A Comparison of Universal and Mean-Variance Efficient Portfolios

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Universal and MVE Portfolios

- Investment Dynamics
- Universal Portfolios
- Mean-Variance Efficient (MVE) Portfolios
- Examples
- Conclusions

Investment Problem

Fixed parameters:

- \blacksquare M = number of assets
- ightharpoonup N = number of periods to invest

Market determines:

• $p_{n,m}$ = price of asset m at beginning of period n

You choose:

• $a_{n,m}$ = number of asset m owned at beginning of per. n

To control:

• h_n = wealth at beginning of per. n

Wealth

Assume:

- No new wealth added to portfolio
- Prices at the end of a period are equal those at beginning of next

Wealth at beginning of period n:

$$h_n = a_{n,1}p_{n,1} + \dots + a_{n,M}p_{n,M}$$

Wealth at beginning of period n + 1:

$$h_{n+1} = a_{n,1}p_{n+1,1} + \dots + a_{n,M}p_{n+1,M}$$

Period Returns

Return for period n:

$$R_{n} = h_{n+1}/h_{n}$$

$$= \frac{a_{n,1}p_{n,1}}{h_{n}} \frac{p_{n+1,1}}{p_{n,1}} + \dots + \frac{a_{n,M}p_{n,M}}{h_{n}} \frac{p_{n+1,M}}{p_{n,M}}$$

$$= b_{n,1}X_{n,1} + \dots + b_{n,M}X_{n,M}$$

$$= b_{n}^{t}X_{n}$$

where

$$b_{n,m} = \frac{a_{n,m}p_{n,m}}{h_n}, \quad b_n = (b_{n,1} \dots b_{n,M})^t$$

 $X_{n,m} = \frac{p_{n+1,m}}{p_{n,m}}, \quad X_n = (X_{n,1} \dots X_{n,M})^t$

Cumulative Returns

Return from period 1 to N:

$$S_{N} = \frac{h_{N+1}}{h_{1}}$$

$$= \left(\frac{h_{N+1}}{h_{N}}\right) \left(\frac{h_{N}}{h_{N-1}}\right) \cdots \left(\frac{h_{2}}{h_{1}}\right)$$

$$= \prod_{n=1}^{N} R_{n}$$

$$= \prod_{n=1}^{N} b_{n}^{t} X_{n}$$

Doubling Rate

Assuming that $b_n^t X_n > 0$ for n = 1, ..., N:

$$S_N = \exp\left(\sum_{n=1}^N \log b_n^t X_n\right)$$
$$= \exp\left(NW_N\right)$$

where

$$W_N = \frac{1}{N} \sum_{n=1}^{N} \log b_n^t X_n$$

is called the "doubling" rate of the portfolio.

Constant Rebalanced Portfolios

Constant Rebalanced Portfolio (CRP):

$$b_1 = b_2 = \dots = b_N = b$$

Buy-and-Hold (B&H) Portfolio:

$$b = e_m = (0, 0, \dots, 0, 1, 0, \dots, 0)^t$$

Best CRP:

$$b_N^* = \operatorname*{argmax}_{b \in B} S_N(b) = \operatorname*{argmax}_{b \in B} W_N(b)$$

where

$$B = \{b \in \mathcal{R}^M : b_m \ge 0, m = 1, 2, \dots, M; \sum_{m=1}^M b_m = 1\}$$

CRP Example [Blum97]

Two Stocks:

- Stock #1: $X_{n,1} = 1$, n = 1, ..., N
- Stock #2: $X_{n,2}=\left\{ egin{array}{ll} \frac{1}{2} & n=\mbox{even} \\ 2 & n=\mbox{odd} \end{array} \right.$

Portfolios:

