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Existence and Uniqueness for the Multivariate Discrete Terminal Wealth Relative

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Abstract: In this paper, the multivariate fractional trading ansatz of money management from Vince (Vince 1990) is discussed. In particular, we prove existence and uniqueness of an “optimal f ” of the respective optimization problem under reasonable assumptions on the trade return matrix. This result generalizes a similar result for the univariate fractional trading ansatz. Furthermore, our result guarantees that the multivariate optimal f solutions can always be found numerically by steepest ascent methods.

Keywords: fractional trading; optimal f ; multivariate discrete terminal wealth relative; risk and money management; portfolio theory

1. Introduction

Risk and money management for investment issues has always been at the heart of finance. Going back to the 1950s, Markowitz (1991) invented the “modern portfolio theory”, where the additive expectation of a portfolio of different investments was maximized subject to a given risk expressed by volatility of the portfolio.

When the returns of the portfolio are no longer calculated additive, but multiplicative in order to respect the needs of compound interest, the resulting optimization problem is known as “fixed fractional trading”. In fixed fractional trading strategies an investor always wants to risk a fixed percentage of his current capital for future investments given some distribution of historic trades of his trading strategy.

A first example of fractional trading was established in the 1950s by (Kelly 1956) who found a criterion for an asymptotically optimal investment strategy for one investment instrument. Similarly, Vince in the 1990s (see (Vince 1990, 1992)) used the fractional trading ansatz to optimize his position sizing. Although at first glance these two methods look quite different, they are in fact closely related as could be shown in (Maier-Paape 2016). However, only recently (Vince 2009) extended the fractional trading ansatz to portfolios of different investment instruments. The situation with M investment instruments (systems) and N coincident realizations of absolute returns of these M systems results in a trade return matrix T described in detail in (2). Given this trade return matrix, the “Terminal Wealth Relative” (TWR) can be constructed (see (4)) measuring the multiplicative gain of a portfolio resulting from a fixed vector $\varphi = (\varphi_1, \dots, \varphi_M)$ of fractional investments into the M systems. In order to find an optimal investment among all fractions φ the TWR has to be maximized

$$\underset{\varphi \in \mathfrak{G}}{\text{maximize}} \quad \text{TWR}(\varphi), \quad (1)$$

where \mathfrak{G} is the definition set of the TWR (see Definition 1 and (8)).

Whereas (Vince 2009) only stated this optimization problem and illustrated it with examples, in Section 3 we give as our main result the necessary analysis. In particular, we investigate the definition set \mathcal{G} of the TWR and fix reasonable assumptions (Assumption 1) under which (1) has a unique solution. This unique solution may lie in \mathcal{G} or on $\partial\mathcal{G}$ as different examples in Section 4 show. Our result extend the results of (Maier-Paape 2013; Zhu 2007) ($M = 1$ case only) and parts of the PhD of (Hermes 2016) on the discrete multivariate TWR. One of the main ingredients to show the uniqueness of the maximum of (1) is the concavity of the function $[\text{TWR}(\cdot)]^{1/N}$ (see Lemma 5). Uniqueness and concavity furthermore guarantee that the solution of (1) can always be found numerically by simply following steepest ascent.

Before we start our analysis, some more remarks on related papers are in order. In (Maier-Paape 2015) showed that the fractional trading ansatz on one investment instrument leads to tremendous drawdowns, but that effect can be reduced largely when several stochastic independent trading systems are used coincidentally. Under which conditions this diversification effect works out in the here considered multivariate TWR situation is still an open question. Furthermore, several papers investigated risk measures in the context of fractional trading with one investment instrument ($M = 1$; see (De Prado et al. 2013; Maier-Paape 2013, 2016; Vince and Zhu 2013)). Related investigations for the multivariate TWR using the drawdown can be found in (Vince 2009).

In the following sections, we analyse the multivariate case of a discrete Terminal Wealth Relative. That means we consider multiple investment strategies where every strategy generates multiple trading returns. As noted before this situation can be seen as a portfolio approach of a discrete Terminal Wealth Relative (cf. (Vince 2009)). For example, one could consider an investment strategy applied to several assets, the strategy producing trading returns on each asset. However, in an even broader sense, one could also consider several distinct investment strategies applied to several distinct assets or even classes of assets.

2. Definition of a Terminal Wealth Relative

The subject of consideration in this paper is the multivariate case of the discrete Terminal Wealth Relative for several trading systems analogous to the definition of Ralph Vince in (Vince 2009). For $1 \leq k \leq M$, $M \in \mathbb{N}$, we denote the k -th trading system by (system k). A trading system is an investment strategy applied to a financial instrument. Each system generates periodic trade returns, e.g., monthly, daily or the like. The absolute trade return of the i -th period of the k -th system is denoted by $t_{i,k}$, $1 \leq i \leq N, 1 \leq k \leq M$. Thus, we have the joint return matrix

period	(system 1)	(system 2)	...	(system M)
1	$t_{1,1}$	$t_{1,2}$...	$t_{1,M}$
2	$t_{2,1}$	$t_{2,2}$...	$t_{2,M}$
\vdots	\vdots	\vdots	\ddots	\vdots
N	$t_{N,1}$	$t_{N,2}$...	$t_{N,M}$

and define

$$T := \left(t_{i,k} \right)_{\substack{1 \leq i \leq N \\ 1 \leq k \leq M}} \in \mathbb{R}^{N \times M}. \tag{2}$$

Just as in the univariate case (cf. (Maier-Paape 2013) or (Vince 1990)), we assume that each system produced at least one loss within the N periods. That means

$$\forall k \in \{1, \dots, M\} \exists i_0 = i_0(k) \in \{1, \dots, N\} \text{ such that } t_{i_0,k} < 0. \tag{3}$$

Thus, we can define the biggest loss of each system as

$$\hat{t}_k := \max_{1 \leq i \leq N} \{|t_{i,k}| \mid t_{i,k} < 0\} > 0, \quad 1 \leq k \leq M.$$

For better readability, we define the rows of the given return matrix, i.e., the return of the i -th period, as

$$t_{i\cdot} := (t_{i,1}, \dots, t_{i,M}) \in \mathbb{R}^{1 \times M}$$

and the vector of all biggest losses as

$$\hat{t} := (\hat{t}_1, \dots, \hat{t}_M) \in \mathbb{R}^{1 \times M}.$$

Having the biggest losses at hand, it is possible to “normalize” the k -th column of T by $1/\hat{t}_k$ such that each system has a maximal loss of -1 . Using the componentwise quotient, the normalized trade matrix return then has the rows

$$(t_{i\cdot}/\hat{t}) := \left(\frac{t_{i,1}}{\hat{t}_1}, \dots, \frac{t_{i,M}}{\hat{t}_M} \right) \in \mathbb{R}^{1 \times M}, \quad 1 \leq i \leq N.$$

For $\varphi := (\varphi_1, \dots, \varphi_M)^\top$, $\varphi_k \in [0, 1]$, we define the Holding Period Return (HPR) of the i -th period as

$$\text{HPR}_i(\varphi) := 1 + \sum_{k=1}^M \varphi_k \frac{t_{i,k}}{\hat{t}_k} = 1 + \langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle_{\mathbb{R}^M}, \tag{4}$$

where $\langle \cdot, \cdot \rangle_{\mathbb{R}^M}$ denotes the standard scalar product on \mathbb{R}^M . To shorten the notation, the marking of the vector space \mathbb{R}^M at the scalar product is omitted, if the dimension of the vectors is clear. Similar to the univariate case, the gain (or loss) in each system is scaled by its biggest loss. Therefore the HPR represents the gain (loss) of one period, when investing a fraction of φ_k/\hat{t}_k of the capital in (system k) for all $1 \leq k \leq M$, thus risking a maximal loss of φ_k in the k -th trading system.

