

**Multiperiod Asset-Liability Management  
in a Mean-Variance Framework  
with Exogenous and Endogenous Liabilities**

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# Outline

1. Objective and Motivation of the paper
2. Reformulation of the problem
3. Main Theorem
4. Asset-only vs Asset-Liability (exogenous) Optimization
5. Exogenous vs Endogenous Liabilities
6. Conclusion

# Objective of the paper

## General:

- Extend Markowitz's basic intuition to the multiperiod setting for AL portfolios by studying the optimal policies and minimum variance frontiers (MVF) implied by discrete-time multiperiod AL portfolio selection models.
- Give a geometric decomposition of these problems that drastically simplifies the analysis and numerical implementation of such model settings.

## Specific: Investigate impact of

- taking liabilities into account
- investment horizon,
- rebalancing frequency,
- exogeneity/endogeneity of liabilities,
- determination of optimal funding ratios
- numerically efficient incorporation of optimization constraints.

# Challenges

- Discrete-time multiperiod mean variance model only recently solved for the asset-only case (Li and Ng (2000)).
- AL surplus optimization requires working with two state variables: the aggregate value of (i) assets and (ii) liabilities.
- Treating liability as exogenous or endogenous.
- Economic interpretation of the structure of the implied solutions and MVF in a multi-period setting.
- Inclusion of intertemporal constraints of different forms (for instance, downside risk limitations).

## Solution

- Discrete-time multiperiod mean variance model: we embed the AL-MV problem in an equivalent mean-second moments problem accessible to dynamics programming.
- State variables for assets and liabilities: we adopt a geometric approach that drastically simplifies the computations.
- Economic interpretation in **exogenous case**: an orthogonal geometric decomposition represents the implied policies (MVF) as linear combinations of some simple basis strategies (returns)
- Economic interpretation in **endogenous case**: a non-trivial geometric decomposition represents the implied policies (MVF) as linear combinations of some basis strategies (returns).

## Contribution to Existing Literature

- Exogenous Liabilities:
  - General recursive analytical representation of the implied policies and MVF (extending results in Li and Ng (2000)).
  - Orthogonal representation of  $k$ -period policies and MVF as a linear combination of  $k+1$  basis policies and returns (extending results in Li and Ng (2000) and Hansen and Richard (1987)).
  - Inclusion of liabilities only affects the multi period minimum second moment (MSM) return (extending results in Keel and Müller (1995))
  - In the i.i.d. setting: closed form solutions
- Endogenous Liabilities:
  - ...to our knowledge, no related work!

## Notation

- For given aggregate initial wealth  $x_0$  and initial liabilities  $l_0$  the investor is allowed to rebalance portfolios at dates  $0, 1, \dots, T - 1$ .
- There are two assets and two liabilities with gross returns  $R_t = (R_t^0, \tilde{R}_t, Q_t^0, \tilde{Q}_t)'$  with aggregate dynamics given by

$$x_{t+1} = R_t^0 x_t + R_t^1 u_t, \quad l_{t+1} = Q_t^0 l_t + Q_t^1 v_t \quad , \quad (1)$$

where  $R_t^1 = \tilde{R}_t - R_t^0$ ,  $Q_t^1 = \tilde{Q}_t - Q_t^0$ .

- The initial balance sheet to be optimized takes the form as given below.

Balance Sheet at time $t$			
Benchmark Asset:	$x_t - u_t$	Benchmark Liability:	$l_t - v_t$
Other Assets:	$u_t$	Other Liabilities:	$v_t$
Total:	$x_t$	Surplus:	$S_t$

## Problem Reformulation

- Define a set  $\mathcal{A}(z_t)$ ,  $z_t = (x_t, l_t)'$ , of linear constraints. This set includes, e.g., short-selling restrictions, positivity constraints on the surplus etc.
- For a final surplus  $S_T := x_T - l_T$  and risk aversion  $w > 0$  the following AL problem has to be solved

$$(P) \quad \begin{cases} \max_{u,v \in \mathcal{A}(z)} [\mathbb{E}(S_T) - w \text{var}(S_T)] \\ \text{s.t. (1)} \end{cases}$$

- For a free parameter  $\lambda$  set  $\gamma = \lambda/w$  and define the optimization problem

$$(P1) \quad \begin{cases} \max_{u,v \in \mathcal{A}(z)} [\mathbb{E}(\gamma S_T - S_T^2)] \\ \text{s.t. (1)} \end{cases}$$

- If  $\phi^*$  is a solution to (P1) for given  $(\lambda^*, w)$ , then it is also a solution to (P) for:

$$\lambda^* = 1 + 2w\mathbb{E}(S_T)|_{\phi^*} .$$

- The implied MVF are the same and problem (P1) is accessible to dynamic programming



## General Solution

Let us first state the main result in the most general form:

**Theorem 1.** *Given a mean-variance AL optimization problem in (P1) given the AL dynamics in (1) the optimal final surplus  $S_T^*$  can be decomposed into two returns,*

$$S_T^* = S_{T-k}^{0e} + \gamma(z)S_{T-k}^{1e}, \quad k = 0, \dots, T - 1,$$

*with  $S_{T-k}^{0e} = \mathbf{1}'D_{T-k}^e(z)z_{T-k}$  and  $S_{T-k}^{1e} = \sum_{i=0}^{k-1} R_{T-k+i}^{1e}(z)$ ,  $D_{T-k}^e$  and  $S_{T-k}^{1e}$  are projections upon properly defined vector spaces.*

Before clarifying the above theorem and presenting the explicit expressions for the projections, we remark:

- The surplus can be decomposed into two returns - their features to be explored!
- Under the above representation, the risk-aversion parameter  $\gamma$  is becoming state-dependent for  $k < T - 1$ .

