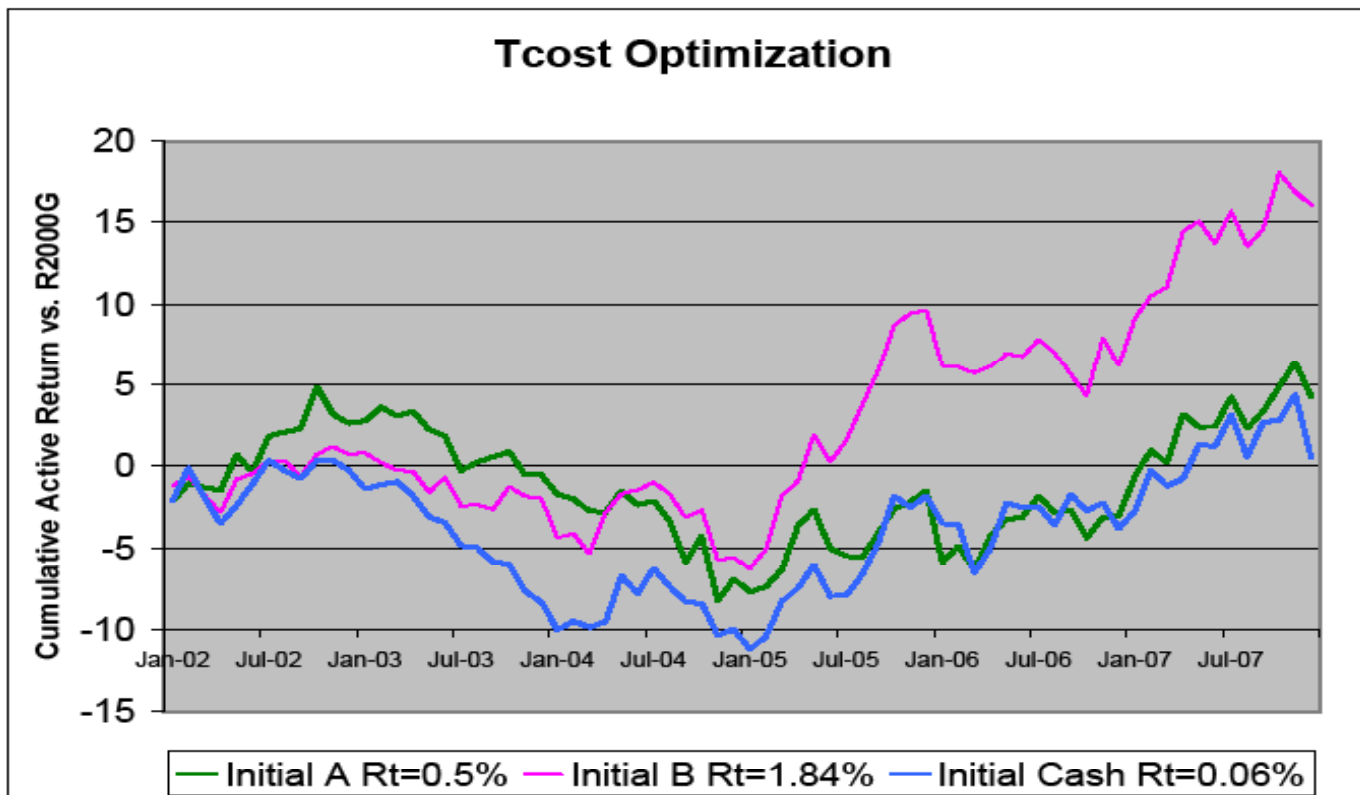
- Long-Only Active Portfolio Benchmarked to Russell 2000 Growth
- Max Alpha (Predicted)
- Tracking Error no more than 4.75%
- Security level active-weight bounds
- Beta neutral to the benchmark
- Tcost (25 bps impact + 3¢/share)
- No more than 200 securities in the portfolio



Annualized Performance	Initial Portfolio		
	A	B	C
Active Returns	3.67%	3.71%	3.41%
Active Returns (after Tcost)	0.80%	0.84%	0.54%
Active Risk	4.62%	4.59%	4.74%
Tcost Impact	-2.89%	-2.89%	-2.89%

Tcost Optimization





Annualized Performance	Initial Portfolio		
	A	B	C
Active Returns	0.50%	1.84%	0.06%
Active Risk	4.75%	4.81%	4.51%
Tcost Impact	-0.20%	-0.21%	-0.26%

Does Robust Optimization help?

- S. Ceria and R. Stubbs. (2006). “Incorporating Estimation Errors into Portfolio Selection: Robust Portfolio Construction.” *Journal of Asset Management* 7. pp. 109-127.
- J.H. Lee, D. Stefek and A. Zheleznyak (2006). Robust Portfolio Optimization: A Closer Look. *Barra Research Reports*.
- F. Fabozzi, P. Kolm, D. Pachamanova and S. Focardi. (2007). Robust Portfolio Optimization. *The Journal of Portfolio Management*.
- Michaud, R. O. (1989). The Markowitz Optimization Enigma: Is “Optimized” Optimal. *The Financial Analyst Journal* V. 45. pp.31-42.