The Terminal Wealth Relative (TWR) as the gain (or loss) after the given N periods, when the fraction φ_k is invested in (system k) over all periods, is then given as

$$\begin{aligned} \text{TWR}_N(\varphi) &:= \prod_{i=1}^N \text{HPR}_i(\varphi) \\ &= \prod_{i=1}^N \left(1 + \sum_{k=1}^M \varphi_k \frac{t_{i,k}}{\hat{t}_k} \right) = \prod_{i=1}^N \left(1 + \langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle \right). \end{aligned} \tag{5}$$

Note that in the $M = 1$ -dimensional case a risk of a full loss of our capital corresponds to a fraction of $\varphi = 1 \in \mathbb{R}$. Here in the multivariate case we have a loss of 100% of our capital every time there exists an $i_0 \in \{1, \dots, N\}$ such that $\text{HPR}_{i_0}(\varphi) = 0$. That is for example if we risk a maximal loss of $\varphi_{k_0} = 1$ in the k_0 -th trading system (for some $k_0 \in \{1, \dots, M\}$) and simultaneously letting $\varphi_k = 0$ for all other $k \in \{1, \dots, M\}$. However these degenerate vectors of fractions are not the only examples that produce a Terminal Wealth Relative (TWR) of zero. Since we would like to risk at most 100% of our capital (which is quite a meaningful limitation), we restrict $\text{TWR}_N : \mathfrak{G} \rightarrow \mathbb{R}$ to the domain \mathfrak{G} given by the following definition:

Definition 1. A vector of fractions $\varphi \in \mathbb{R}_{\geq 0}^M$ is called admissible if $\varphi \in \mathfrak{G}$ holds, where

$$\begin{aligned} \mathfrak{G} &:= \{ \varphi \in \mathbb{R}_{\geq 0}^M \mid \text{HPR}_i(\varphi) \geq 0, \forall 1 \leq i \leq N \} \\ &= \{ \varphi \in \mathbb{R}_{\geq 0}^M \mid \langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle \geq -1, \forall 1 \leq i \leq N \}. \end{aligned}$$

Furthermore, we define

$$\mathfrak{R} := \{ \varphi \in \mathfrak{G} \mid \exists 1 \leq i_0 \leq N \text{ s.t. } \text{HPR}_{i_0}(\varphi) = 0 \}.$$

With this definition we now have a risk of 100% for each vector of fractions $\boldsymbol{\varphi} \in \mathfrak{R}$ and a risk of less than 100% for each vector of fractions $\boldsymbol{\varphi} \in \mathfrak{G} \setminus \mathfrak{R}$. Since

$$\text{HPR}_i(\mathbf{0}) = 1 \quad \text{for all } 1 \leq i \leq N$$

we can find an $\varepsilon > 0$ such that

$$\Lambda_\varepsilon := \{\boldsymbol{\varphi} \in \mathbb{R}_{\geq 0}^M \mid \|\boldsymbol{\varphi}\| \leq \varepsilon\} \subset \mathfrak{G},$$

and thus in particular $\mathfrak{G} \neq \emptyset$ holds. $\|\cdot\| = \sqrt{\langle \cdot, \cdot \rangle}$ denotes the Euclidean norm on \mathbb{R}^M .

Observe that the i -th period results in a loss if $\text{HPR}_i(\boldsymbol{\varphi}) < 1$, that means $\langle (t_i \cdot / \hat{t})^\top, \boldsymbol{\varphi} \rangle = \text{HPR}_i(\boldsymbol{\varphi}) - 1 < 0$. Hence the biggest loss over all periods for an investment with a given vector of fractions $\boldsymbol{\varphi} \in \mathfrak{G}$ is

$$r(\boldsymbol{\varphi}) := \max \left\{ - \min_{1 \leq i \leq N} \{ \langle (t_i \cdot / \hat{t})^\top, \boldsymbol{\varphi} \rangle, 0 \} \right\}. \tag{6}$$

Consequently, we have a biggest loss of

$$r(\boldsymbol{\varphi}) = 1 \quad \forall \boldsymbol{\varphi} \in \mathfrak{R}$$

and

$$r(\boldsymbol{\varphi}) \in [0, 1) \quad \forall \boldsymbol{\varphi} \in \mathfrak{G} \setminus \mathfrak{R}.$$

Note that for $\boldsymbol{\varphi} \in \mathfrak{G}$ we do not have an a priori bound for the fractions $\varphi_k, k = 1, \dots, M$. Thus it may happen that there are $\boldsymbol{\varphi} \in \mathfrak{G} \setminus \mathfrak{R}$ with $\varphi_k > 1$ for some (or even for all) $k \in \{1, \dots, M\}$, or at least $\sum_{k=1}^M \varphi_k > 1$, indicating a risk of more than 100% for the individual trading systems, but the combined risk of all trading systems $r(\boldsymbol{\varphi})$ can still be less than 100%. So the individual risks can potentially be eliminated to some extent through diversification. As a drawback of this favorable effect the optimization in the multivariate case may result in vectors of fractions $\boldsymbol{\varphi} \in \mathfrak{G}$ that require a high capitalization of the individual trading systems. Thus, we assume leveraged financial instruments and ignore margin calls or other regulatory issues.

Before we continue with the TWR analysis, let us state a first auxiliary lemma for \mathfrak{G} .

Lemma 1. *The set \mathfrak{G} in Definition 1 is convex, as is $\mathfrak{G} \setminus \mathfrak{R}$.*

Proof. All the conditions $\varphi_k \geq 0, k = 1, \dots, M$ and

$$\text{HPR}_i(\boldsymbol{\varphi}) \geq 0 \Leftrightarrow \langle (t_i \cdot / \hat{t})^\top, \boldsymbol{\varphi} \rangle \geq -1, \quad i = 1, \dots, N$$

define half spaces (which are convex). Since \mathfrak{G} is the intersection of a finite set of half spaces, it is itself convex.

A similar reasoning yields that $\mathfrak{G} \setminus \mathfrak{R}$ is convex, too. \square

3. Optimal Fraction of the Discrete Terminal Wealth Relative

If we develop this line of thought a little further a necessary condition for the return matrix T for the optimization of the Terminal Wealth Relative gets clear:

Lemma 2. *Assume there is a vector $\boldsymbol{\varphi}_0 \in \Lambda_\varepsilon$ with $r(\boldsymbol{\varphi}_0) = 0$ then*

$$\{s \cdot \boldsymbol{\varphi}_0 \mid s \in \mathbb{R}_{\geq 0}\} \subset \mathfrak{G} \setminus \mathfrak{R}.$$

If in addition there is an $1 \leq i_0 \leq N$ such that $\text{HPR}_{i_0}(\boldsymbol{\varphi}_0) > 1$ then

$$\text{TWR}_N(s \cdot \boldsymbol{\varphi}_0) \xrightarrow{s \rightarrow \infty} \infty.$$

Proof. If

$$r(\boldsymbol{\varphi}_0) = \max \left\{ - \min_{1 \leq i \leq N} \{ \langle (t_{i \cdot} / \hat{t})^\top, \boldsymbol{\varphi}_0 \rangle \}, 0 \right\} = 0,$$

it follows that

$$\text{HPR}_i(\boldsymbol{\varphi}_0) \geq 1 \quad \text{for all } 1 \leq i \leq N. \tag{7}$$

For arbitrary $s \in \mathbb{R}_{\geq 0}$ the function

$$s \mapsto \text{HPR}_i(s\boldsymbol{\varphi}_0) = 1 + \langle (t_{i \cdot} / \hat{t})^\top, s\boldsymbol{\varphi}_0 \rangle = 1 + s \underbrace{\langle (t_{i \cdot} / \hat{t})^\top, \boldsymbol{\varphi}_0 \rangle}_{\geq 0} \geq 1$$

is monotonically increasing in s for all $i = 1, \dots, N$ and by that we have

$$s\boldsymbol{\varphi}_0 \in \mathfrak{G} \setminus \mathfrak{R}.$$

Moreover, if there is an i_0 with $\text{HPR}_{i_0}(\boldsymbol{\varphi}_0) > 1$ then

$$\text{HPR}_{i_0}(s\boldsymbol{\varphi}_0) \xrightarrow{s \rightarrow \infty} \infty$$

and by that

$$\text{TWR}_N(s \cdot \boldsymbol{\varphi}_0) \xrightarrow{s \rightarrow \infty} \infty. \quad \square$$

An investment where the holding period returns are greater than or equal to 1 for all periods denotes a “risk free” investment ($r(\boldsymbol{\varphi}) = 0$) and considering the possibility of an unbounded leverage, it is clear that the overall profit can be maximized by investing an infinite quantity. Assuming arbitrage free investment instruments, any risk free investment can only be of short duration, hence by increasing $N \in \mathbb{N}$ the condition $\text{HPR}_i(\boldsymbol{\varphi}_0) \geq 1$ will eventually burst, cf. (7). Thus, when optimizing the Terminal Wealth Relative, we are interested in settings that fulfill the following assumption

$$\forall \boldsymbol{\varphi} \in \partial B_\varepsilon(0) \cap \Lambda_\varepsilon \exists i_0 = i_0(\boldsymbol{\varphi}) \text{ such that } \langle (t_{i_0 \cdot} / \hat{t})^\top, \boldsymbol{\varphi} \rangle < 0,$$

always yielding $r(\boldsymbol{\varphi}) > 0$.

With that at hand, we can formulate the optimization problem for the multivariate discrete Terminal Wealth Relative

$$\underset{\boldsymbol{\varphi} \in \mathfrak{G}}{\text{maximize}} \quad \text{TWR}_N(\boldsymbol{\varphi}) \tag{8}$$

and analyze the existence and uniqueness of an optimal vector of fractions for the problem under the assumption

Assumption 1. We assume that each of the trading systems in (2) produced at least one loss (cf. (3)) and furthermore

- (a) $\forall \boldsymbol{\varphi} \in \partial B_\varepsilon(0) \cap \Lambda_\varepsilon \exists i_0 = i_0(\boldsymbol{\varphi}) \in \{1, \dots, N\}$
such that $\langle (t_{i_0 \cdot} / \hat{t})^\top, \boldsymbol{\varphi} \rangle < 0$ (no risk free investment)
- (b) $\frac{1}{N} \sum_{i=1}^N t_{i,k} > 0 \quad \forall k = 1, \dots, M$ (each trading system is profitable)

(c) $\ker(T) = \{0\}$ (linear independent trading systems)

Assumption 1(a) ensures that, no matter how we allocate our portfolio (i.e., no matter what direction $\varphi \in \mathfrak{G}$ we choose), there is always at least one period that realizes a loss, i.e., there exists an i_0 with $HPR_{i_0}(\varphi) < 1$. Or in other words, not only are the investment systems all fraught with risk (cf. (3)), but there is also no possible risk free allocation of the systems.