## Definitions

Before continuing, we have to introduce some definitions:

**Definition 1.** Let  $\mathbb{P} : L_2 \rightarrow L_2$  be an orthogonal projection, that is  $\mathbb{P}$  is self-adjoint and  $\mathbb{P}^2 = \mathbb{P}$ . For any finite dimensional subspace  $\mathcal{S} \subset L_2$ ,  $\mathcal{S}^\perp$  is the orthogonal complement of  $\mathcal{S}$ . By  $\langle \cdot, \cdot \rangle$  we mean the scalar product in the space  $L_2$  of square integrable random variables.

**Definition 2.** A final surplus  $S_T^*$  belongs to the Minimum Variance Frontier (MVF) if the variance of the surplus is a minimum given a targeted expected return of the final surplus.

Further, we denote the vector of assets and liabilities by  $z_t = (x_t, l_t)'$ , and by  $d_t = (u_t, v_t)'$  the vector of amounts invested in the returns  $R_t^1$  and  $Q_t^1$ .

**Definition 3.** We define a matrix  $M_z^y(x)$  and its first leading principle submatrix  $M^y(x)$  as:

$$\begin{array}{c}
 \begin{array}{|ccc|}
 \hline
 1 & 0 & 0 \\
 \hline
 0 & \langle x, y \rangle & \langle y, z \rangle \\
 \hline
 0 & \langle x, z \rangle & \langle z, z \rangle \\
 \hline
 \end{array} \\
 \begin{array}{l}
 M^y(x) \nearrow \\
 \\
 \nearrow M_z^y(x)
 \end{array}
 \end{array}$$

With the above definition the determinant  $|M_z^y(x)|$  equals the determinant of the second principle submatrix. Recall that the determinant is a multilinear skewsymmetric form. Therefore, for two scalars  $\alpha$  and  $\beta$  we would have

$$|M_z^y(\alpha x_1 + \beta x_2)| = \alpha |M_z^y(x_1)| + \beta |M_z^y(x_2)|.$$

To abbreviate notation, we write  $|M_z^y|$  for  $|M_z^y(y)|$  and  $|M_y^z(z)|$  as they are both

the same. Finally, we will work with the following notation for  $t = 0, \dots, T$ :

$$\begin{aligned}
 D_t &= \begin{pmatrix} R_t^0 & 0 \\ 0 & Q_t^0 \end{pmatrix}, \quad G_t = \begin{pmatrix} R_t^1 & 0 \\ 0 & Q_t^1 \end{pmatrix}, \\
 \tilde{D}_t &= \begin{pmatrix} R_t^0 & 0 \\ 0 & \tilde{Q}_t \end{pmatrix}, \quad \hat{D}_t = \begin{pmatrix} \hat{R}_t & 0 \\ 0 & Q_t^0 \end{pmatrix} \\
 \mathbf{I} &= \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \quad e_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \quad e_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}.
 \end{aligned}$$

# One-Period Optimization

Consider none of the constraints to be binding and an investor who...

- a) ...totally neglects existence of liabilities
- b) ...assumes liabilities to be exogenously determined
- c) ...assumes endogenous liabilities.

In case a) the investor solves

$$(P2-a) \quad \begin{cases} \max_u \mathbb{E}[\gamma x_T - \frac{1}{2}x_T^2] \\ s.t. \quad x_{t+1} = R_t^0 x_t + R_t^1 u_t, \quad t = T - 1. \end{cases}$$

The investor with exogenous liabilities, case b), solves the problem,

$$(P2-b) \quad \begin{cases} \max_u \mathbb{E}[\gamma \mathbf{I}' z_T - \frac{1}{2} z_T' \mathbf{I} \mathbf{I}' z_T] \\ s.t. \quad \mathbf{I}' z_{t+1} = \mathbf{I}' D_t z_t + \mathbf{I}' G_t e_1 u_t, \quad t = T - 1, \end{cases}$$

whereas the investor with the endogenous liabilities, case c), solves

$$(P2-c) \quad \begin{cases} \max_d \mathbb{E}[\gamma \mathbf{I}' z_T - \frac{1}{2} z_T' \mathbf{I} \mathbf{I}' z_T] \\ s.t. \quad \mathbf{I}' z_{t+1} = \mathbf{I}' D_t z_t + \mathbf{I}' G_t d_t, \quad t = T - 1. \end{cases}$$