What is Robust MVO?

A mean-variance optimization (MVO) problem:

α is the expected absolute or excess return of stocks from a universe

$$\begin{aligned} \max \quad & \alpha'w \\ & w'Vw \leq \sigma^2 \\ & e'w = 1 \\ & w \geq 0. \end{aligned} \tag{1}$$

Its active counterpart can be written as:

Here, α is the expected active return of stocks from a universe

$$\begin{aligned} \max \quad & \alpha'w \\ & (w - w^b)'V(w - w^b) \leq \tau^2 \\ & e'w = 1 \\ & w \geq 0. \end{aligned} \tag{2}$$

A very important and well-known property for the active problem is:

$$\alpha'w^b = 0.$$

A benchmark portfolio's alpha is zero! (See Grinold and Kahn (1999)).

What is Robust MVO?

Due to the fact that the true α is not known, and we can only obtain an estimation of the “true” α . Assume

$$\alpha \sim N(\alpha^*, \Omega)$$

a normal distribution of **known** covariance.

For any small probability p , we can construct a $(1 - p)$ confidence region for some κ_p as

$$\mathcal{B} = \{\alpha | (\alpha - \hat{\alpha})' \Omega^{-1} (\alpha - \hat{\alpha}) \leq (\kappa_p)^2\},$$

where $\hat{\alpha}$ is the estimated expected return of the stocks. The robust optimization tries to optimize the “worst case scenario” out of all possibilities, or over some most likely cases.

$$\begin{aligned} \max \min_{\{\alpha \in \mathcal{B}\}} \quad & \alpha' w \\ & w' V w \leq \tau^2 \\ & e' w = 1 \\ & w \geq 0. \end{aligned} \tag{3}$$

What is Robust MVO?

Absolute Return-Risk Robust MVO Formulation

The robust optimization problem can be formulated as,

$$\begin{aligned} \max \quad & \hat{\alpha}w - \kappa_p \sqrt{w' \Omega w} \\ & w' V w \leq \sigma^2 \\ & e' w = 1 \\ & w \geq 0. \end{aligned} \quad (4)$$

This is a second-order cone programming (SOCP) problem instead of a quadratic programming problem (QP) largely due to the appearance of the square root.

Because

$$\begin{aligned} & \min_{\alpha} \{ \alpha' w \mid (\alpha - \hat{\alpha})' \Omega^{-1} (\alpha - \hat{\alpha}) \leq (\kappa_p)^2 \} \\ &= \hat{\alpha}w + \kappa_p \min_x \{ x' u \mid x' x \leq 1 \} \\ &= \hat{\alpha}w - \kappa_p \|u\| \\ &= \hat{\alpha}w - \kappa_p \sqrt{w' \Omega w}. \end{aligned} \quad (5)$$

$$\max_{\alpha \in \mathcal{B}} \min_{\alpha \in \mathcal{B}} \alpha' w = \max \hat{\alpha}w - \kappa_p \sqrt{w' \Omega w}.$$

Our Results

Demystification of the Robust Optimization

$$\begin{aligned} \max \quad & \hat{\alpha}w - \kappa_p \sqrt{w' \Omega w} \\ & w' V w \leq \sigma^2 \\ & e' w = 1 \\ & w \geq 0. \end{aligned} \tag{6}$$

There exists a λ such that the solution of the following problem has an identical solution to the above Robust MVO,

$$\begin{aligned} L(\lambda) = \max \quad & \hat{\alpha}w - \lambda w' \Omega w \\ & w' V w \leq \sigma^2 \\ & e' w = 1 \\ & w \geq 0. \end{aligned} \tag{7}$$

This is a Quadratic Penalty Function problem.

Our Results

Why Our Claim is true?

$$\begin{aligned} \max \quad & \hat{\alpha}'w \\ & w'Vw \leq \sigma^2 \\ & e'w = 1 \\ & w \geq 0. \end{aligned}$$

It is well-known that the regular MVO can be solved by

$$\begin{aligned} \max \quad & \hat{\alpha}'w - \mu w'Vw \\ & e'w = 1 \\ & w \geq 0, \end{aligned}$$

for some μ . (μ is formally called the Lagrangian Multiplier.)

$$\begin{aligned} \max \quad & \hat{\alpha}w - \kappa_p \sqrt{w'\Omega w} \\ & w'Vw \leq \sigma^2 \\ & e'w = 1 \\ & w \geq 0. \end{aligned}$$

By the same logic, the above Robust Problem can be solved through

$$\begin{aligned} \max \quad & \hat{\alpha}w \\ & \sqrt{w'\Omega w} \leq \theta \quad \text{or} \quad w'\Omega w \leq \theta^2 \\ & w'Vw \leq \sigma^2 \\ & e'w = 1 \\ & w \geq 0. \end{aligned}$$

Our Results

The Implication of Our Claim

Since V is the risk model, α^* is the expected absolute returns,

$$\alpha \sim N(\alpha^*, V).$$

On account of Axioma's Robust model's assumption $\alpha \sim N(\alpha^*, \Omega)$, we shall conclude that

$$\Omega = \delta V.$$

If this is the case, the Efficient Frontier from the Robust Optimization would be identical to the one from Regular MVO.

$$L(\lambda) = \max_{\substack{\hat{\alpha}w - \lambda\delta w'Vw \\ w'Vw \leq \sigma^2 \\ e'w = 1 \\ w \geq 0.}} \quad (8)$$

A Big Question

Does Robust MVO Enhance Portfolio Alpha?