The matrix T from (2) can be viewed as a linear mapping

$$T : \mathbb{R}^M \rightarrow \mathbb{R}^M,$$

“ $\ker(T)$ ” denotes the kernel of the matrix T in Assumption 1(c). Thus, this assumption is the linear independence of the trading systems, i.e., the linear independence of the columns

$$t_{\cdot,k} \in \mathbb{R}^N, \quad k \in \{1, \dots, M\}$$

of the matrix T . Hence with Assumption 1(c) it is not possible that there exists an $1 \leq k_0 \leq M$ and a $\psi \in \mathbb{R}^M \setminus \{0\}$ such that

$$(-\psi_{k_0}) \begin{pmatrix} t_{1,k_0} \\ \vdots \\ t_{N,k_0} \end{pmatrix} = \sum_{\substack{k=1 \\ k \neq k_0}}^M \psi_k \begin{pmatrix} t_{1,k} \\ \vdots \\ t_{N,k} \end{pmatrix},$$

which would make (system k_0) obsolete. So Assumption 1(c) is no actual restriction of the optimization problem.

Now we point out a first property of the Terminal Wealth Relative .

Lemma 3. Let the return matrix $T \in \mathbb{R}^{N \times M}$ (as in (2)) satisfy Assumption 1(a) then, for all $\varphi \in \mathfrak{G} \setminus \{0\}$, there exists an $s_0 = s_0(\varphi) > 0$ such that $TWR_N(s_0\varphi) = 0$. In fact $s_0\varphi \in \mathfrak{R}$.

Proof. For some arbitrary $\varphi \in \mathfrak{G} \setminus \{0\}$ we have $\frac{\varepsilon}{\|\varphi\|} \cdot \varphi \in \partial B_\varepsilon(0) \cap \Lambda_\varepsilon$. Then Assumption 1(a) yields the existence of an $i_0 \in \{1, \dots, N\}$ with $\langle (t_{i_0 \cdot} / \hat{t})^\top, \varphi \rangle < 0$. With

$$j_0 := \operatorname{argmin}_{1 \leq i \leq N} \{ \langle (t_{i \cdot} / \hat{t})^\top, \varphi \rangle \} \in \{1, \dots, N\}$$

and

$$s_0 := -\frac{1}{\langle (t_{j_0 \cdot} / \hat{t})^\top, \varphi \rangle} > 0$$

we get that

$$HPR_{j_0}(s_0\varphi) = 1 + \langle (t_{j_0 \cdot} / \hat{t})^\top, s_0\varphi \rangle = 1 + s_0 \langle (t_{j_0 \cdot} / \hat{t})^\top, \varphi \rangle = 0$$

and $HPR_i(s_0\varphi) \geq 0$ for all $i \neq j_0$. Hence $TWR_N(s_0\varphi) = 0$ and clearly $s_0\varphi \in \mathfrak{R}$ (cf. Definition 1). \square

Furthermore, the following holds.

Lemma 4. Let the return matrix $T \in \mathbb{R}^{N \times M}$ (as in (2)) satisfy Assumption 1(a) then the set \mathfrak{G} is compact.

Proof. For all $\varphi \in \partial B_\varepsilon(0) \cap \Lambda_\varepsilon$ Assumption 1(a) yields an $i_0(\varphi) \in \{1, \dots, N\}$ such that $\langle (t_{i_0 \cdot} / \hat{t})^\top, \varphi \rangle < 0$. With that we define

$$m : \partial B_\varepsilon(0) \cap \Lambda_\varepsilon \rightarrow \mathbb{R}, \varphi \mapsto m(\varphi) := \min_{1 \leq i \leq N} \{ \langle (t_{i \cdot} / \hat{t})^\top, \varphi \rangle \} < 0.$$

This function is continuous on the compact support $\partial B_\varepsilon(0) \cap \Lambda_\varepsilon$. Thus, the maximum exists

$$M := \max_{\varphi \in \partial B_\varepsilon(0) \cap \Lambda_\varepsilon} m(\varphi) < 0.$$

Consequently the function

$$g : \partial B_\varepsilon(0) \cap \Lambda_\varepsilon \rightarrow \mathbb{R}_{\geq 0}^M, \varphi \mapsto \frac{1}{|m(\varphi)|} \cdot \varphi$$

is well defined and continuous. Since for all $\varphi \in \partial B_\varepsilon(0) \cap \Lambda_\varepsilon$

$$\langle (t_{i\cdot}/\hat{t})^\top, \frac{1}{|m(\varphi)|} \varphi \rangle = \frac{\langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle}{|\min_{1 \leq i \leq N} \{ \langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle \}|} \geq -1 \quad \forall 1 \leq i \leq N$$

with equality for at least one index $\tilde{i}_0 \in \{1, \dots, N\}$, we have

$$\text{HPR}_i \left(\frac{1}{|m(\varphi)|} \varphi \right) \geq 0 \quad \forall 1 \leq i \leq N$$

and

$$\text{HPR}_{\tilde{i}_0} \left(\frac{1}{|m(\varphi)|} \varphi \right) = 0,$$

hence

$$\frac{1}{|m(\varphi)|} \varphi \in \mathfrak{R}.$$

Altogether we see that

$$g(\partial B_\varepsilon(0) \cap \Lambda_\varepsilon) = \left\{ \frac{1}{|m(\varphi)|} \cdot \varphi \mid \varphi \in \partial B_\varepsilon(0) \cap \Lambda_\varepsilon \right\} = \mathfrak{R},$$

thus the set \mathfrak{R} is bounded and connected as image of the compact set $\partial B_\varepsilon \cap \Lambda_\varepsilon$ under the continuous function g and by that the set \mathfrak{G} is compact. \square

Now we take a closer look at the third assumption for the optimization problem.

Lemma 5. *Let the return matrix $T \in \mathbb{R}^{N \times M}$ (as in (2)) satisfy Assumption 1(c) then $\text{TWR}_N^{1/N}$ is concave on $\mathfrak{G} \setminus \mathfrak{R}$. Moreover if there is a $\varphi_0 \in \mathfrak{G} \setminus \mathfrak{R}$ with $\nabla \text{TWR}_N(\varphi) = \mathbf{0}$, then $\text{TWR}_N^{1/N}$ is even strictly concave in φ_0 .*

Proof. For $\varphi \in \mathfrak{G} \setminus \mathfrak{R}$ the gradient of $\text{TWR}_N^{1/N}$ is given by the column vector

$$\begin{aligned} \nabla \text{TWR}_N^{1/N}(\varphi) &= \text{TWR}_N^{1/N}(\varphi) \cdot \frac{1}{N} \sum_{i=1}^N \frac{1}{1 + \sum_{k=1}^M \varphi_k \frac{t_{i,k}}{\hat{t}_k}} \cdot \begin{pmatrix} t_{i,1}/\hat{t}_1 \\ t_{i,2}/\hat{t}_2 \\ \vdots \\ t_{i,M}/\hat{t}_M \end{pmatrix} \\ &= \text{TWR}_N^{1/N}(\varphi) \cdot \frac{1}{N} \sum_{i=1}^N \frac{1}{1 + \langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle} \cdot (t_{i\cdot}/\hat{t})^\top \in \mathbb{R}^{M \times 1}, \end{aligned} \tag{9}$$

where $\text{TWR}_N^{1/N}(\varphi) > 0$. The Hessian-matrix is then given by

$$\begin{aligned} \text{Hess}_{\text{TWR}_N^{1/N}}(\varphi) &= \nabla \left[\left(\nabla \text{TWR}_N^{1/N}(\varphi) \right)^\top \right] \\ &= \nabla \left[\text{TWR}_N^{1/N}(\varphi) \cdot \frac{1}{N} \sum_{i=1}^N \frac{1}{1 + \langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle} (t_{i\cdot}/\hat{t})^\top \right] \end{aligned}$$

$$\begin{aligned}
 &= \nabla \text{TWR}_N^{1/N}(\boldsymbol{\varphi}) \cdot \frac{1}{N} \sum_{i=1}^N \frac{1}{1 + \langle (t_{i\cdot}/\hat{t})^\top, \boldsymbol{\varphi} \rangle} (t_{i\cdot}/\hat{t}) \\
 &+ \text{TWR}_N^{1/N}(\boldsymbol{\varphi}) \frac{1}{N} \sum_{i=1}^N \left(-\frac{1}{(1 + \langle (t_{i\cdot}/\hat{t})^\top, \boldsymbol{\varphi} \rangle)^2} (t_{i\cdot}/\hat{t})^\top \cdot (t_{i\cdot}/\hat{t}) \right) \\
 &= \text{TWR}_N^{1/N}(\boldsymbol{\varphi}) \underbrace{\left[\frac{1}{N^2} \sum_{i=1}^N \mathbf{y}_i^\top \sum_{i=1}^N \mathbf{y}_i - \frac{1}{N} \sum_{i=1}^N \mathbf{y}_i^\top \mathbf{y}_i \right]}_{=: -1/N \cdot B(\boldsymbol{\varphi}) \in \mathbb{R}^{M \times M}}
 \end{aligned}$$