## A. The Solutions

According to [Theorem 1](#), the MVF-surplus for all problems is given by

$$S_T^* = S_{T-1}^{0j} + \gamma S_{T-1}^{1j} = \mathbf{I}' D_{T-1}^j z_{T-1} + \gamma R_{T-1}^{1j}, \quad j \in \{a, b, c\}.$$

Depending on the investor's assumption about liabilities, we get for case [c](#))

$$D^c = \begin{pmatrix} \mathbb{P}_{R^{1\perp}Q^1}^{\mathcal{D}_{QR}\mathcal{D}_{RR}}(R^0) & 0 \\ 0 & \mathbb{P}_{R^1Q^{1\perp}}^{\mathcal{D}_{QQ}\mathcal{D}_{RQ}}(Q^0) \end{pmatrix},$$

$$R^{1c} = \mathbb{P}_{R^1Q^1}^{\mathcal{D}_{QI}\mathcal{D}_{RI}}(\mathbb{1}).$$

For case [b](#))

$$D^b = \begin{pmatrix} \mathbb{P}_{R^{1\perp}}(R^0) & 0 \\ 0 & \mathbb{P}_{R^{1\perp}}(Q^0) \end{pmatrix},$$

$$R^{1b} = \mathbb{P}_{R^1}(\mathbb{1}).$$

Finally, for case [a](#))

$$D^a = \begin{pmatrix} \mathbb{P}_{R^{1\perp}}(R^0) & 0 \\ 0 & Q^0 \end{pmatrix},$$

$$R^{1a} = R^{1b} = \mathbb{P}_{R^1}(\mathbb{1}).$$

## B. Asset-Only vs. Exogenous Liabilities

The results for cases a) and b) seem to be interpretable in a straightforward manner. The formalism using projections was introduced in Leippold, Trojani, and Vanini (2002a). In particular,

- the projections are explicitly given as

$$\mathbb{P}_{R^{1\perp}}(R^0) = R_{T-1}^0 - \frac{\mathbb{E}(R_{T-1}^1 R_{T-1}^0)}{\mathbb{E}\left((R_{T-1}^1)^2\right)} R_{T-1}^1,$$

$$\mathbb{P}_{R^{1\perp}}(Q^0) = Q_{T-1}^0 - \frac{\mathbb{E}(R_{T-1}^1 Q_{T-1}^0)}{\mathbb{E}\left((R_{T-1}^1)^2\right)} R_{T-1}^1,$$

$$\mathbb{P}_{R_{T-1}^1}(\mathbb{1}) = \frac{\mathbb{E}(R_{T-1}^1)}{\mathbb{E}\left((R_{T-1}^1)^2\right)} R_{T-1}^1.$$

- $S_T^*$  is a linear combination of two returns:

$$S_T^* = \begin{cases} x_{T-1} \mathbb{P}_{R_{T-1}^{1,\perp}}(R^0) - l_{T-1} Q^0 & + \gamma \mathbb{P}_{R_{T-1}^1}(\mathbb{1}), & \text{for case a)} \\ x_{T-1} \mathbb{P}_{R_{T-1}^{1,\perp}}(R^0) - l_{T-1} \mathbb{P}_{R_{T-1}^{1,\perp}}(Q^0) & + \gamma \mathbb{P}_{R_{T-1}^1}(\mathbb{1}), & \text{for case b)} \end{cases}$$

- There is a subtle difference between the surplus for case a) and b):
  - $S_{T-1}^{0b}$  is the minimum-second-moment (MSM) surplus, the difference of an asset-only and a liabilities-only MSM payoff.
  - $S_{T-1}^{0a}$  is the minimum-second-moment (MSM) surplus only if  $l_{T-1} = 0$  or  $Q^0 = 0$ .
  - $S_{T-1}^{1j} = \mathbb{P}_{R_{T-1}^1}(\mathbb{1})$ ,  $j \in \{a, b\}$ , is the asset excess return nearest to the “risk-free” pay-off  $\mathbb{1}$ .
  - The MSM surplus is the return of a portfolio with generally non zero initial position in AL while  $\mathbb{P}_{R_{T-1}^1}(\mathbb{1})$  is an asset excess return
- With exogenous liability (case b)), the surplus is decomposed into two orthogonal returns, such that

$$\begin{aligned}\mathbb{E} [S_{T-1}^{0b} S_{T-1}^{1b}] &= 0, \\ \mathbb{E} [(S_{T-1}^{1b})^2] &= \mathbb{E} [S_{T-1}^{1b}].\end{aligned}$$

- In the asset-only case, only the second equation above holds, i.e.,

$$\mathbb{E} [(S_{T-1}^{1a})^2] = \mathbb{E} [S_{T-1}^{1a}],$$

but  $\mathbb{E} [S_{T-1}^{0a} S_{T-1}^{1a}] \neq 0$ . These differences have some impacts on the MVF.