Just consider a simpler case where $\Omega = \text{diag}(\sigma_i^2)$.

$$\begin{aligned} \max \quad & \hat{\alpha}w - \lambda \{w'\Omega w\} \\ & w'Vw \leq \sigma^2 \\ & e'w = 1 \\ & w \geq 0. \end{aligned} \tag{9}$$

$$\begin{aligned} \max \quad & \hat{\alpha}w - \mu w'(V + (\lambda/\mu)\Omega)w \\ & e'w = 1 \\ & w \geq 0. \end{aligned} \tag{10}$$

The Robust Optimization suggests use a different risk model that increases the idiosyncratic risks for stocks of which the alphas might be harder to predict.

A Mistake in the Current Robust MVO Formulation

Active Returns/Risks Robust Reformulation

For active returns MVO, the corresponding robust problem is

$$\begin{aligned} \max \min_{\{\alpha \in \mathcal{B}\}} \quad & \alpha'w \\ & (w - w^b)'V(w - w^b) \leq \tau^2 \\ & e'w = 1 \\ & w \geq 0. \end{aligned} \quad (11)$$

Ceria and Stubbs' (2005) from Axioma claimed that the above objective function is equivalent to

$$\max \hat{\alpha}(w - w^b) - \kappa_p \sqrt{(w - w^b)' \Omega (w - w^b)}$$

Actually, due to the fact that $\alpha'w^b = 0$, their objective is not entirely accurate. We show that a correct form of objective function should be

A Mistake in the Current Robust MVO Formulation

$$\max \quad \hat{\alpha}w - \kappa_p \sqrt{w' \Omega w - \frac{(w' \Omega w^b)^2}{w^b' \Omega w^b}},$$

or

$$\begin{aligned} \max \quad & \hat{\alpha}w - \lambda \left(w' \Omega w - \frac{(w' \Omega w^b)^2}{w^b' \Omega w^b} \right) \\ & (w - w^b)' V (w - w^b) \leq \sigma^2 \\ & e' w = 1 \\ & w \geq 0. \end{aligned} \tag{12}$$

What is the Remedy?

- **Robust MVO does not differ much from regular MVO, it is equivalent to the quadratic penalty function method;**
- **Robust MVO requires to estimate another set of parameters;**
- **Robust Optimization is a one-period static solution.**

Two-Stage Optimization --- A Solution

- Stage 1: Find a path-independent “ideal” portfolio

- Max Alpha

1a. The Tracking Error upper bound;

1b. Linear side-constraints; (Such as security-level bounds, factor-bets);

1c. Upper bound on number of securities (optional).