where $\mathbf{y}_i := \frac{1}{1 + \langle (t_{i\cdot}/\hat{t})^\top, \boldsymbol{\varphi} \rangle} (t_{i\cdot}/\hat{t}) \in \mathbb{R}^{1 \times M}$ is a row vector. The matrix $B(\boldsymbol{\varphi})$ can be rearranged as

$$\begin{aligned}
 B(\boldsymbol{\varphi}) &= \sum_{i=1}^N \mathbf{y}_i^\top \mathbf{y}_i - \frac{1}{N} \left(\sum_{i=1}^N \mathbf{y}_i^\top \right) \left(\sum_{i=1}^N \mathbf{y}_i \right) \\
 &= \sum_{i=1}^N \mathbf{y}_i^\top \mathbf{y}_i - \frac{1}{N} \left[\sum_{i=1}^N \mathbf{y}_i^\top \left(\sum_{u=1}^N \mathbf{y}_u \right) \right] - \frac{1}{N} \left[\sum_{i=1}^N \left(\sum_{v=1}^N \mathbf{y}_v^\top \right) \mathbf{y}_i \right] \\
 &\quad + \frac{1}{N^2} \left(\sum_{i=1}^N 1 \right) \left(\sum_{v=1}^N \mathbf{y}_v^\top \right) \left(\sum_{u=1}^N \mathbf{y}_u \right) \\
 &= \sum_{i=1}^N \left[\mathbf{y}_i^\top \mathbf{y}_i - \mathbf{y}_i^\top \frac{1}{N} \left(\sum_{u=1}^N \mathbf{y}_u \right) - \frac{1}{N} \left(\sum_{v=1}^N \mathbf{y}_v^\top \right) \mathbf{y}_i \right. \\
 &\quad \left. + \frac{1}{N^2} \left(\sum_{v=1}^N \mathbf{y}_v^\top \right) \left(\sum_{u=1}^N \mathbf{y}_u \right) \right] \\
 &= \sum_{i=1}^N \left[\mathbf{y}_i^\top \left(\mathbf{y}_i - \frac{1}{N} \sum_{u=1}^N \mathbf{y}_u \right) - \frac{1}{N} \left(\sum_{v=1}^N \mathbf{y}_v^\top \right) \left(\mathbf{y}_i - \frac{1}{N} \sum_{u=1}^N \mathbf{y}_u \right) \right] \\
 &= \sum_{i=1}^N \left[\left(\mathbf{y}_i^\top - \frac{1}{N} \sum_{v=1}^N \mathbf{y}_v^\top \right) \underbrace{\left(\mathbf{y}_i - \frac{1}{N} \sum_{u=1}^N \mathbf{y}_u \right)}_{:= \mathbf{w}_i \in \mathbb{R}^{1 \times M}} \right] \\
 &= \sum_{i=1}^N \mathbf{w}_i^\top \mathbf{w}_i.
 \end{aligned}$$

Since the matrices $\mathbf{w}_i^\top \mathbf{w}_i$ are positive semi-definite for all $i = 1, \dots, N$, the same holds for $B(\boldsymbol{\varphi})$ and therefore $\text{TWR}_N^{1/N}$ is concave. Furthermore, if there is a $\boldsymbol{\varphi}_0 \in \mathfrak{G} \setminus \mathfrak{R}$ with

$$\begin{aligned}
 &\nabla \text{TWR}_N(\boldsymbol{\varphi}_0) = 0 \\
 \text{TWR}_N(\boldsymbol{\varphi}_0) > 0 &\Leftrightarrow \sum_{i=1}^N \frac{1}{1 + \langle (t_{i\cdot}/\hat{t})^\top, \boldsymbol{\varphi}_0 \rangle} (t_{i\cdot}/\hat{t}) = 0 \\
 &\Leftrightarrow \sum_{i=1}^N \mathbf{y}_i = 0,
 \end{aligned}$$

where $\mathbf{y}_i = \mathbf{y}_i(\boldsymbol{\varphi}_0)$, the matrix $B(\boldsymbol{\varphi}_0)$ further reduces to

$$B(\boldsymbol{\varphi}_0) = \sum_{i=1}^N \mathbf{y}_i^\top \mathbf{y}_i.$$

If $B(\varphi_0)$ is not strictly positive definite there is a $\psi = (\psi_1, \dots, \psi_M)^\top \in \mathbb{R}^M \setminus \{0\}$ such that

$$0 = \psi^\top B(\varphi_0) \psi = \sum_{i=1}^N \psi^\top y_i^\top y_i \psi = \sum_{i=1}^N \underbrace{\langle y_i^\top, \psi \rangle^2}_{\geq 0}$$

and we get that

$$\begin{aligned} \langle y_i^\top, \psi \rangle &= \frac{1}{1 + \langle (t_{i\cdot}/\hat{t})^\top, \varphi_0 \rangle} \langle (t_{i\cdot}/\hat{t})^\top, \psi \rangle = 0 \quad \forall 1 \leq i \leq N \\ \Leftrightarrow \langle (t_{i\cdot}/\hat{t})^\top, \psi \rangle &= 0 \quad \forall 1 \leq i \leq N, \end{aligned}$$

yielding a non trivial element in $\ker(T)$ and thus contradicting Assumption 1(c). Hence matrix $B(\varphi_0)$ is strictly positive definite and $\text{TWR}_N^{1/N}$ is strictly concave in φ_0 . \square

With this at hand we can state an existence and uniqueness result for the multivariate optimization problem.

Theorem 2. (optimal f existence) Given a return matrix $T = \begin{pmatrix} t_{i,k} \end{pmatrix}_{\substack{1 \leq i \leq N \\ 1 \leq k \leq M}}$ as in (2) that fulfills Assumption 1, then there exists a solution $\varphi_N^{opt} \in \mathfrak{G}$ of the optimization problem (8)

$$\underset{\varphi \in \mathfrak{G}}{\text{maximize}} \quad \text{TWR}_N(\varphi). \tag{10}$$

Furthermore, one of the following statements holds:

- (a) φ_N^{opt} is unique, or
- (b) $\varphi_N^{opt} \in \partial \mathfrak{G}$.

For both cases $\varphi_N^{opt} \neq 0$, $\varphi_N^{opt} \notin \mathfrak{R}$ and $\text{TWR}_N(\varphi_N^{opt}) > 1$ hold true.

Proof. We show existence and partly uniqueness of a maximum of the N -th root of TWR_N , yielding existence and partly uniqueness of a solution φ_N^{opt} of (10) with the claimed properties.

With Lemmas 1 and 4, the support \mathfrak{G} of the Terminal Wealth Relative is convex and compact. Hence the continuous function $\text{TWR}_N^{1/N}$ attains its maximum on \mathfrak{G} . For $\varphi = 0$ we get from (9)

$$\nabla \text{TWR}_N^{1/N}(0) = \underbrace{\text{TWR}_N^{1/N}(0)}_{=1} \cdot \frac{1}{N} \sum_{i=1}^N (t_{i\cdot}/\hat{t})^\top,$$

which is a vector whose components are strictly positive due to Assumption 1(b). Therefore $0 \in \mathfrak{G}$ is not a maximum of $\text{TWR}_N^{1/N}$ and a global maximum reaches a value greater than

$$\text{TWR}_N^{1/N}(0) = 1.$$

Since for all $\varphi \in \mathfrak{R}$

$$\text{TWR}_N^{1/N}(\varphi) = 0$$

holds, a maximum can not be attained in \mathfrak{R} either.

Now if there is a maximum on $\partial \mathfrak{G}$, assertion (b) holds together with the claimed properties. Alternatively, a maximum φ_0 is attained in the interior $\mathring{\mathfrak{G}}$. In this case, Lemma 5 yields the strict concavity of $\text{TWR}_N^{1/N}$ at φ_0 . Suppose there is another maximum $\varphi^* \in \mathfrak{G} \setminus \mathfrak{R}$ then the straight line connecting both maxima

$$L := \{t \cdot \varphi_0 + (1 - t) \cdot \varphi^* \mid t \in [0, 1]\}$$

is fully contained in the convex set $\mathfrak{G} \setminus \mathfrak{R}$ (cf. Lemma 1). Because of the concavity of $\text{TWR}_N^{1/N}$ all points of L have to be maxima, which is a contradiction to the strict concavity of $\text{TWR}_N^{1/N}$ in φ_0 . Thus, the maximum is unique and assertion (a) holds together with the claimed properties. \square

In the remainder of this section, we will further discuss case (b) in Theorem 2. We aim to show that the maximum $\varphi_N^{opt} \in \partial\mathfrak{G}$ is unique either, but we proof this using a completely different idea. In order to lay the grounds for this, first, we give a lemma:

Lemma 6. *If $T \in \mathbb{R}^{N \times M}$ from (2) is a return map satisfying Assumption 1 and if $M \geq 2$, then each return map $\tilde{T} \in \mathbb{R}^{N \times (M-1)}$, which results from T after eliminating one of its columns, is also a return map satisfying Assumption 1.*

Proof. Since each of the M trading systems of the return matrix $T \in \mathbb{R}^{N \times M}$ has a biggest loss \hat{t}_k , $1 \leq k \leq M$, the same holds for the $(M - 1)$ trading systems of the reduced matrix $\tilde{T} \in \mathbb{R}^{N \times (M-1)}$.