## C. Mean-Variance-Frontier I

Any MVF can be represented in the  $(\mathbb{V}(S), \mathbb{E}(S))$ -space by a curvature parameter  $A$ , by a horizontal shift parameter  $B$  and a vertical shift parameter  $C$ . In particular, the one-period MVF in the unrestricted case can be represented as

$$\mathbb{V}(S_T) = A_j \mathbb{E}(S_T)^2 + 2B_j \mathbb{E}(S_T) + C_j, \quad j \in \{a, b, c\}.$$

The Minimum Variance Portfolio (MVP) has expectation value and variance according to

$$\mathbb{E}(S^{MVP}) = -B_j/A_j, \quad \mathbb{V}(S^{MVP}) = -B_j^2/4A_j + C_j, \quad j \in \{a, b, c\}.$$

For the MVF's of a) and b) the following holds:

$$\begin{aligned} A_b &= \frac{1}{\mathbb{E}[S^{1b}]} - 1, & A_a &= \frac{1}{\mathbb{E}[S^{1a}]} - 1, \\ B_b &= -\frac{\mathbb{E}[S^{0b}]}{\mathbb{E}[S^{1b}]}, & B_a &= \frac{\mathbb{E}[S^{0a}S^{1a}]}{\mathbb{E}[S^{1a}]} - \frac{\mathbb{E}[S^{0a}]}{\mathbb{E}[S^{1a}]}, \\ C_b &= \mathbb{E}\left[(S^{0b})^2\right] + \frac{\mathbb{E}[S^{0b}]^2}{\mathbb{E}[S^{1b}]}, & C_a &= \mathbb{E}\left[(S^{0a})^2\right] - \frac{2\mathbb{E}[S^{0a}]\mathbb{E}[S^{0a}S^{1a}]}{\mathbb{E}[S^{1a}]} + \frac{\mathbb{E}[S^{0a}]^2}{\mathbb{E}[S^{1a}]}. \end{aligned}$$

From the expression above, the following implications are derived

- The AL MVF is affected by the introduction of liabilities in only two ways (cf. also Keel and Müller (1995)):
  - By a “vertical” shift caused by the parameter  $C$  and a “sidewise” shift caused by the parameter  $B$ .
  - Not by a change of curvature in the parameter  $A$ .
- These shifts are caused by the orthogonality property of the two surplus returns in the exogenous liability case.
- This induces a pure translation of the MVF in the mean-variance space, caused by a lower global MSM surplus  $S_{T-1}^{0j}$ ,  $j \in \{a, b\}$ .
- The direction of the translation is north-west in the  $(\mathbb{V}, \mathbb{E})$ -plane and can be computed explicitly.

## Mean-Variance-Frontier: Asset-Only vs Exogeneous Liabilities

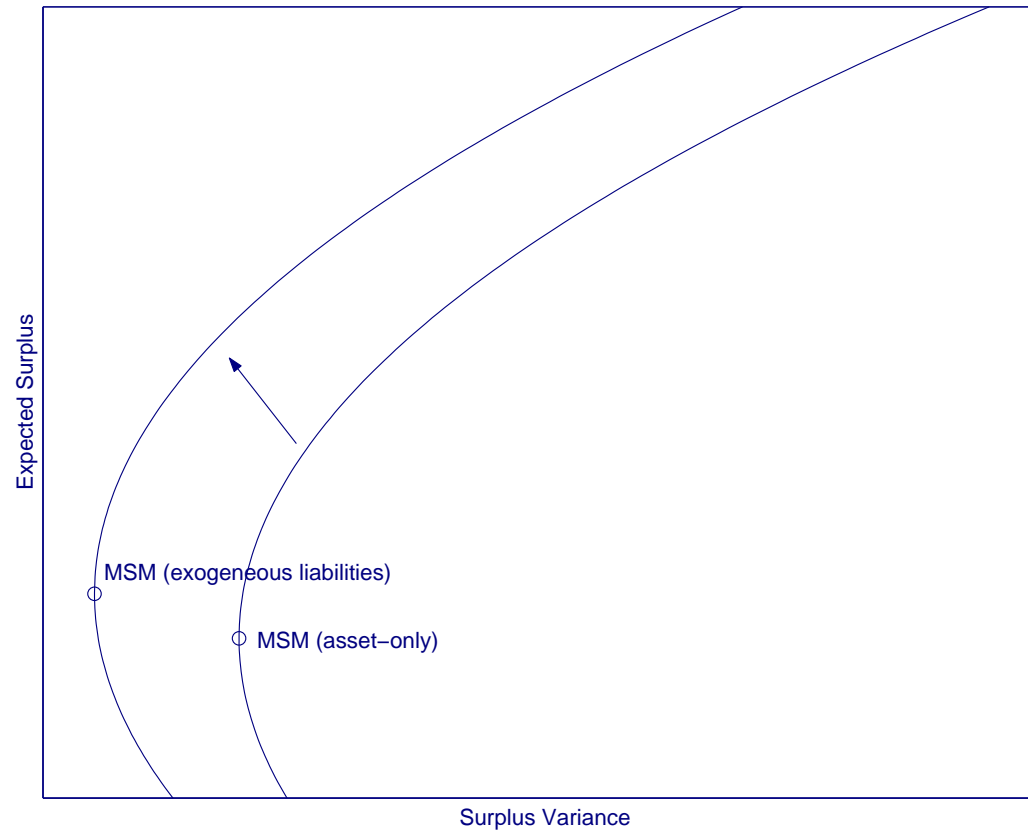


Figure 1: Mean-Variance-Frontiers. Moving from an asset-only optimization to a optimization where exogeneous liabilities are taken into account, shifts the MVF in the upper-right corner of the  $(\mathbb{V}(S), \mathbb{E}(S))$ -plane.