- Stage 2: Find the tradable portfolio

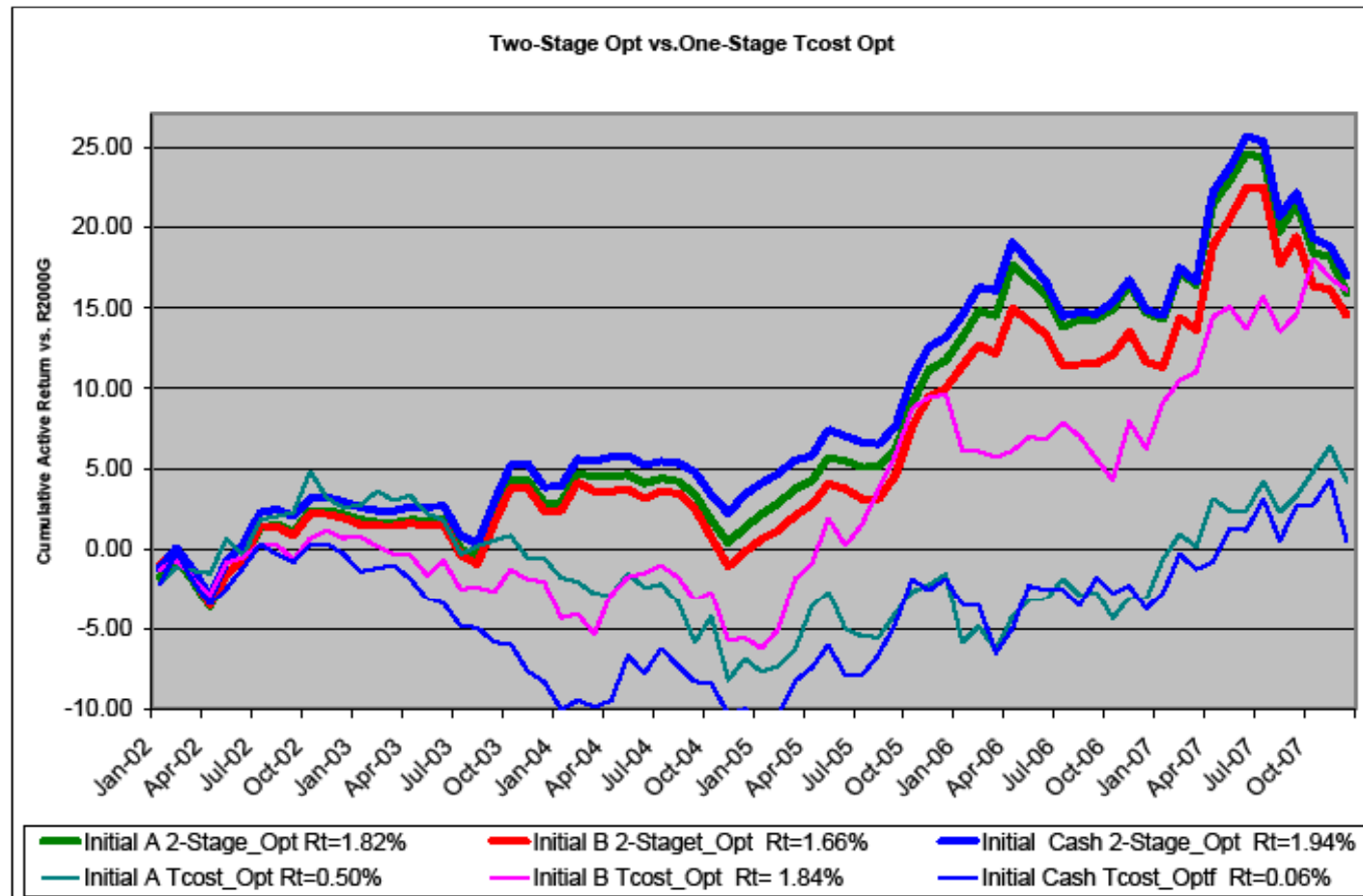
Max Utility Function

same constraints as Stage 1, 1c) shall be included.

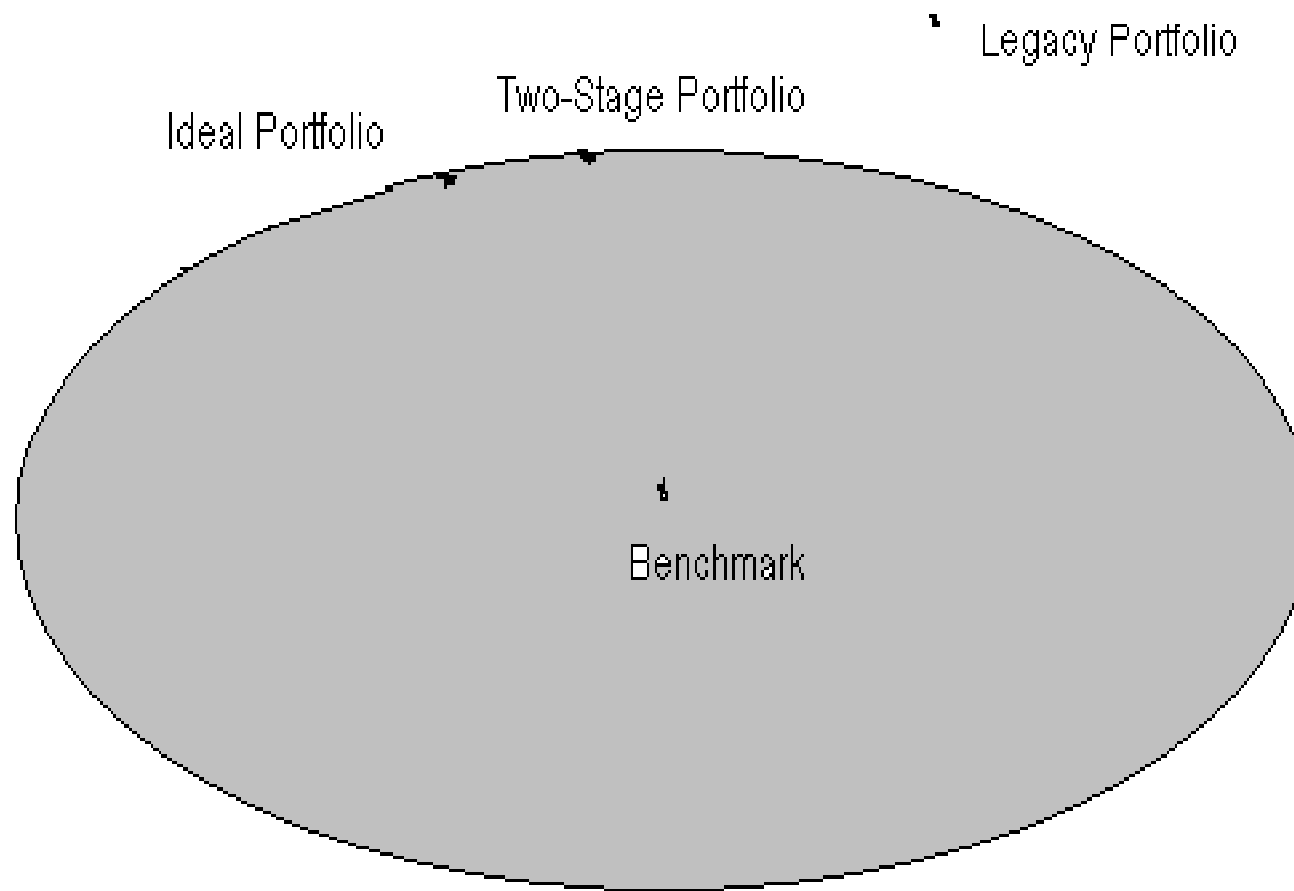
Utility Function = $\alpha - \lambda (\text{tracking_err vs. } \underline{\text{ideal portfolio}})^2$

