

# Universal Portfolios & GAs

A brief comparison between the performance of Thomas Cover universal portfolios and GAs' ones in investment decision

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# Introduction

A bankroll is described as such

$$X_n = X_0(1 + f)^S(1 - f)^F$$

$$\left(\frac{X_n}{X_0}\right)^{1/n} = (1 + f)^{S/n}(1 - f)^{F/n}$$

The term on the left is the growth rate per bet on average and we want to find an  $f$  that maximizes this quantity via the right-hand side. In log-world this is equivalent to

$$\frac{1}{n} (\log X_n - \log X_0) = \frac{S}{n} \log(1 + f) + \frac{F}{n} \log(1 - f),$$

or in the limit for  $n$  large we obtain:

$$\mathbb{E}[g(f)] = p \log(1 + f) + (1 - p) \log(1 - f)$$

# Introduction

Suppose we have  $n$  assets, with random return vector  $x \in \mathbb{R}^n$ , where the  $x_i$  are of the form  $x_i = \frac{p_i(\text{new})}{p_i(\text{old})}$ , i.e., relative price changes. In the spirit of Kelly's approach, we allocate fractions across these assets, so that the logarithmic growth rate is given as  $\log f^\top x$ .

$$f \in \Delta(n) \doteq \left\{ f \in \mathbb{R}^n \mid \sum_i f_i = 1, f \geq 0 \right\}$$

# Introduction

Thus, We would have the relative price change realizations  $x_t$  in time  $t$  and allocations  $f_t$ , so that the logarithmic portfolio growth over time is given by

$$\sum_{t=1}^T \log f_t^T x_t$$

Equivalently the average logarithmic growth rate is given by:

$$\frac{1}{T} \sum_{t=1}^T \log f_t^T x_t$$

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A Constant Rebalancing Portfolio (CRP) is one where the  $f_t = f$  are constant over time with the implicit assumption that the return distribution is somewhat stationary and hence from a sequential decision perspective we maximize the expected logarithmic growth rate by picking the expectation maximizer in each step assuming i.i.d. returns.

The algorithm constructs a portfolio over time, so that when  $T \rightarrow \infty$ , for any sequence of relative price changes  $x_1, \dots, x_T$ , then

$$\frac{1}{T} \sum_{t=1}^T \log f_t^T x_t \rightarrow \max_{f \in \mathbb{R}^n} \frac{1}{T} \sum_{t=1}^T \log f^T x_t$$

# Universal Portfolios

The total return after  $n$  days is  $T_n = \prod_{i=1}^n \mathbf{f}_i^T \mathbf{x}_i$ . With Universal Portfolios, we can approximate the return of optimal Constant Rebalancing Portfolio, as we defined it as a fixed wealth allocation throughout all days. Given a fixed wealth allocation  $\mathbf{f} \in \Delta^m$ , the return of CRP on day  $n$  is  $S_n(\mathbf{f}) = \prod_{i=1}^n \mathbf{f}^T \mathbf{x}_i$ . With Universal Portfolio algorithm, we can achieve

$$\frac{T_n}{S_n(\mathbf{f}^*)} \geq \frac{1}{(n+1)^{m-1}}$$

Thus we get an average daily return comparable with optimal CRP. Such implementations of Cover's algorithm are exponential in the number of stocks with worst-case run times of  $\Theta(n^{m-1})$ .

# Universal Portfolios

Here  $\mathbf{f}^*$  is the optimal CRP. Considering the average return per day, we have

$$\left( \frac{T_n}{S_n(\mathbf{f}^*)} \right)^{\frac{1}{n}} \geq \frac{1}{(n+1)^{\frac{m-1}{n}}}$$

As  $n \rightarrow \infty$ , the right hand side approach 1. Thus we get an average daily return comparable with optimal CRP.