For \tilde{T} , Assumption 1(b),(c) follow straight from the respective properties of the matrix T .

Now let, without loss of generality, \tilde{T} be the matrix that results from T by eliminating the last column, i.e., the M -th trading system is omitted. Let $t_i^{(M-1)} \in \mathbb{R}^{M-1}$, $i = 1, \dots, N$, denote the rows of \tilde{T} and $\hat{t}^{(M-1)} \in \mathbb{R}^{M-1}$ the vector of biggest losses of \tilde{T} . Then for Assumption 1(a) we have to show that

$$\forall \varphi^{(M-1)} \in \partial B_\varepsilon^{(M-1)}(0) \cap \Lambda_\varepsilon^{(M-1)} \exists i_0 = i_0(\varphi^{(M-1)}) \in \{1, \dots, N\},$$

such that

$$\langle (t_{i_0 \cdot}^{(M-1)} / \hat{t}^{(M-1)})^\top, \varphi^{(M-1)} \rangle < 0. \tag{11}$$

Using Assumption 1(a) for matrix T and

$$\varphi^M := \begin{pmatrix} \varphi_1^{(M-1)} \\ \vdots \\ \varphi_{M-1}^{(M-1)} \\ 0 \end{pmatrix} \in \partial B_\varepsilon^{(M)}(0) \cap \Lambda_\varepsilon^{(M)},$$

the inequality

$$\langle (t_{i_0 \cdot} / \hat{t})^\top, \varphi^{(M)} \rangle < 0,$$

holds true. Thus, (11) holds likewise. \square

Having this at hand, we can now extend Theorem 2.

Corollary 1. (optimal uniqueness) *In the situation of Theorem 2 the uniqueness also holds for case (b), i.e., a maximum $\varphi_N^{opt} \in \partial\mathfrak{G}$ is also a unique maximum of $\text{TWR}_N(\varphi)$ in \mathfrak{G} .*

Proof. Assume that the optimal solution $\varphi_0 := \varphi_N^{opt} \in \partial\mathfrak{G}$ is not unique, then there exists an additional optimal solution $\varphi^* \in \partial\mathfrak{G}$ with $\varphi^* \neq \varphi_0$. Since $\mathfrak{G} \setminus \mathfrak{R}$ is convex (c.f. Lemma 1), the line connecting both solutions

$$L := \{t \cdot \varphi_0 + (1 - t) \cdot \varphi^* \mid t \in [0, 1]\}$$

is fully contained in $\mathfrak{G} \setminus \mathfrak{R}$. Because of the concavity of $\text{TWR}_N^{1/N}$ on $\mathfrak{G} \setminus \mathfrak{R}$ (c.f. Lemma 5), all points on L are optimal solutions. Therefore L must be a subset of $\partial\mathfrak{G} \setminus \mathfrak{R}$, since we have seen that an optimal solution in the interior $\mathring{\mathfrak{G}}$ would be unique. Hence, there is (at least) one $k_0 \in \{1, \dots, M\}$ such that, for all investment vectors in L , the trading system (system k_0) is not invested. i.e., the k_0 -th component of φ_0 , of φ^* and of all vectors in L is zero.

Without loss of generality, let $k_0 = M$. Then

$$\varphi_0 = \begin{pmatrix} \varphi_1 \\ \vdots \\ \varphi_{M-1} \\ 0 \end{pmatrix} \neq \begin{pmatrix} \varphi_1^* \\ \vdots \\ \varphi_{M-1}^* \\ 0 \end{pmatrix} = \varphi^*$$

are two optimal solutions for

$$\text{TWR}_N(\varphi) \stackrel{!}{=} \max$$

However, with that, the $(M - 1)$ -dimensional investment vectors $\varphi_0^{(M-1)} := (\tilde{\varphi}_1, \dots, \tilde{\varphi}_{M-1})^\top$ and $\varphi^{*,(M-1)} := (\varphi_1^*, \dots, \varphi_{M-1}^*)^\top$ are two distinct optimal solutions for

$$\text{TWR}_N^{(M-1)}\left(\begin{pmatrix} \varphi_1 \\ \vdots \\ \varphi_{M-1} \end{pmatrix}\right) := \prod_{i=1}^N \left(1 + \sum_{k=1}^{M-1} \varphi_k \frac{t_{i,k}}{\hat{t}}\right) \stackrel{!}{=} \max .$$

With Lemma 6 the return map $\tilde{T} \in \mathbb{R}^{N \times (M-1)}$, which results from T after eliminating the M -th column (i.e., (system M)) satisfies Assumption 1. Applying Theorem 2 to the sub-dimensional optimization problem, yields that $\varphi_0^{(M-1)}$ and $\varphi^{*,(M-1)}$ again lie at the boundary of the admissible set of investment vectors $\mathfrak{G}^{(M-1)} \subset \mathbb{R}^{M-1}$.

Hence, we have two distinct optimal solutions on the boundary $\partial\mathfrak{G}^{(M-1)}$ for the optimization problem with $(M - 1)$ investment systems. By induction this reasoning leads to the existence of two distinct optimal solutions for an optimization problem with just one single trading system. However, for that type of problem, we already know that the solution is unique (see for example (Maier-Paape 2013)), which causes a contradiction to our assumption. Thus, also for case (b) we have the uniqueness of the solution $\varphi_N^{opt} \in \partial\mathfrak{G}$. \square

Remark 1. Note that Assumption 1(c) is necessary for uniqueness. To give a counterexample imagine a return matrix T with two equal columns, meaning the same trading system is used twice. Let φ^{opt} be the optimal f for this one dimensional trading system. Then it is easy to see that $(\varphi^{opt}, 0)$, $(0, \varphi^{opt})$ and the straight line connecting these two points yield TWR optimal solutions for the return matrix T .

4. Examples

As an example we fix the joint return matrix $T := (t_{i,k})_{\substack{1 \leq i \leq 6 \\ 1 \leq k \leq 4}}$ for $M = 4$ trading systems and the returns from $N = 6$ periods given through the following table.

period	(system 1)	(system 2)	(system 3)	(system 4)
1	2	1	-1	1
2	2	-1/2	2	-1
3	-1/2	1	-1	2
4	1	2	2	-1
5	-1/2	-1/2	2	1
6	-1	-1	-1	-1

(12)

Obviously every system produced at least one loss within the 6 periods, thus the vector $\hat{t} = (\hat{t}_1, \hat{t}_2, \hat{t}_3, \hat{t}_4)^\top$ with

$$\hat{t}_k = \max_{1 \leq i \leq 6} \{|t_{i,k}| \mid t_{i,k} < 0\} = 1, \quad k = 1, \dots, 4,$$

is well-defined. For $\varphi \in \mathfrak{G} \setminus \mathfrak{R}$ the TWR_6 takes the form

$$\begin{aligned} TWR_6(\varphi) &= (1 + 2\varphi_1 + \varphi_2 - \varphi_3 + \varphi_4)(1 + 2\varphi_1 - \frac{1}{2}\varphi_2 + 2\varphi_3 - \varphi_4) \\ &\quad (1 - \frac{1}{2}\varphi_1 + \varphi_2 - 1\varphi_3 + 2\varphi_4)(1 + \varphi_1 + 2\varphi_2 + 2\varphi_3 - \varphi_4) \\ &\quad (1 - \frac{1}{2}\varphi_1 - \frac{1}{2}\varphi_2 + 2\varphi_3 + 1\varphi_4)(1 - \varphi_1 - \varphi_2 - \varphi_3 - \varphi_4), \end{aligned}$$

where the set of admissible vectors is given by

$$\begin{aligned} \mathfrak{G} &= \{\varphi \in \mathbb{R}_{\geq 0}^4 \mid \langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle \geq -1, \forall 1 \leq i \leq 6\} \\ &= \{\varphi \in \mathbb{R}_{\geq 0}^4 \mid \langle (t_{6\cdot}/\hat{t})^\top, \varphi \rangle = \min_{i=1,\dots,6} \langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle \geq -1\} \\ &= \{\varphi \in [0, 1]^4 \mid \varphi_1 + \varphi_2 + \varphi_3 + \varphi_4 \leq 1\}. \end{aligned}$$