- μ (Tcost against the legacy portfolio)

Two-Stage vs. One-Stage



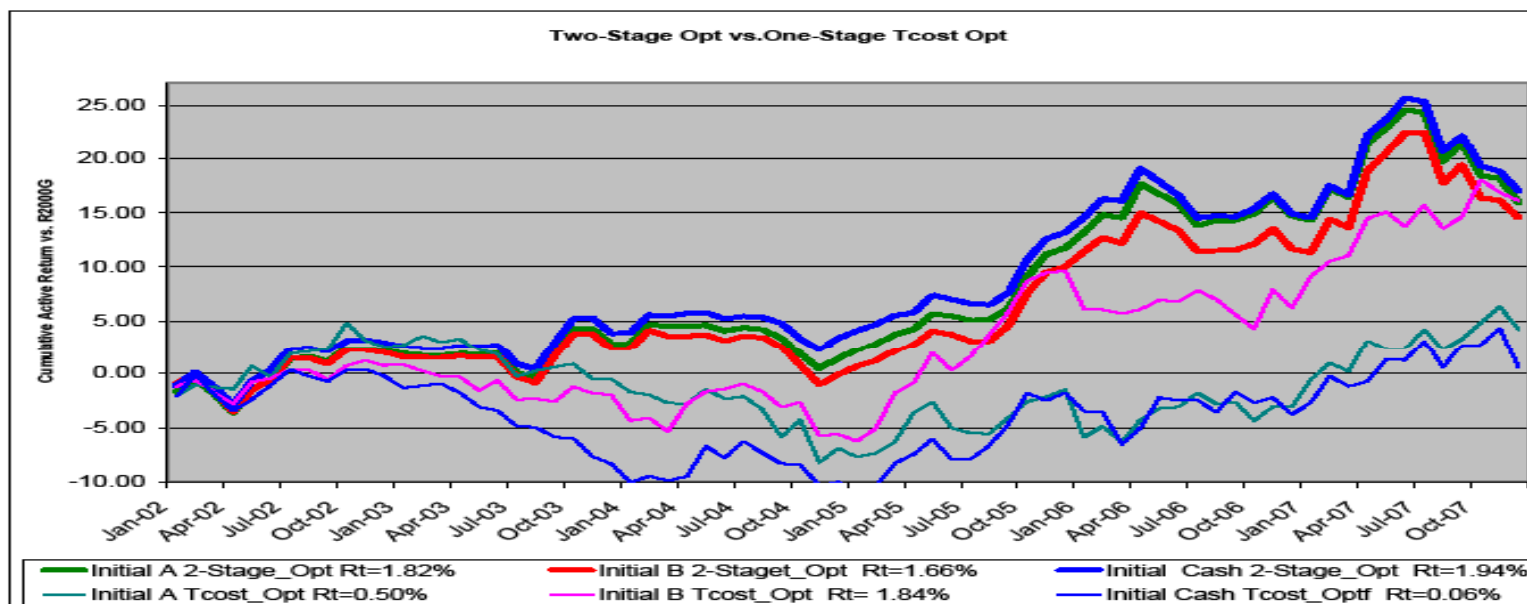
Two-Stage Optimization



Two-Stage Optimization --- A Solution

- tradable portfolio
=Function (alpha, risk model, bounds, legacy portfolio, λ , μ)
- *How to decide λ and μ is an art.*