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# Results from the paper

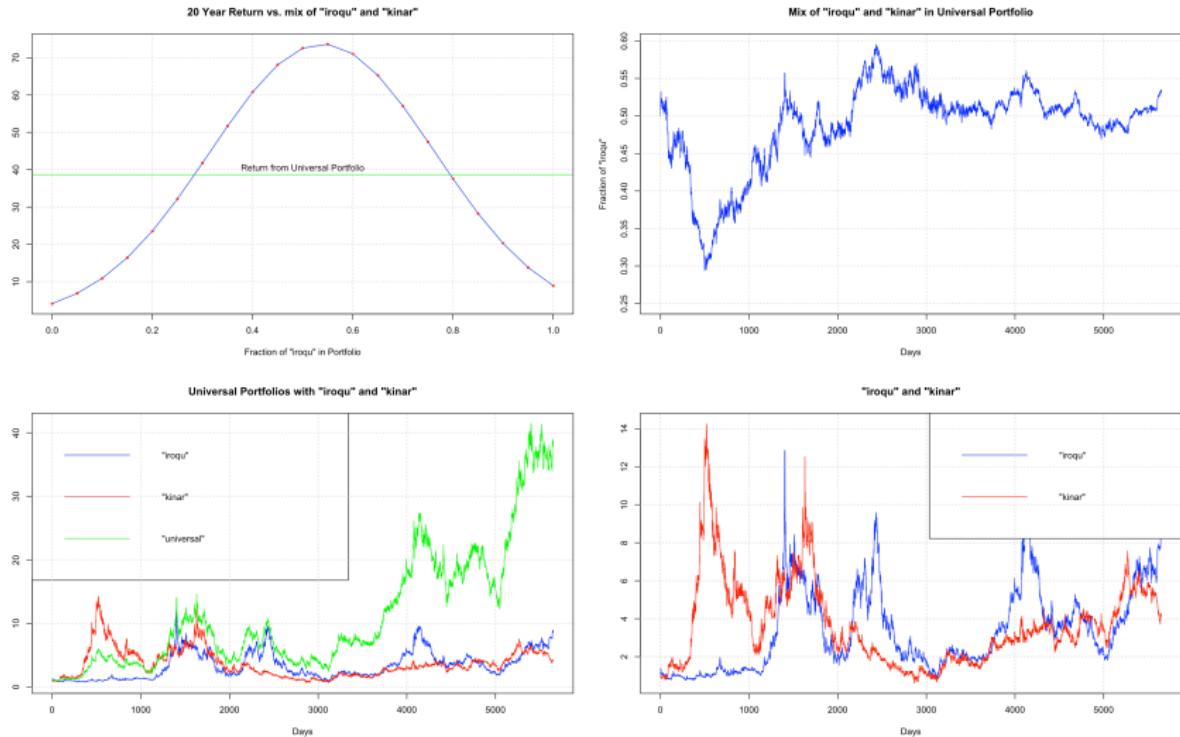
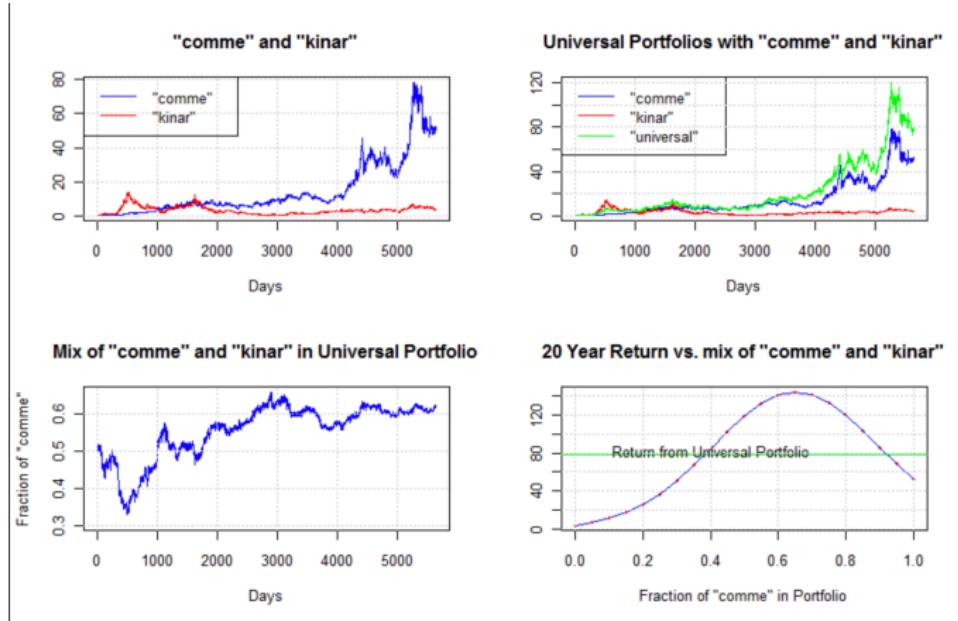
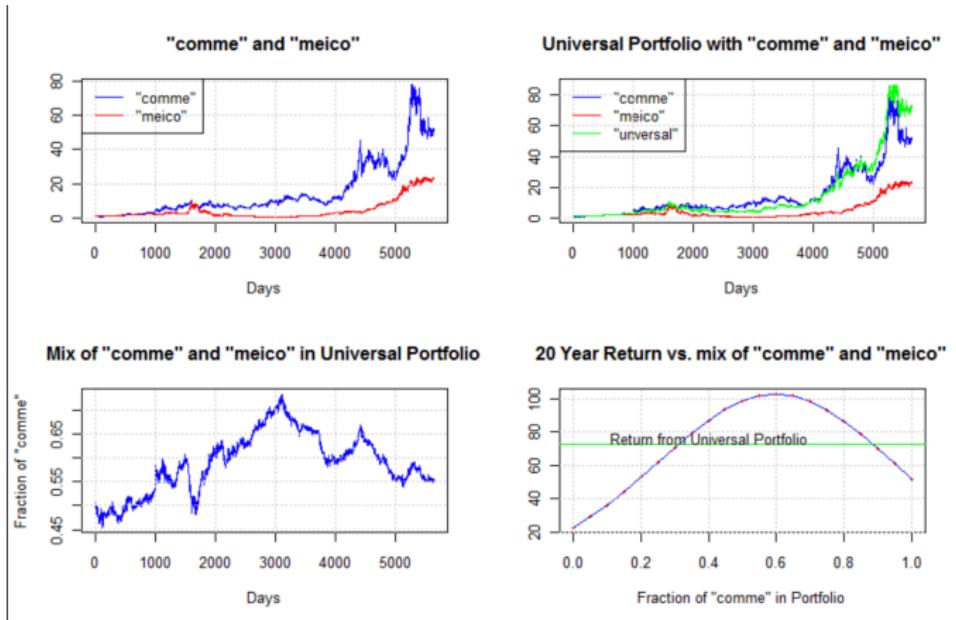