## D. Endogenous Liabilities

Writing out explicitly the terms for  $D^c$  and  $R^{1c}$ , we arrive at

$$D^c = \begin{pmatrix} R^0 - \frac{\langle R^0, Q^1 \rangle \langle R^1, Q^1 \rangle - \langle R^0, R^1 \rangle \langle Q^1, Q^1 \rangle}{\langle R^1, Q^1 \rangle^2 - \langle R^1, R^1 \rangle \langle Q^1, Q^1 \rangle} R^1 & \frac{\langle Q^0, Q^1 \rangle \langle R^1, Q^1 \rangle - \langle Q^1, Q^1 \rangle \langle R^1, Q^0 \rangle}{\langle R^1, Q^1 \rangle^2 - \langle R^1, R^1 \rangle \langle Q^1, Q^1 \rangle} R^1 \\ \frac{\langle R^0, R^1 \rangle \langle R^1, Q^1 \rangle - \langle R^1, R^1 \rangle \langle R^0, Q^1 \rangle}{\langle R^1, Q^1 \rangle^2 - \langle R^1, R^1 \rangle \langle Q^1, Q^1 \rangle} Q^1 & Q^0 - \frac{\langle R^1, Q^0 \rangle \langle R^1, Q^1 \rangle - \langle R^1, R^1 \rangle \langle Q^0, Q^1 \rangle}{\langle R^1, Q^1 \rangle^2 - \langle R^1, R^1 \rangle \langle Q^1, Q^1 \rangle} Q^1 \end{pmatrix},$$

and

$$R^{1c} = \frac{(\langle Q^1, \mathbb{1} \rangle \langle R^1, Q^1 \rangle - \langle R^1, \mathbb{1} \rangle \langle Q^1, Q^1 \rangle) R^1 + (\langle R^1, \mathbb{1} \rangle \langle R^1, Q^1 \rangle - \langle Q^1, \mathbb{1} \rangle \langle R^1, R^1 \rangle) Q^1}{\langle R^1, Q^1 \rangle^2 - \langle R^1, R^1 \rangle \langle Q^1, Q^1 \rangle}.$$

The above expressions are rather involved and there is now way to detect the mathematical structure of the problem. Luckily, using the definition of  $M$ , we can simplify

$$D^c = \begin{pmatrix} R^0 - \frac{|M_{Q^1}^{R^1}(R^0)|}{|M_{Q^1}^{R^1}|} R^1 - \frac{|M_{R^1}^{Q^1}(R^0)|}{|M_{Q^1}^{R^1}|} Q^1 & 0 \\ 0 & Q^0 - \frac{|M_{Q^1}^{R^1}(Q^0)|}{|M_{Q^1}^{R^1}|} Q^1 - \frac{|M_{R^1}^{Q^1}(Q^0)|}{|M_{Q^1}^{R^1}|} R^1 \end{pmatrix},$$

and

$$R^{1c} = \frac{|M_{Q^1}^{R^1}(\mathbb{1})|}{|M_{Q^1}^{R^1}|} R^1 + \frac{|M_{R^1}^{Q^1}(\mathbb{1})|}{|M_{Q^1}^{R^1}|} Q^1.$$

To simplify the notation, we define<sup>1</sup>

$$\begin{aligned} \mathcal{D}_{QR} &= \frac{|M_{Q^1}^{R^1}(R^0)|}{|M_{Q^1}^{R^1}|}, & \mathcal{D}_{RR} &= \frac{|M_{R^1}^{Q^1}(R^0)|}{|M_{Q^1}^{R^1}|}, & \mathcal{D}_{QQ} &= \frac{|M_{Q^1}^{R^1}(Q^0)|}{|M_{Q^1}^{R^1}|}, \\ \mathcal{D}_{RQ} &= \frac{|M_{R^1}^{Q^1}(Q^0)|}{|M_{Q^1}^{R^1}|}, & \mathcal{D}_{Q\mathbb{1}} &= \frac{|M_{Q^1}^{R^1}(\mathbb{1})|}{|M_{Q^1}^{R^1}|}, & \mathcal{D}_{R\mathbb{1}} &= \frac{|M_{R^1}^{Q^1}(\mathbb{1})|}{|M_{Q^1}^{R^1}|}. \end{aligned}$$

From the positive definiteness of the expected return matrix,  $\mathcal{D} \in (0, \infty)$ .

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<sup>1</sup>The first subscript of  $\mathcal{D}$  determines the subscript of  $M$  in the nominator. The second subscript of  $\mathcal{D}$  enters as the argument in the parenthesis of  $M$  in the nominator and superscript of  $M$  can just be deduced from the subscript of  $\mathcal{D}$ .