Two-Stage Optimization



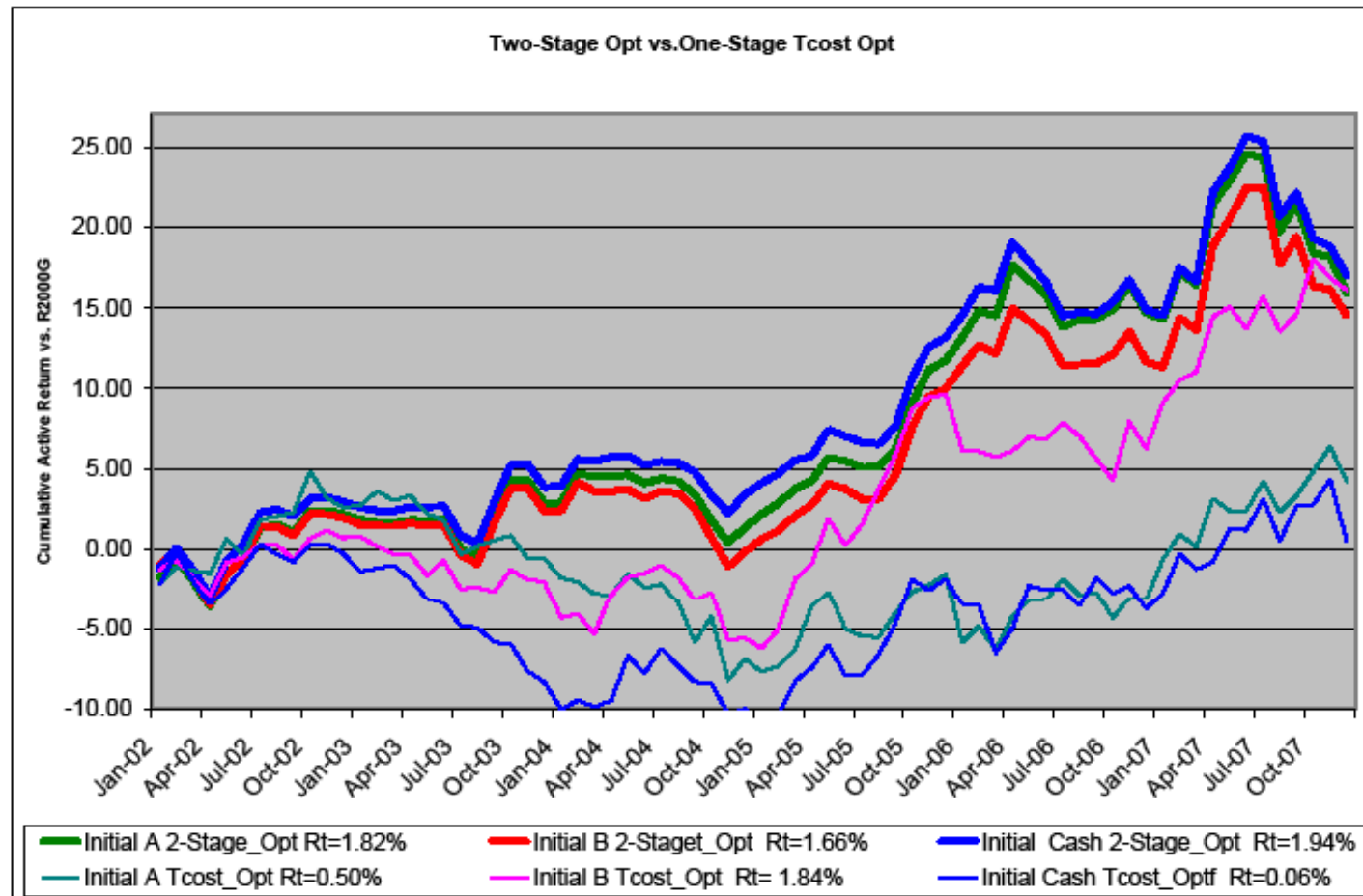
	Initial Portfolio		
	A	B	C
Active Returns	1.82%	1.66%	1.94%
Active Risk	4.20%	4.28%	4.20%
Tcost Impact	-1.43%	-1.44%	-1.42%