- $S_N \leq 2$ for any B&H portfolio
- Return of $b = (\frac{1}{2}, \frac{1}{2})^t$ increases exponentially by 9/8 every two days:

$$R_n = \begin{cases} \left(\frac{1}{2}\right)(1) + \left(\frac{1}{2}\right)\left(\frac{1}{2}\right) = \frac{3}{4} & n = \text{even} \\ \left(\frac{1}{2}\right)(1) + \left(\frac{1}{2}\right)(2) = \frac{3}{2} & n = \text{odd} \end{cases}$$

Equally Weighted CRP [Wang00]

- Equally weighted CRP: $b = (\frac{1}{M}, \dots, \frac{1}{M})^t$
- Assume X_n are IID random vectors with $E[X_n] = \bar{X}$ and $V = E[(X_n E(X_n))(X_n E(X_n))^t]$:
 - Expected return: $E(R_n) = \frac{1}{M} \sum_{m=1}^{M} \bar{X}_m$
 - Variance of return:

$$\begin{aligned} \operatorname{var}(R_n) &= \frac{1}{M} \left(\frac{1}{M} \sum_{m=1}^{M} v_{m,m} \right) \\ &+ \left(\frac{M^2 - M}{M^2} \right) \left(\frac{1}{M^2 - M} \sum_{m=1}^{M} \sum_{k \neq m}^{M} v_{m,k} \right) \\ &= \left(\frac{1}{M} \right) (\operatorname{avg. var.}) + \left(\frac{M^2 - M}{M^2} \right) (\operatorname{avg. cov.}) \end{aligned}$$

Universal Portfolios

- Origin: T. Cover, "Universal Portfolios", Mathematical Finance, 1991
- Objective: Track return of best CRP chosen after asset returns revealed (b_N^*)
- Algorithm: Return-weighted average of all CRPs
- Performance:
 - Yields average return of all CRPs
 - Return exceeds geometric mean of asset returns
 - Asymptotically "tracks" best CRP return

Universal Portfolio Algorithm

Return-weighted average of all CRPs:

Initialize:
$$\hat{b}_1 = \left(\frac{1}{M}, \cdots, \frac{1}{M}\right)^t$$

For
$$n > 1$$
: $\hat{b}_n = \frac{\int_B b S_n(b) db}{\int_B S_n(b) db}$

where

$$B = \{ b \in \mathcal{R}^M : b_m \ge 0, m = 1, 2, \dots, M; \sum_{m=1}^M b_m = 1 \}$$

Performance

Regardless of the stock return sequence:

Portfolio return is the average of all CRPs' returns

$$\hat{S}_N = \prod_{n=1}^N \hat{b}_n^t X_n = \frac{\int_B S_n(b)db}{\int_B db}$$

Return is greater than geometric mean

$$\hat{S}_N \ge \left(\prod_{m=1}^M S_N(e_m)\right)^{1/M}$$

Asymptotic Performance

Under many assumptions:

$$\frac{\hat{S}_N}{S_N^*} \sim \left(\sqrt{\frac{2\pi}{N}}\right)^{M-1} \frac{(M-1)!}{|J^*|^{1/2}}$$

where

 $J^* = \text{Hessian of } W_N(b) \text{ at its maximum for large N}$ $S_N^* = \text{return of best CRP } b_N^*$

and \sim means the ratio of left- and right-hand sides equals one for large N

Assumptions for Asymptotic Perf.

The stock sequence X_1, X_2, \ldots , satisfies:

- $a \le X_{n,m} \le q, m = 1, ..., M, n = 1, 2, ...,$ for some $0 < a \le q < \infty$
- $J_N^* \to J^*$ for some positive definite matrix J^*
- $b_N^* \to b^*$ for some b^* in the interior of the B

and there exists a function W(b) such that

- $W_N(b) \nearrow W(b)^*$ for any portfolio $b \in B$
- W(b) is strictly concave
- ullet W(b) has bounded third partial derivatives
- W(b) achieves its maximum at b^* in the interior of the B