**Lemma 1.** *The expressions*

$$\begin{aligned}
& R^0 - \frac{|M_{Q^1}^{R^1}(R^0)|}{|M_{Q^1}^{R^1}|} R^1, \quad \frac{|M_{R^1}^{Q^1}(R^0)|}{|M_{Q^1}^{R^1}|} Q^1, \quad Q^0 - \frac{|M_{Q^1}^{R^1}(Q^0)|}{|M_{Q^1}^{R^1}|} Q^1, \\
& \frac{|M_{R^1}^{Q^1}(Q^0)|}{|M_{Q^1}^{R^1}|} R^1, \quad \frac{|M_{Q^1}^{R^1}(\mathbb{1})|}{|M_{Q^1}^{R^1}|} R^1, \quad \frac{|M_{R^1}^{Q^1}(\mathbb{1})|}{|M_{Q^1}^{R^1}|} Q^1,
\end{aligned}$$

*are all (orthogonal) projections.*

*Proof.* The lemma follows from the multilinearity of the determinant.  $\square$

Hence,

$$D^c = \begin{pmatrix} \mathbb{P}_{R^1 \perp}^{\mathcal{D}_{QR}}(R^0) - \mathbb{P}_{Q^1}^{\mathcal{D}_{RR}}(R^0) & 0 \\ 0 & \mathbb{P}_{Q^1 \perp}^{\mathcal{D}_{QQ}}(Q^0) - \mathbb{P}_{R^1}^{\mathcal{D}_{RQ}}(Q^0) \end{pmatrix},$$

and

$$R^{1c} = \mathbb{P}_{R^1}^{\mathcal{D}_{Q^1}}(\mathbb{1}) + \mathbb{P}_{Q^1}^{\mathcal{D}_{R^1}}(\mathbb{1}).$$

Going one step further we can represent the diagonal elements of the matrix

$D^c$  as projections on the span of  $(R^{1\perp}, Q^1)$  and  $(R^1, Q^{1\perp})$ , i.e.,

$$D^c = \begin{pmatrix} \mathbb{P}_{R^{1\perp}Q^1}^{\mathcal{D}_{QR}\mathcal{D}_{RR}}(R^0) & 0 \\ 0 & \mathbb{P}_{R^1Q^{1\perp}}^{\mathcal{D}_{QQ}\mathcal{D}_{RQ}}(Q^0) \end{pmatrix},$$

Similarly,  $R^{1c}$  is a projection of  $\mathbf{l}$  on the span of  $(R^1, Q^1)$ ,

$$R^{1c} = \mathbb{P}_{R^1Q^1}^{\mathcal{D}_{Ql}\mathcal{D}_{Rl}}(\mathbf{l}).$$

Note that the orthogonality properties will simplify the subsequent optimization procedure for the subsequent time steps. Indeed, tedious calculations show that

$$\mathbb{P}_{R^{1\perp}Q^1}^{\mathcal{D}_{QR}\mathcal{D}_{RR}} \left( \mathbb{P}_{R^1Q^1}^{\mathcal{D}_{Ql}\mathcal{D}_{Rl}}(\mathbf{l}) \right) = \mathbb{P}_{R^1Q^{1\perp}}^{\mathcal{D}_{QQ}\mathcal{D}_{RQ}} \left( \mathbb{P}_{R^1Q^1}^{\mathcal{D}_{Ql}\mathcal{D}_{Rl}}(\mathbf{l}) \right) = 0,$$

and

$$\mathbb{P}_{R^{1\perp}Q^1}^{\mathcal{D}_{QR}\mathcal{D}_{RR}} \left( \mathbb{P}_{R^1Q^{1\perp}}^{\mathcal{D}_{QQ}\mathcal{D}_{RQ}}(Q^0) \right) = \mathbb{P}_{R^1Q^{1\perp}}^{\mathcal{D}_{QQ}\mathcal{D}_{RQ}} \left( \mathbb{P}_{R^{1\perp}Q^1}^{\mathcal{D}_{QR}\mathcal{D}_{RR}}(R^0) \right) = 0.$$

Whenever a restriction is binding, the projection changes, but the basic structure of the optimal surplus as the sum of two returns is preserved. Indeed, whenever a constraint in  $u$  and  $v$  is binding, the optimization problem is equivalent to an optimization where either the assets or the liabilities are “exogenized”.