Figure: Results of Universal Portfolio for 2 stocks

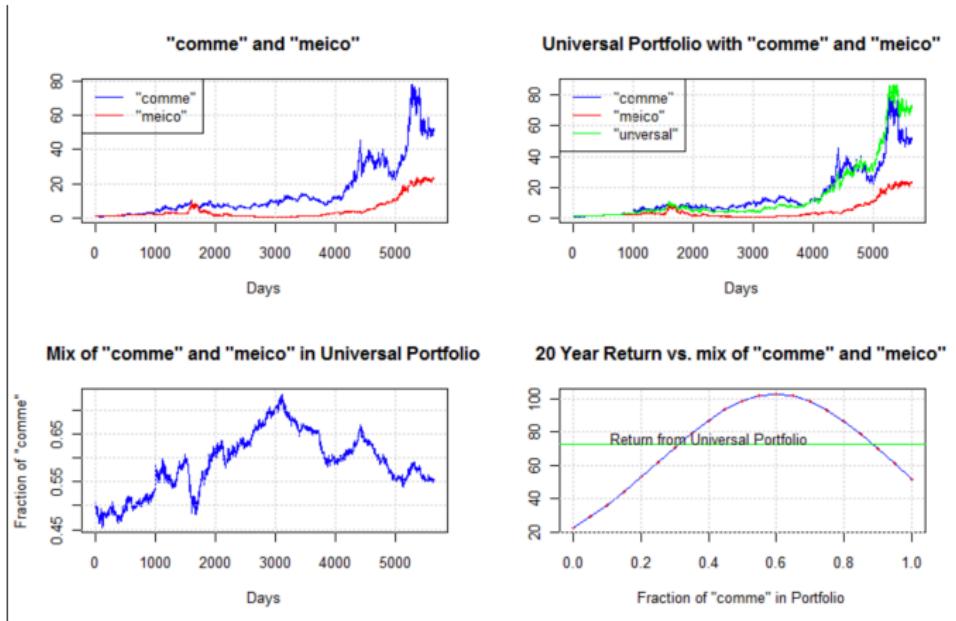
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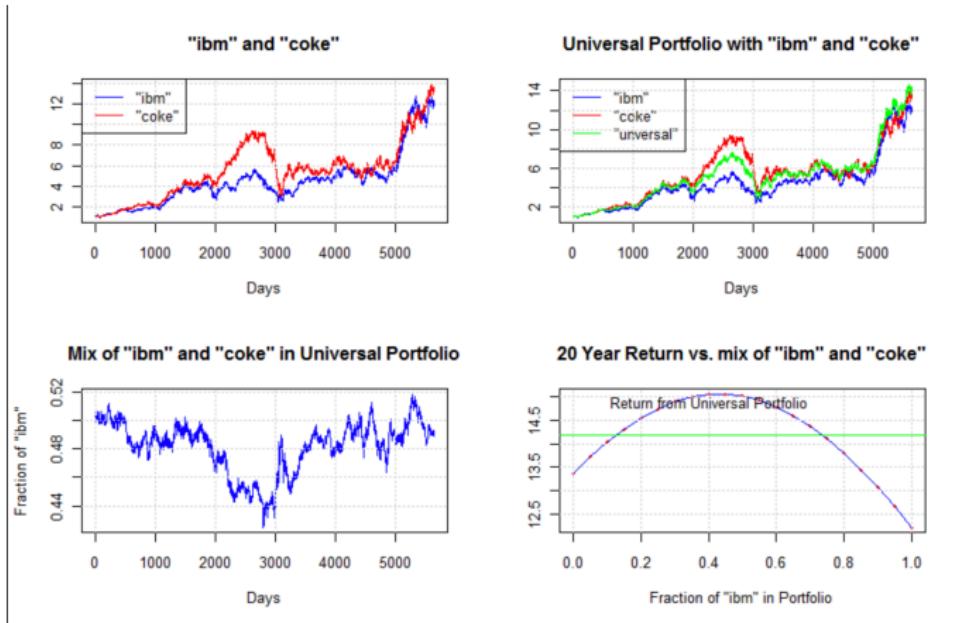