Outline of Proof

- Show LHS ≥ RHS:
 - Expand $W_N(b)$ in Taylor series about the best CRP b_N^* to three terms:
 - ullet First term is zero because gradient is zero at b_N^*
 - ullet Second term is related to J_N^*
 - Third term is bounded
 - Manipulate expression to look like a Gaussian CDF
 - Bound the Gaussian CDF
- Show LHS ≤ RHS:
 - Follows from Laplace's method of integration

Mean-Variance Efficient Portfolios

- Origins: Markowitz (1952) and Sharpe (1963) (1990 Nobel Laureates)
- Objective: To minimize the variance of portfolio returns, while maintaining a desired average return

Algorithm:

- A solution to a constrained optimization problem
- Requires estimates of asset return mean and covariance matrix

Performance:

- Tracks objective portfolio if asset returns are IID
- Otherwise, no guarantee

Objective of MVE Portfolio

Minimize:
$$b_n^t V_n b_n$$

Subject to: 1)
$$\sum_{m=1}^{M} b_{n,m} = 1$$

2)
$$b_n^t \bar{X}_n = \bar{R}_n$$

where

 \bar{X}_n = estimate of asset return mean

 V_n = estimate of asset return covariance matrix

 \bar{R}_n = desired average return

The MVE Portfolio

Mean and covariance estimators:

$$\bar{X}_n = \frac{1}{\Lambda} \sum_{k=1}^{n-1} \lambda^{n-k-1} X_k$$

$$V_n = \frac{1}{\Lambda} \sum_{k=1}^{n-1} \lambda^{n-k-1} (X_k - \bar{X}_n) (X_k - \bar{X}_n)^t$$

where $0 < \lambda \le 1$ and $\Lambda = \sum_{k=1}^{n-1} \lambda^{n-k-1}$

The MVE Portfolio:

$$b_n = w_0(1 - \bar{R}_n) + w_1\bar{R}_n$$

where w_0 and w_1 are portfolios that yield on average 0% and 100% returns, respectively

The 0% and 100% Avg. Ret. Ports.

$$w_0 = \frac{1}{D} \left(c_1 V_n^{-1} 1 - c_2 V_n^{-1} \bar{X}_n \right)$$

$$w_1 = \frac{1}{D} \left(c_1 V_n^{-1} 1 - c_2 V_n^{-1} \bar{X}_n \right) + \frac{1}{D} \left(c_3 V_n^{-1} \bar{X}_n - c_2 V_n^{-1} 1 \right)$$

where

$$1^{t} = (1, ..., 1)$$

$$c_{1} = \bar{X}_{n}^{t} V_{n}^{-1} \bar{X}_{n}$$

$$c_{2} = \bar{X}_{n}^{t} V_{n}^{-1} 1$$

$$c_{3} = 1^{t} V_{n}^{-1} 1$$

$$D = c_{1} c_{3} - c_{2}^{2}$$

Performance

If asset returns are IID then:

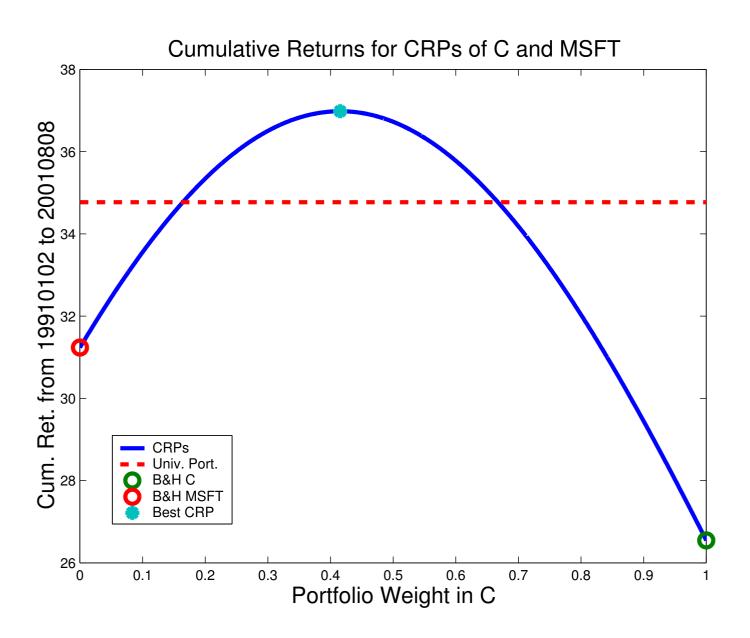
- Estimators converge to their true values by LLN
- MVE portfolio asymptotically achieves objective