## E. The Mean-Variance-Frontier with Endogenous Liabilities

With endogenous liabilities, the parameters of the MVF are given by

$$\begin{aligned}A_c &= \frac{1}{\mathbb{E}[S^{1c}]} - 1, \\B_c &= \frac{\mathbb{E}[S^{0c}S^{1c}]}{\mathbb{E}[S^{1c}]} - \frac{\mathbb{E}[S^{0c}]}{\mathbb{E}[S^{1c}]}, \\C_c &= \mathbb{E}[(S^{0c})^2] - \frac{2\mathbb{E}[S^{0c}]\mathbb{E}[S^{0c}S^{1c}]}{\mathbb{E}[S^{1c}]} + \frac{\mathbb{E}[S^{0c}]^2}{\mathbb{E}[S^{1c}]}.\end{aligned}$$

Besides that  $S^{0c}$  is no longer the MSM portfolio, the following holds:

**Proposition 1.** *Endogenous liabilities influence both shift parameters as well as the convexity parameter of the MVF. Exogenizing or neglecting liabilities in the optimization program makes the convexity independent of liabilities.*



We can even go one step further:

**Proposition 2.** *The risk-return tradeoff<sup>2</sup> on the MVF with exogenous liabilities is always less or equal as the risk-return tradeoff on the MVF with endogenous liabilities. The MVFs for b) and c) collide as  $Q_1 \rightarrow 0$ .*

*Proof.* We have to show that the curvature of the MVF with endogenous liabilities is always less or equal than the curvature with exogenous liabilities, i.e.,

$$\frac{1}{\mathbb{E} [\mathbb{P}_{R^1}(\mathbf{1})]} \geq \frac{\mathbb{E} \left[ \left( \mathbb{P}_{R^1 Q^1}^{\mathcal{D}_{Q^1} \mathcal{D}_{R^1}}(\mathbf{1}) \right)^2 \right]}{\mathbb{E} \left[ \mathbb{P}_{R^1 Q^1}^{\mathcal{D}_{Q^1} \mathcal{D}_{R^1}}(\mathbf{1}) \right]^2}.$$

Note, from the positive definiteness assumption of the return matrix, all determinants in the above expression are strictly positive. After some algebraic manipulation, the above inequality can be considerably simplified to

$$\frac{\langle R^1, R^1 \rangle}{\langle \mathbf{1}, R^1 \rangle^2} \geq \frac{\langle R^1, R^1 \rangle \langle Q^1, Q^1 \rangle - \langle R^1, Q^1 \rangle^2}{\langle \mathbf{1}, R^1 \rangle^2 \langle Q^1, Q^1 \rangle - 2 \langle \mathbf{1}, R^1 \rangle \langle \mathbf{1}, Q^1 \rangle \langle R^1, Q^1 \rangle + \langle R^1, R^1 \rangle \langle \mathbf{1}, Q^1 \rangle^2}.$$

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<sup>2</sup>As measured by the Sharpe-Ratio

We now multiply the above equation with  $\langle \mathbf{1}, R^1 \rangle$  and write the RHS as

$$\langle R^1, R^1 \rangle \frac{\langle Q^1, Q^1 \rangle - \frac{\langle R^1, Q^1 \rangle^2}{\langle R^1, R^1 \rangle}}{\langle Q^1, Q^1 \rangle - 2 \frac{\langle \mathbf{1}, Q^1 \rangle \langle R^1, Q^1 \rangle}{\langle \mathbf{1}, R^1 \rangle} + \frac{\langle R^1, R^1 \rangle \langle \mathbf{1}, Q^1 \rangle^2}{\langle \mathbf{1}, R^1 \rangle^2}}.$$

Obviously, we have proven our result, when we can show that

$$\frac{\langle R^1, Q^1 \rangle^2}{\langle R^1, R^1 \rangle} \geq 2 \frac{\langle \mathbf{1}, Q^1 \rangle \langle R^1, Q^1 \rangle}{\langle \mathbf{1}, R^1 \rangle} - \frac{\langle R^1, R^1 \rangle \langle \mathbf{1}, Q^1 \rangle^2}{\langle \mathbf{1}, R^1 \rangle^2}$$

We can write the latter inequality as

$$\langle R^1, Q^1 \rangle^2 \langle \mathbf{1}, R^1 \rangle^2 \geq 2 \langle \mathbf{1}, Q^1 \rangle \langle R^1, Q^1 \rangle \langle \mathbf{1}, R^1 \rangle \langle R^1, R^1 \rangle - \langle \mathbf{1}, Q^1 \rangle \langle R^1, R^1 \rangle^2.$$

This, in turn, is just the same as

$$(\langle R^1, Q^1 \rangle \langle \mathbf{1}, R^1 \rangle - \langle \mathbf{1}, Q^1 \rangle \langle R^1, R^1 \rangle)^2 \geq 0,$$