Since for all $\varphi \in \mathfrak{G}$

$$\langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle \geq \langle (t_{6\cdot}/\hat{t})^\top, \varphi \rangle \geq -1 \quad \forall i = 1, \dots, 6$$

we have

$$\langle (t_{i\cdot}/\hat{t})^\top, \varphi \rangle = -1 \text{ for some } i \in \{1, \dots, 6\} \Rightarrow \langle (t_{6\cdot}/\hat{t})^\top, \varphi \rangle = -1.$$

Accordingly we get

$$\begin{aligned} \mathfrak{R} &= \{\varphi \in \mathfrak{G} \mid \exists 1 \leq i_0 \leq 6 \text{ s.t. } \langle (t_{i_0\cdot}/\hat{t})^\top, \varphi \rangle = -1\} \\ &= \{\varphi \in [0, 1]^4 \mid \varphi_1 + \varphi_2 + \varphi_3 + \varphi_4 = 1\}. \end{aligned}$$

When examining the 6-th row $t_{6\cdot} = (-1, -1, -1, -1)$ of the matrix T we observe that Assumption 1(a) is fulfilled with $i_0 = 6$. To see that let, for some $\varepsilon > 0$, $\varphi \in \partial B_\varepsilon \cap \Lambda_\varepsilon$, then

$$\langle (t_{6\cdot}/\hat{t})^\top, \varphi \rangle = -\varphi_1 - \varphi_2 - \varphi_3 - \varphi_4 < 0.$$

For Assumption 1(b) one can easily check that all four systems are “profitable”, since the mean values of all four columns in (12) are strictly positive. Lastly, for Assumption 1(c) we check that the rows of matrix T are linearly independent

$$\det \begin{vmatrix} t_{1\cdot} \\ t_{2\cdot} \\ t_{3\cdot} \\ t_{4\cdot} \end{vmatrix} = \det \begin{vmatrix} 2 & 1 & -1 & 1 \\ 2 & -1/2 & 2 & -1 \\ -1/2 & 1 & -1 & 2 \\ 1 & 2 & 2 & -1 \end{vmatrix} = 22.75 \neq 0.$$

Thus, Theorem 2 yields the existence and uniqueness of an optimal investment fraction $\varphi_6^{opt} \in \mathfrak{G}$ with $\varphi_6^{opt} \neq 0$, $\varphi_6^{opt} \notin \mathfrak{R}$ and $TWR_6(\varphi_6^{opt}) > 1$, which can numerically be computed

$$\varphi_6^{opt} \approx \begin{pmatrix} 0.2362 \\ 0.0570 \\ 0.1685 \\ 0.1012 \end{pmatrix}.$$

In the above example, a crucial point is that there is one row in the return matrix where the k -th entry is the biggest loss of (system k), $k = 1, \dots, 4$. Such a row in the return matrix implies, that all trading systems realized their biggest loss simultaneously, which can be seen as a strong evidence against a sufficient diversification of the systems. Hence we analyze Assumption 1(a) a little closer to see what happens if this is not the case.

With the help of Assumption 1(a), for all $\varphi \in \partial B_\varepsilon(0) \cap \Lambda_\varepsilon$, there is a row of the return matrix $t_{i_0 \cdot}$, $i_0 \in \{1, \dots, N\}$ such that $\langle (t_{i_0 \cdot} / \hat{t})^\top, \varphi \rangle < 0$. The sets

$$\{\varphi \in \mathbb{R}^M \mid \langle (t_{i \cdot} / \hat{t})^\top, \varphi \rangle = 0\}, \quad i = 1, \dots, N$$

describe the hyperplanes generated by the normal direction $(t_{i \cdot} / \hat{t})^\top \in \mathbb{R}^M, i = 1, \dots, N$. Thus, each φ from the set $\partial B_\varepsilon(0) \cap \Lambda_\varepsilon$ has to be an element of one of the half spaces

$$H_i := \{\varphi \in \mathbb{R}^M \mid \langle (t_{i \cdot} / \hat{t})^\top, \varphi \rangle < 0\}, \quad i = 1, \dots, N.$$

In other words the set $\partial B_\varepsilon(0) \cap \Lambda_\varepsilon$ has to be a subset of a union of half spaces

$$(\partial B_\varepsilon(0) \cap \Lambda_\varepsilon) \subset \bigcup_{i=1}^N H_i.$$

If there exists an index i_0 such that $t_{i_0,k} = -\hat{t}_k$ for all $1 \leq k \leq M$, then the normal direction of the corresponding hyperplane is

$$(t_{i_0 \cdot} / \hat{t})^\top = \begin{pmatrix} -1 \\ -1 \\ \vdots \\ -1 \end{pmatrix} \in \mathbb{R}^M, \tag{13}$$

hence

$$(\partial B_\varepsilon(0) \cap \Lambda_\varepsilon) \subset \mathbb{R}_{\geq 0}^M \subset H_{i_0}$$

and therefore Assumption 1(a) is fulfilled. Figure 1 shows a hyperplane for $M = 2$ and a row of the return matrix where all entries are the biggest losses, that means the normal direction of this hyperplane is the vector

$$\begin{pmatrix} -\hat{t}_1 \\ -\hat{t}_2 \end{pmatrix} / \begin{pmatrix} \hat{t}_1 \\ \hat{t}_2 \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \end{pmatrix}.$$

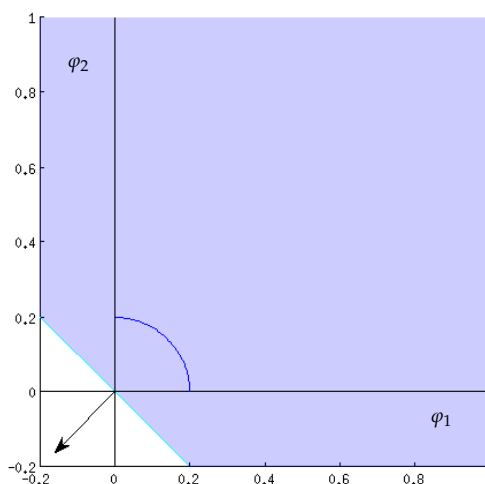


Figure 1. Hyperplane for a return vector consisting of “biggest losses”.

However, it is not necessary for Assumption 1(a) that the set $\partial B_\varepsilon(0) \cap \Lambda_\varepsilon$ is covered by just one hyperplane. Again for $M = 2$ an illustration of possible hyperplanes can be seen in Figure 2. The figure on the left shows a case where Assumption 1(a) is violated and the figure on the right a case where it is satisfied.

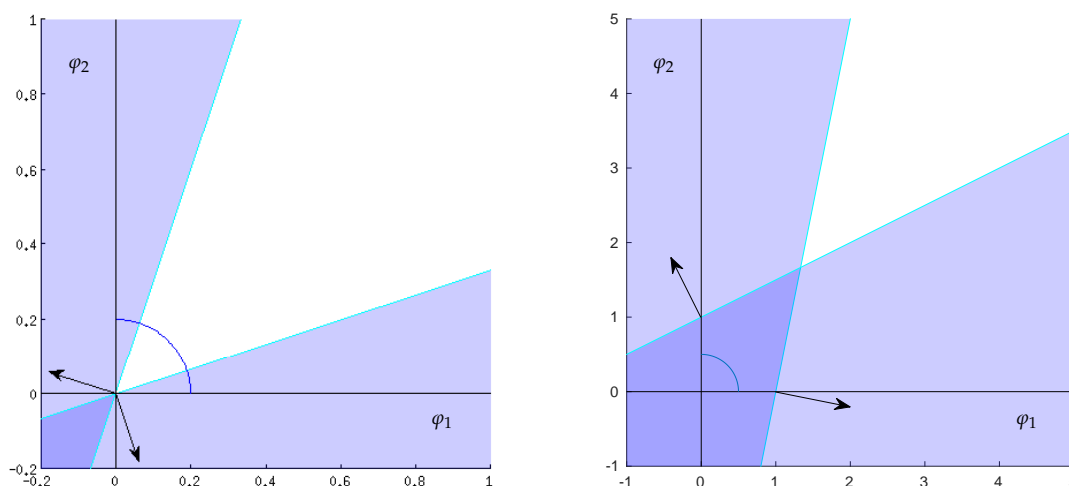


Figure 2. Two hyperplanes and the set $\partial B_\epsilon(0) \cap \Lambda_\epsilon$.

Remark 2. Following our arguments concerning the example in (12) the question arises whether or not a trading system where biggest losses are realized simultaneously always implies insufficient diversification. In (12) this seems to be the case, but is this true in general? One way diversification is often measured in the literature is volatility or variance/standard deviation of the portfolio returns. In terms of modern portfolio theory where portfolios are searched for which either

- minimize risk for a given chance/utility level
- or maximize chance/utility for a given risk level,

the volatility stands for the risk part. Transferring this setup for a trade-off between risk and utility to the TWR “utility function” would result in an optimization problem like

$$\begin{aligned} & \underset{\boldsymbol{\varphi} \in \mathfrak{G}}{\text{maximize}} && \text{TWR}_N^{1/N}(\boldsymbol{\varphi}) \\ & \text{subject to} && r(\boldsymbol{\varphi}) := \boldsymbol{\varphi}^T \boldsymbol{\Sigma} \boldsymbol{\varphi} \leq c, \end{aligned} \tag{14}$$

where $c \geq 0$ is a constant restricting the risk (or volatility) level and $\boldsymbol{\Sigma} \in \mathbb{R}^{M \times M}$ is a symmetric positive definite covariance matrix stemming from a trading game with trade returns as in (2). The optimization problem in (14) is quite similar to the Markowitz portfolio optimization. The only difference is that the Markowitz utility function “expected portfolio return” is exchanged by the concave function $\text{TWR}_N^{1/N}$.