Two-Stage Optimization

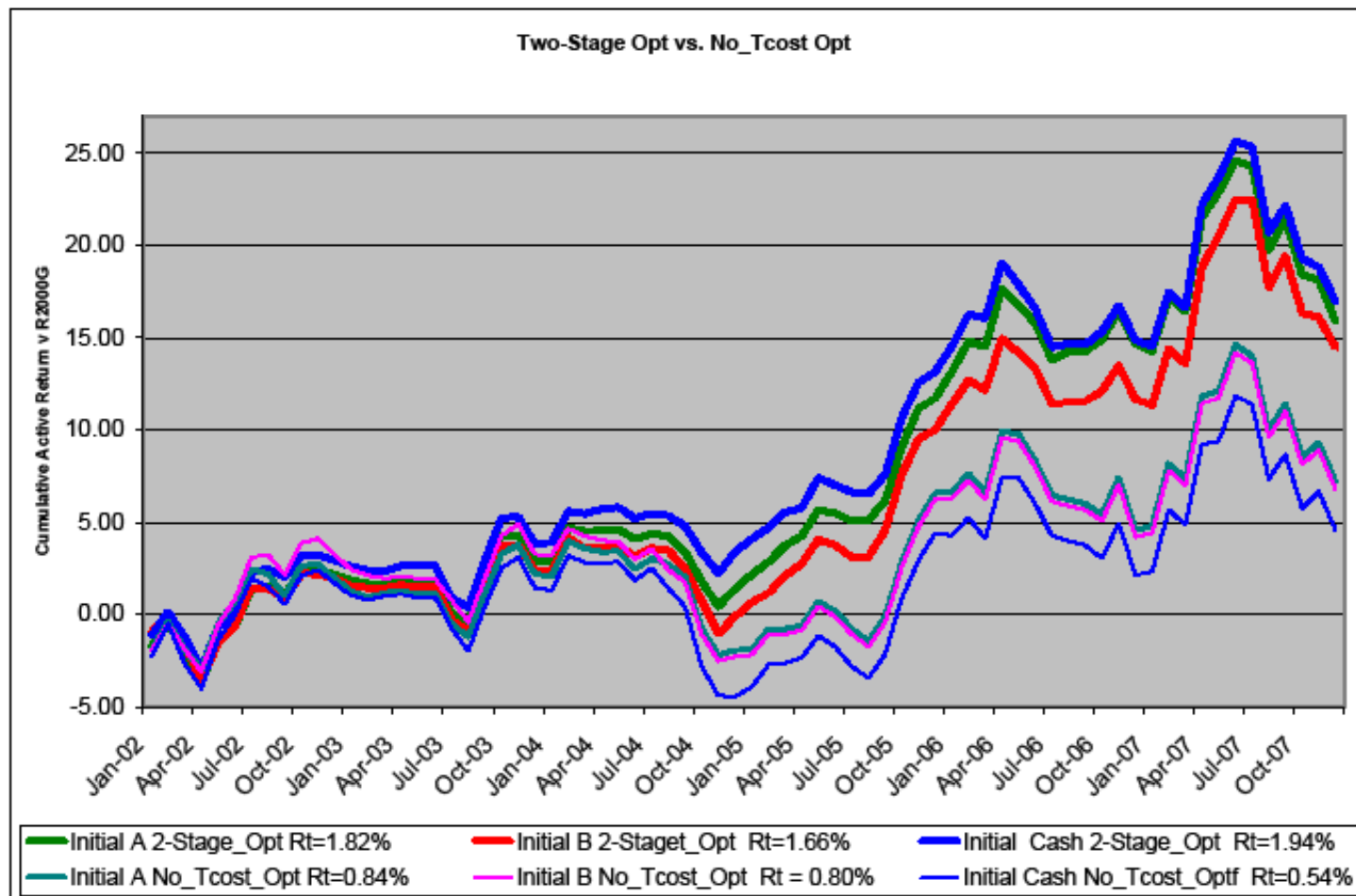
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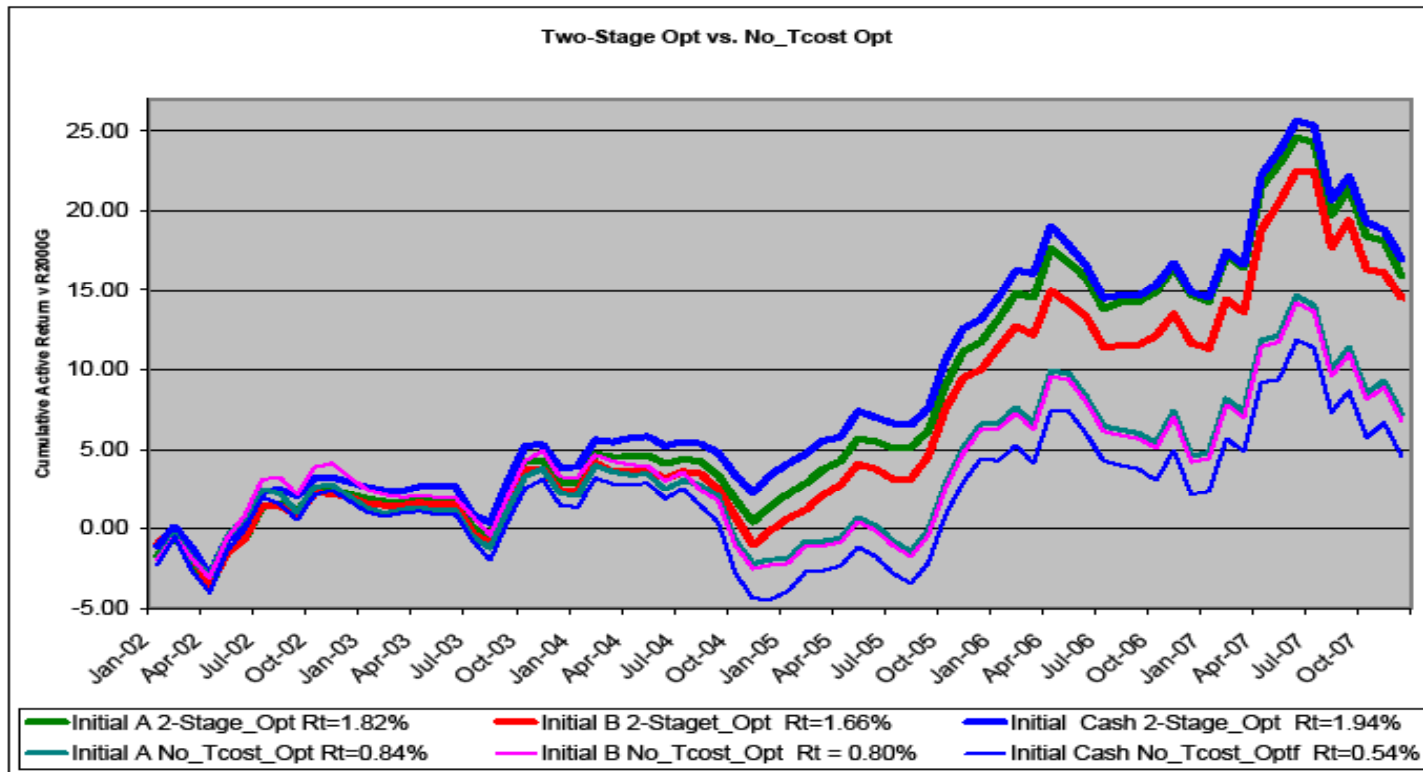
One-Stage Tcost Optimization