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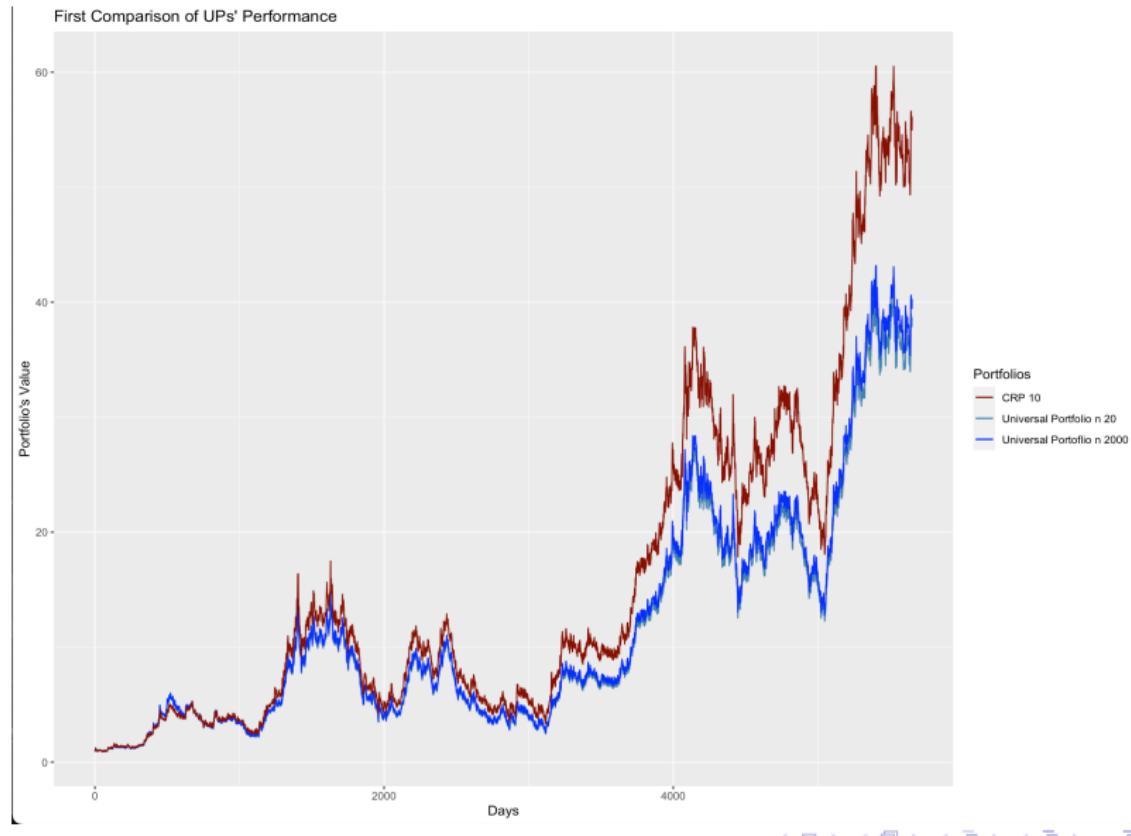
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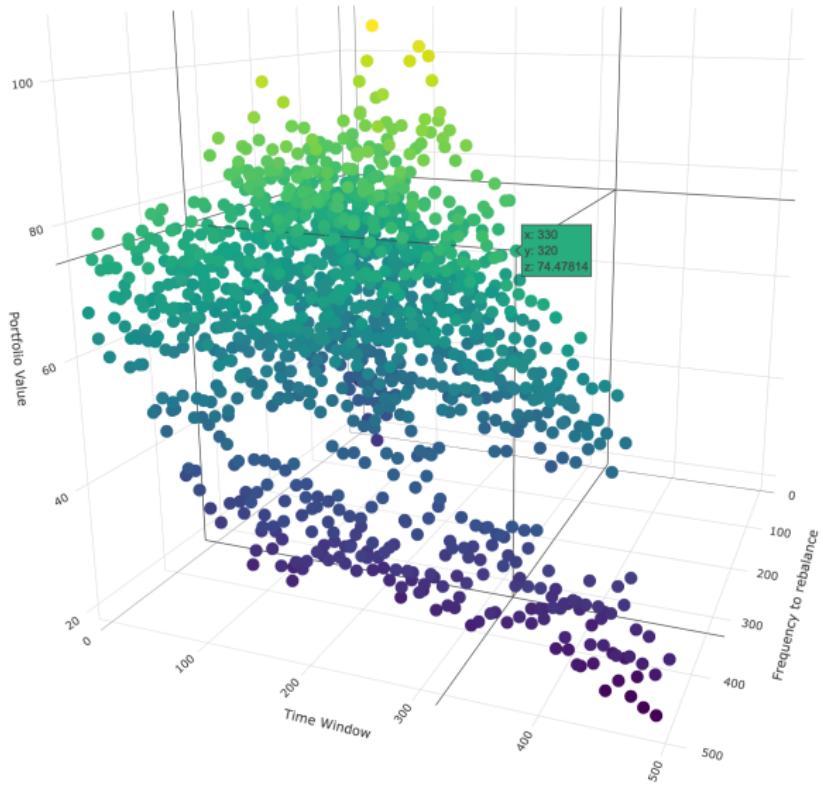
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