Since $\boldsymbol{\varphi}_N^{opt}$ of Theorem 2 solves (10) it is clear that $\boldsymbol{\varphi} = \boldsymbol{\varphi}_N^{opt}$ will solve (14) for all $c \geq r(\boldsymbol{\varphi}_N^{opt})$. On the other hand, for $c \in [0, r(\boldsymbol{\varphi}_N^{opt}))$ solutions of (14) will also be “efficient” in this utility/risk setting. However, the volatility decreases as $c \searrow 0$ and therefore diversification certainly increases. As a matter of fact, among all efficient portfolios, $\boldsymbol{\varphi}_N^{opt}$ has the highest volatility and thus the worst diversification. In that sense diversification of $\boldsymbol{\varphi}_N^{opt}$ is not to be expected.

For the next example we fix the return matrix T as

$$T := \frac{1}{5} \begin{pmatrix} -3 & 3 \\ 9 & 12 \\ 6 & -3 \\ -6 & 3/2 \\ 3 & -15/2 \end{pmatrix}, \tag{15}$$

with $N = 5$ and $M = 2$. Thus, the biggest losses of the two systems are

$$\hat{t}_1 = \frac{6}{5} \quad \text{and} \quad \hat{t}_2 = \frac{3}{2}.$$

To determine the set of admissible investments (and to check Assumption 1) we examine the vectors $(t_i \cdot / \hat{t})$ for $i = 1, \dots, 5$

$$A := \begin{pmatrix} -1/2 & 2/5 \\ 3/2 & 8/5 \\ 1 & -2/5 \\ -1 & 1/5 \\ 1/2 & -1 \end{pmatrix} \tag{16}$$

and solve the linear equations

$$\langle (t_i \cdot / \hat{t})^\top, \boldsymbol{\varphi} \rangle = -1, \quad i = 1, \dots, 5. \tag{17}$$

The solutions for $i = 1, \dots, 5$ are shown in Figure 3.

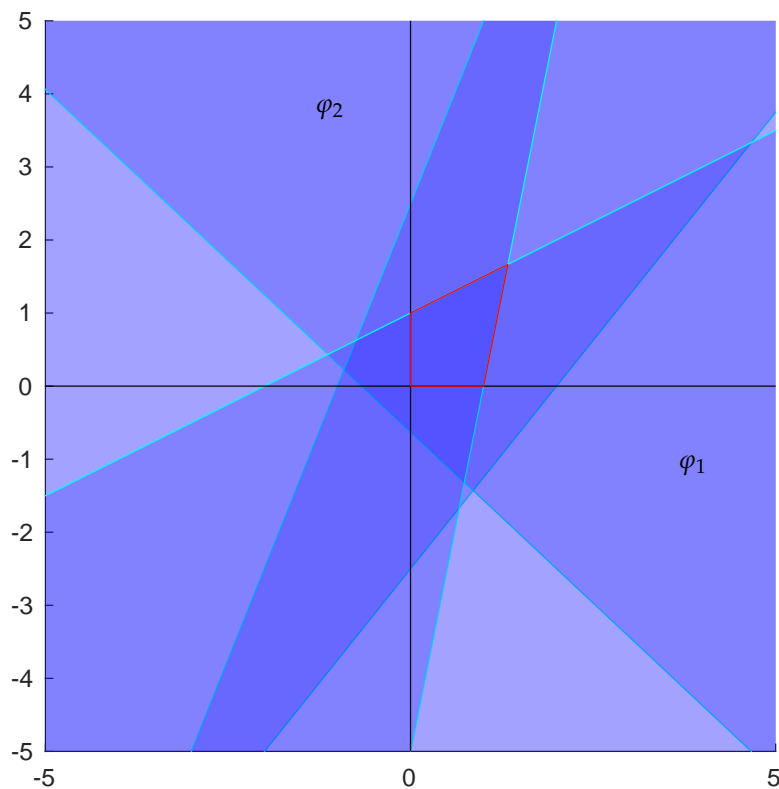


Figure 3. Solutions of the linear equations from (17).

Each solution corresponds to a “cyan” line. The area where the inequality $\langle (t_i \cdot / \hat{t})^\top, \boldsymbol{\varphi} \rangle \geq -1$ holds for some $i \in \{1, \dots, 5\}$ is shaded in “light blue”. The set where the inequalities hold for all $i = 1, \dots, 5$ is the section where all shaded areas overlap, thus the “dark blue” section. Therefore the set of admissible investments is given by

$$\begin{aligned} \mathfrak{S} &= \{ \boldsymbol{\varphi} \in \mathbb{R}_{\geq 0}^2 \mid \langle (t_i \cdot / \hat{t})^\top, \boldsymbol{\varphi} \rangle \geq -1, \forall 1 \leq i \leq 5 \} \\ &= \{ \boldsymbol{\varphi} \in \mathbb{R}_{\geq 0}^2 \mid \varphi_2 \leq 1 + \frac{1}{2}\varphi_1 \text{ and } \varphi_1 \leq 1 + \frac{1}{5}\varphi_2 \}, \end{aligned}$$

with

$$\begin{aligned} \mathfrak{R} &= \{ \boldsymbol{\varphi} \in \mathfrak{G} \mid \exists 1 \leq i_0 \leq 5 \text{ s.t. } \langle (t_{i_0} \cdot / \hat{t})^\top, \boldsymbol{\varphi} \rangle = -1 \} \\ &= \{ \boldsymbol{\varphi} \in \mathbb{R}_{\geq 0}^2 \mid \varphi_2 = 1 + \frac{1}{2}\varphi_1 \text{ or } \varphi_1 = 1 + \frac{1}{5}\varphi_2 \}. \end{aligned}$$

Assumption 1 is fulfilled, since

- (a) the half spaces for rows 4 and 5 of the return matrix cover the whole set $\mathbb{R}_{\geq 0}^2$ (cf. Figure 2b),
- (b) $\frac{1}{5} \sum_{i=1}^5 t_{i,1} = \frac{9}{5} > 0$ and $\frac{1}{5} \sum_{i=1}^5 t_{i,2} = \frac{6}{5} > 0$ and
- (c) obviously, the columns of the return matrix are linearly independent.

A plot of the Terminal Wealth Relative for the return matrix T from (15) can be seen in Figures 4 and 5 with a maximum at

$$\boldsymbol{\varphi}_5^{opt} \approx \begin{pmatrix} 0.4109 \\ 0.3425 \end{pmatrix}. \tag{18}$$

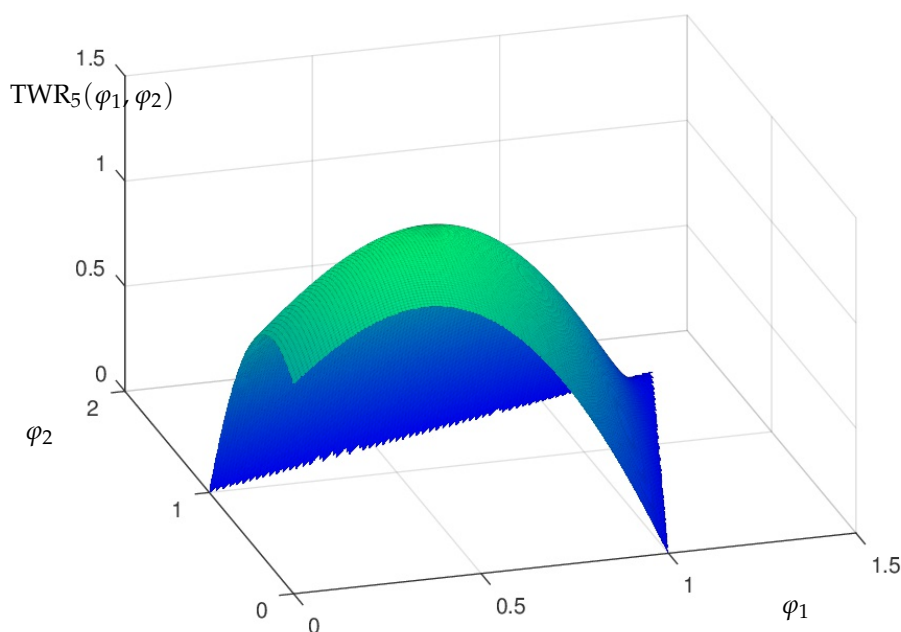


Figure 4. The Terminal Wealth Relative for T from (15).