Otherwise, no guarantee

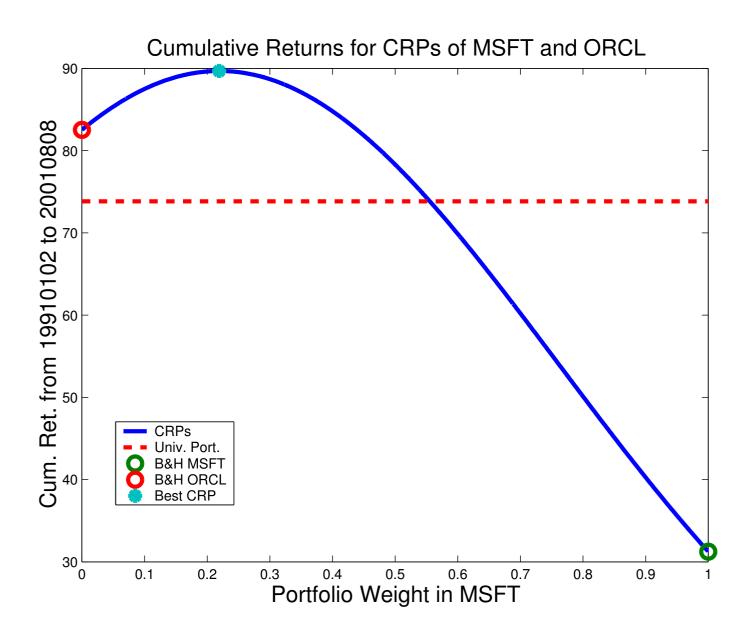
Examples

- Data:
 - Daily adjusted closing prices (yahoo.finance.com)
 - January 2,1991 to August 8, 2001 (2,679 days)
 - Missing data interpolated
- Examples:
 - Universal and constant rebalanced portfolios
 - Mean-Variance efficient portfolios
- Comparison:
 - MVE portfolio allows short sales, but univ. portfolio does not
 - No transaction costs
 - Investor preferences