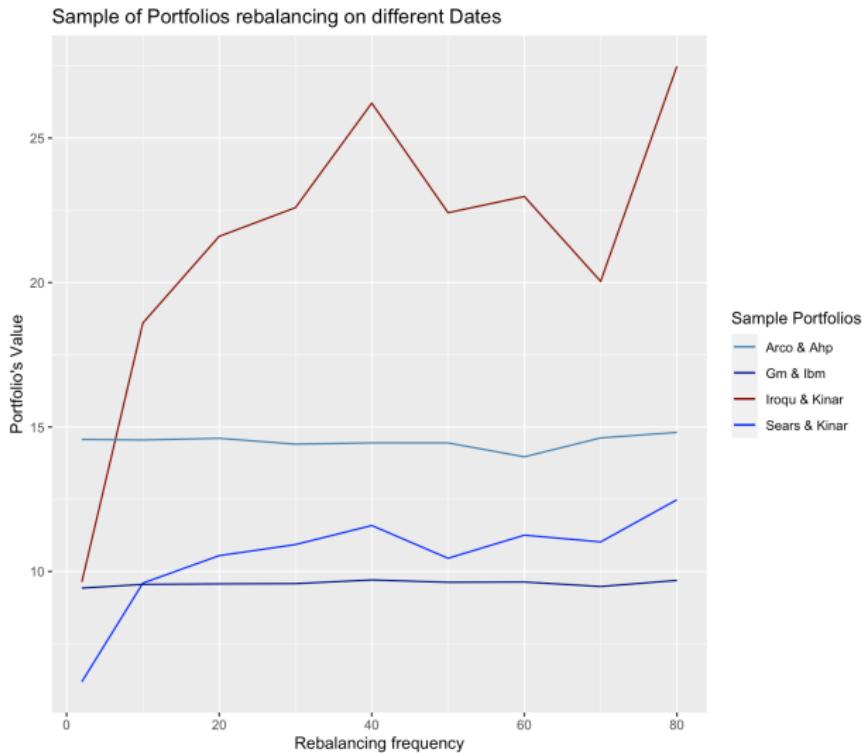
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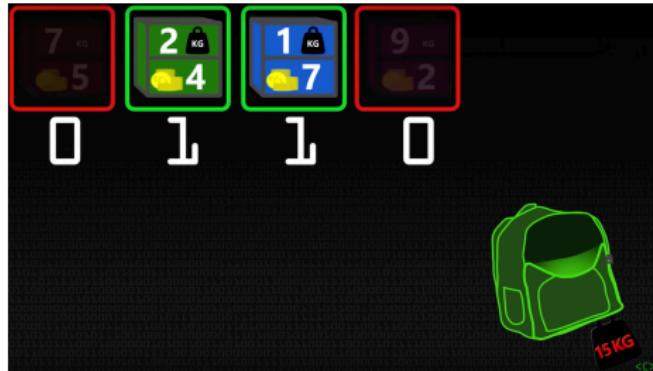
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# Knapsack Problem