Therefore the maximum is clearly attained in the interior $\mathring{\mathfrak{G}}$.

The following example will show that the unique maximum $\boldsymbol{\varphi}_N^{opt}$ of Theorem 2 can indeed be attained on $\partial\mathfrak{G}$, i.e., the case discussed in Corollary 1. For that we add a third investment system to our last example (16) with the new returns

$$t_{1,3}, t_{2,3}, t_{3,3} = 1 \text{ and } t_{4,3}, t_{5,3} = -1 \quad (\text{hence } \hat{t}_3 = 1)$$

such that the vectors $(t_{i \cdot} / \hat{t})$, $i = 1, \dots, 5$, form the matrix

$$\tilde{A} := (a_{i,k})_{\substack{i=1,\dots,5 \\ k=1,\dots,3}} = \begin{pmatrix} -1/2 & 2/5 & 1 \\ 3/2 & 8/5 & 1 \\ 1 & -2/5 & 1 \\ -1 & 1/5 & -1 \\ 1/2 & -1 & -1 \end{pmatrix} \in \mathbb{R}^{5 \times 3}. \tag{19}$$

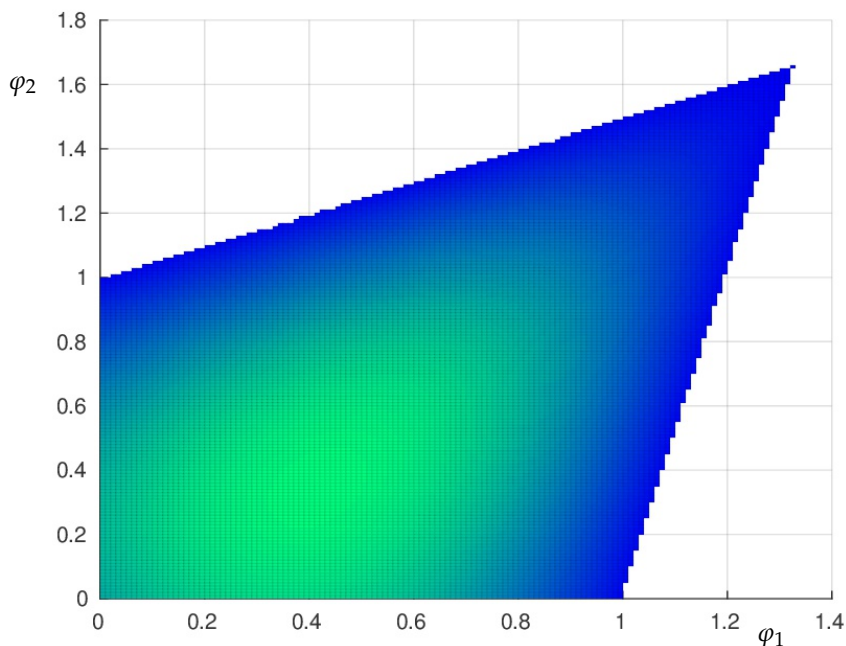


Figure 5. The Terminal Wealth Relative from Figure 4, view from above.

This set of trading systems fulfills Assumption 1(b) since $\sum_{i=1}^{N=5} t_{i,3} = 1 > 0$.

Assumption 1(c) is satisfied as well, because the three columns of \tilde{A} are linearly independent. For Assumption 1(a) we have to show that

$$\forall \boldsymbol{\varphi} \in \partial B_\varepsilon(0) \cap \Lambda_\varepsilon \exists i_0 = i_0(\boldsymbol{\varphi}), \text{ with } \langle (t_{i_0} \cdot / i)^T, \boldsymbol{\varphi} \rangle < 0 \tag{20}$$

holds. If not, we would have an investment vector

$$\hat{\boldsymbol{\varphi}} = (\hat{\varphi}_1, \hat{\varphi}_2, \hat{f}) \in \partial B_\varepsilon(0) \cap \Lambda_\varepsilon,$$

such that (20) is not true for all rows of the matrix \tilde{A} . In particular if we look at lines 4 and 5

$$\begin{aligned} -\hat{\varphi}_1 + \frac{1}{5}\hat{\varphi}_2 - \hat{f} &\geq 0 \\ \frac{1}{2}\hat{\varphi}_1 - \hat{\varphi}_2 - \hat{f} &\geq 0, \end{aligned}$$

the sum of both inequalities still has to be true

$$-\frac{1}{2}\hat{\varphi}_1 - \frac{4}{5}\hat{\varphi}_2 - 2\hat{f} \geq 0,$$

which is a contradiction to $\hat{\boldsymbol{\varphi}}$ being an element of $\partial B_\varepsilon(0) \cap \Lambda_\varepsilon \subset \mathbb{R}_{\geq 0}^3$.

Now we examine the following vector of investments

$$\boldsymbol{\varphi}^* = \begin{pmatrix} \varphi_1^* \\ \varphi_2^* \\ f^* \end{pmatrix} := \begin{pmatrix} \varphi_1^* \\ \varphi_2^* \\ 0 \end{pmatrix}$$

with $(\varphi_1^*, \varphi_2^*)^\top \approx (0.4109, 0.3425)^\top$ the unique maximum of the optimization problem of the reduced set of trading systems from the last example (cf. (18)).

The first derivative of the Terminal Wealth Relative in the direction of the third component at φ^* is given by

$$\frac{\partial}{\partial f} \text{TWR}_5(\varphi^*) = \underbrace{\text{TWR}_5(\varphi^*)}_{>0} \cdot \sum_{i=1}^{N=5} \frac{a_{i,3}}{1 + \langle (t_i \cdot / \hat{t})^\top, \varphi^* \rangle} \approx -0.359 < 0$$

Moreover with φ^* being the optimal solution of the last example in two variables we have

$$\frac{\partial}{\partial \varphi_1} \text{TWR}_5(\varphi_1^*, \varphi_2^*, 0) = 0 = \frac{\partial}{\partial \varphi_2} \text{TWR}_5(\varphi_1^*, \varphi_2^*, 0)$$

and

$$\frac{\partial^2}{\partial \varphi_i^2} \text{TWR}_5(\varphi_1^*, \varphi_2^*, 0) < 0, \quad i = 1, 2.$$

Thus, φ^* is indeed a local maximal point on the boundary of \mathfrak{G} for TWR_5 with the three trading systems in (19). Corollary 1 yields the uniqueness of this maximal solution for

$$\underset{\varphi \in \mathfrak{G}}{\text{maximize}} \quad \text{TWR}_5(\varphi).$$

5. Conclusions

With our main theorems, Theorem 2 and Corollary 1, we were able to give a complete existence and uniqueness theory for the optimization problem (8) of a multivariate Terminal Wealth Relative under reasonable assumptions. Furthermore, due to the convexity of the domain \mathfrak{G} (Lemma 1), the concavity of $[\text{TWR}(\cdot)]^{1/N}$ (see Lemma 5) and the uniqueness of the “optimal f ” solution, it is always guaranteed that simple numerical methods like steepest ascent will find the maximum.

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References

- Hermes, Andreas. 2016. A Mathematical Approach to Fractional Trading. Ph.D. thesis. RWTH Aachen University, Aachen, North Rhine-Westphalia, Germany.
- Kelly, John L., Jr. 1956. A new interpretation of information rate. *Bell System Technical Journal* 35: 917–26.
- De Prado, Marcos Lopez, Ralph Vince, and Qiji Jim Zhu. 2013. Optimal Risk Budgeting Under a Finite Investment Horizon. Available online: https://papers.ssrn.com/sol3/papers.cfm?abstract_id=2364092 (accessed on 1 January 2013).
- Maier-Paape, Stanislaus. 2013. *Existence Theorems for Optimal Fractional Trading*. Aachen: Institute for Mathematics, RWTH Aachen, report no. 67.
- Maier-Paape, Stanislaus. 2015. Optimal f and diversification. *International Federation of Technical Analysis Journal* 15: 4–7.
- Maier-Paape, Stanislaus. 2016. Risk averse fractional trading using the current drawdown. arXiv:1612.02985.
- Markowitz, Harry M. 1991. *Portfolio Selection*. München: FinanzBuch Verlag.
- Vince, Ralph. 1990. *Portfolio Management Formulas: Mathematical Trading Methods for the Futures, Options, and Stock Markets*. Hoboken: John Wiley & Sons, Inc.
- Vince, Ralph. 1992. *The Mathematics of Money Management, Risk Analysis Techniques for Traders*. Hoboken: John Wiley & Sons, Inc.
- Vince, Ralph. 2009. *The Leverage Space Trading Model: Reconciling Portfolio Management Strategies and Economic Theory*. Hoboken: Wiley Trading.

Vince, Ralph, and Qiji Jim Zhu. 2013. Inflection Point Significance for the Investment Size. Available online: https://papers.ssrn.com/sol3/papers.cfm?abstract_id=2230874 (accessed on 27 February 2013).

Zhu, Qiji Jim. 2007. Mathematical analysis of investment systems. *Journal of Mathematical Analysis and Applications* 326: 708–20.



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