Two-Stage vs. One-Stage



Two-Stage vs. No-Tcost Optimization





Annualized Performance	Initial Portfolio		
	A	B	C
Active Returns (Two-Stage)	1.82%	1.66%	1.94%
Active Returns (No-Tcost Opt)	0.80%	0.84%	0.54%
Active Risk	4.57%	4.55%	4.69%
Tcost Impact (Two-Stage)	-1.43%	-1.44%	-1.42%
Tcost Impact (No-Tcost Opt)	-2.89%	-2.89%	-2.89%

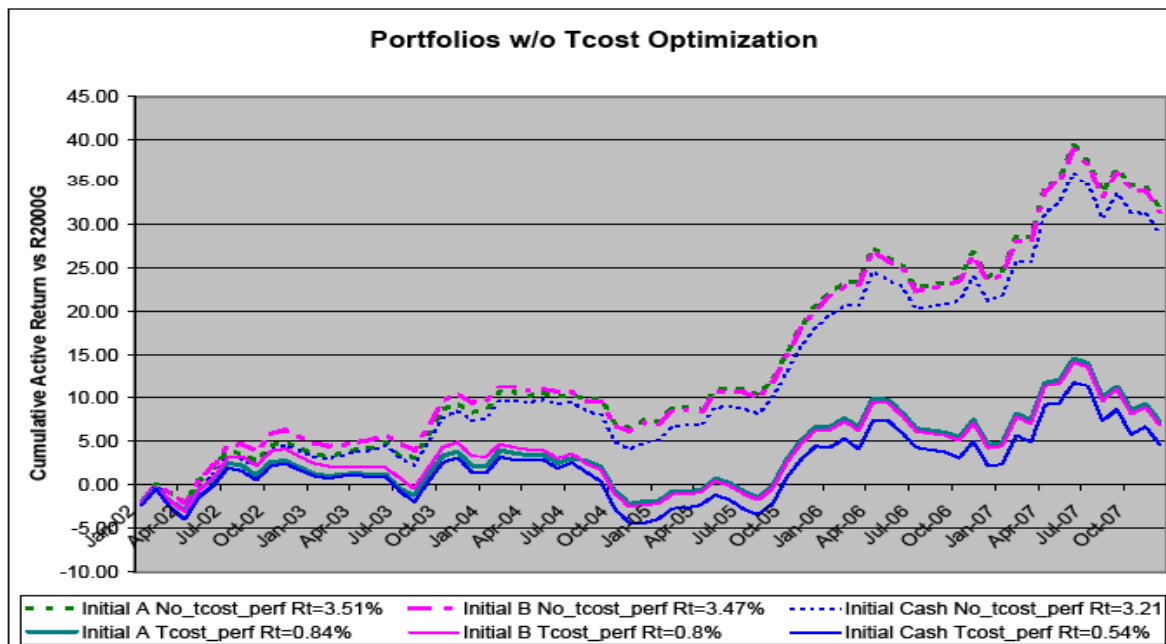
Less is More

- It might *not* be a good idea to *over-analyze* the one-period static optimization problems
- How to make a Portfolio Optimization Process (POP) Robust?
 - 1) form a maximization problem that is *concave* if possible;
having a *unique* optimal solution is even better (When the risk model is a positive-definite matrix *and* risk budget is binding).
 - 2) Constraints that limit number of securities in a portfolio may cause the optimization problem *non-concave*, risk-targeting is a *bad idea*.
 - 3) Find and understand a series of *path-independent* portfolios. The PMs don't need to trade on these portfolios, but they are better to *know* these portfolios.
 - 4) Including transaction cost and market impact into the POP, or adding turnover constraints may create a series of *path-dependent* portfolios.

Less is More

- The Portfolio Optimization Process (POP) uses ex-ante data to achieve an ex-post goal, and it is a single-period proxy to a multi-period stochastic problems.
- The MVO pioneered by Markowitz was developed originally to trade-off between risks and returns
- Currently Optimizer is used as a portfolio construction tool, sometimes, the tool to create a *tradable portfolio*.

Recommendations



Run a Two-Stage Optimization (It requires an optimizer that can handles the 2nd Benchmark.)

If not, run an optimization ignoring Tcost/Turnover to ensure path-independent.

G. Sofianos, S. Takriti and I. Tierens. (2007) Including Trading Costs in Portfolio Optimization. Equity Execution Strategies, Goldman Sachs.

Closing Remarks

- **Robust MVO is equivalent to the Quadratic Penalty Function Approach**
- **“Less is More”**
- **Two-Stage Optimization Enhances Robustness**