**Figure:** The Knapsack problem is a well-known optimization problem in computer science and mathematics. It can be formulated as follows: given a set of items, each with a weight and a value, determine the subset of items that maximizes the total value while staying within a given weight limit.

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# Genetic Algorithm

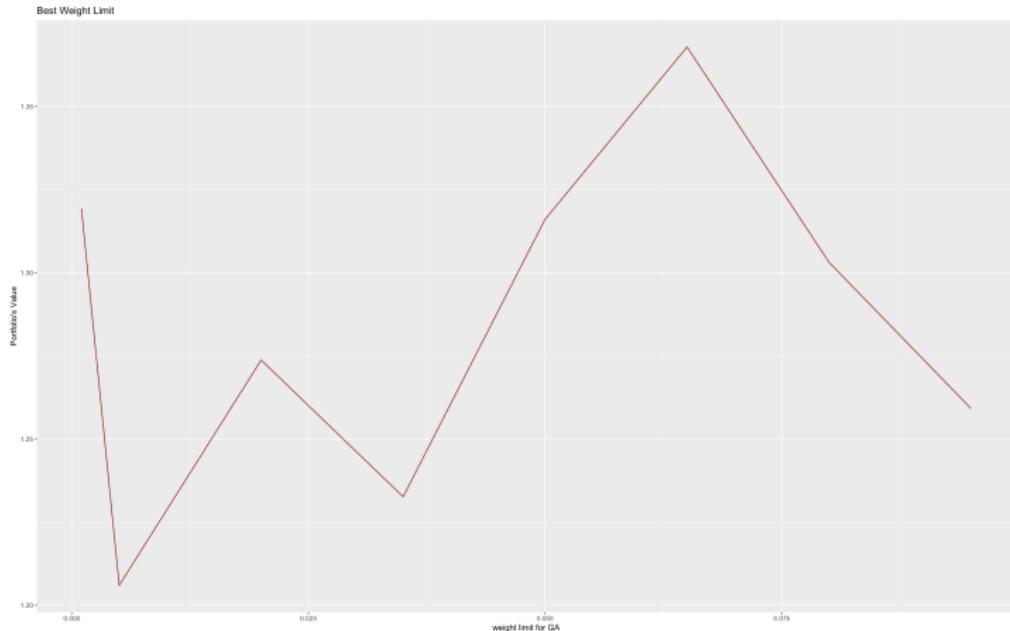


Figure: Performance under different weight limits for the GA

# Genetic Algorithm



Figure: Performance of GA's portfolio compared to other strategies

# End of the Presentation

Thanks for the attention.

All the scripts used can be found by this [github directory](#).

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