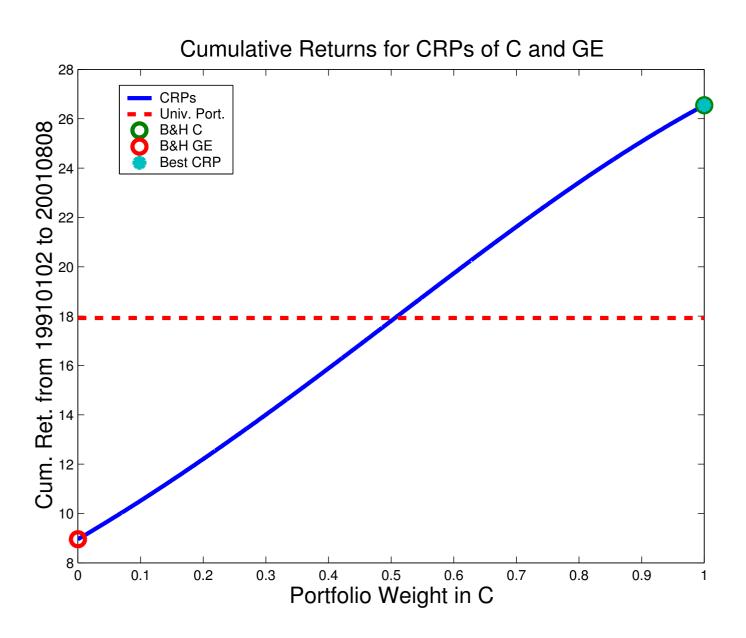
Univ. and CRPs of C & MSFT



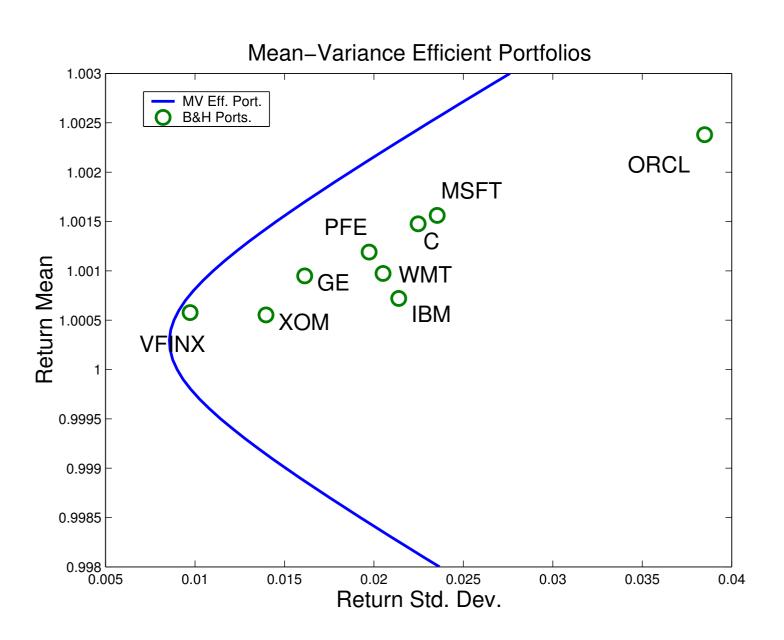
Univ. and CRPs of MSFT & ORCL



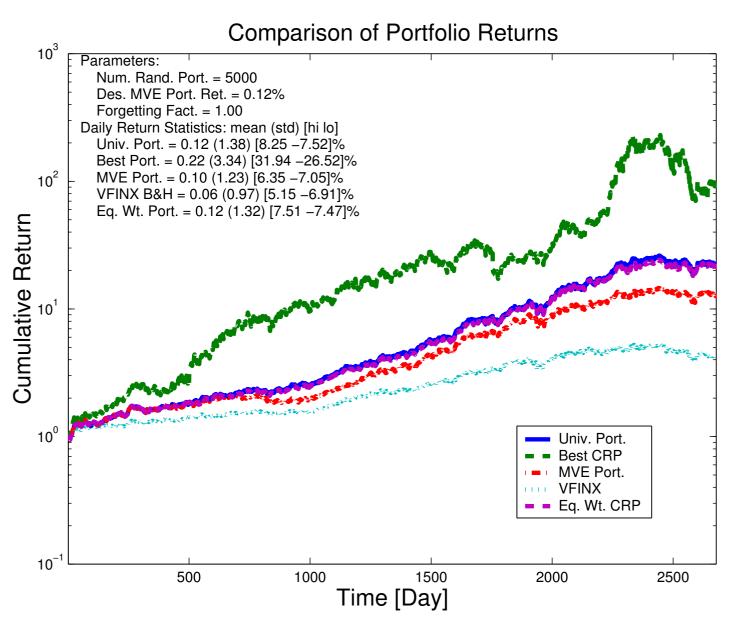
Univ. and CRPs of C & GE



MVE Portfolios



Comparison



Conclusions

- Universal portfolio:
 - Attempts to track return of best CRP given knowledge of future asset returns
 - Yields average return of all CRPs
 - Return greater than geometric mean
 - Asymptotically "tracks" return of best CRP
- MVE portfolio:
 - Attempts to minimize return variance while maintaining a desired average return
 - Obtains objective in IID markets