and we are done. □

# Mean-Variance-Frontiers: Exogenous vs. Endogenous Liabilities

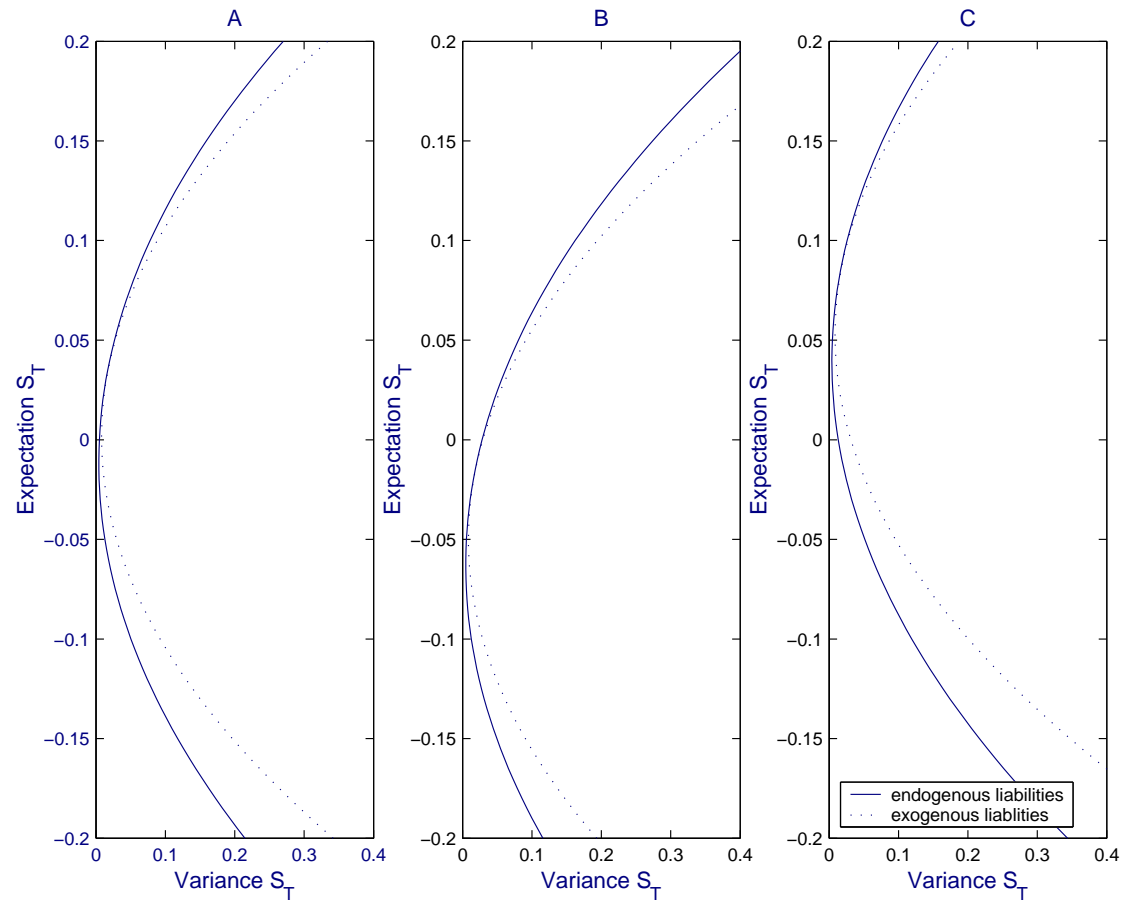


Figure 2: **One-Period MVFs. The curvature of the MVF with endogenous liabilities is smaller than the curvature of the MVF with endogenous liabilities.**

## Multiperiod Model - Exogenous Liabilities

The following results are obtained:

- MSM returns

$$\begin{aligned}R_{T-k}^{0e} &= \mathbb{P}_{(R_{T-k}^1 R_{T-k+1}^{0e})^\perp} (R_{T-k}^0 R_{T-k+1}^{0e}) \\Q_{T-k}^{0e} &= \mathbb{P}_{(R_{T-k}^1 R_{T-k+1}^{0e})^\perp} (Q_{T-k}^0 Q_{T-k+1}^{0e}).\end{aligned}$$

- Optimal surplus

$$S_T = x_{T-k} R_{T-k}^{0e} - l_{T-k} Q_{T-k}^{0e} + \gamma \sum_{i=0}^{k-1} R_{T-k+i}^{1e},$$

where  $R_{T-k+i}^{1e} = \mathbb{P}_{(R_{T-k+i}^1 R_{T-k+i+1}^{0e})} (\mathbb{1})$ .

## A. Interpretation

- The same orthogonal structure prevails as in the one-period model.
- The set
 
$$\{x_{T-k}R_{T-k}^{0e} - l_{T-k}Q_{T-k}^{0e}, R_{T-k}^{1e}, \dots, R_{T-2}^{1e}, R_{T-1}^{1e}\}$$
 is an orthogonal system that spans the MVF.
- The return  $R_{T-k+i}^{1e}$  can be interpreted for any  $i = 0, \dots, k - 1$  as a "local",  $(k - i)$ -period, excess asset return.
- $S_T$  is the orthogonal sum of a  $k$ -period MSM surplus  $x_{T-k}R_{T-k}^{0e} - l_{T-k}Q_{T-k}^{0e}$  and  $k$  "local" excess asset returns  $R_{T-k}^{1e}, \dots, R_{T-1}^{1e}$ .
- Without any further assumptions about return dynamics, the optimal surplus has to be solved recursively.

## B. Minimum Variance Frontier

- Defining  $S_{T-k}^0 = x_{T-k}R_{T-k}^{0e} - l_{T-k}Q_{T-k}^{0e}$ ,  $S_{T-k}^1 = \sum_{i=0}^{k-1} R_{T-k+i}^{1e}$  the  $k$ -period MVF is given by:

$$\begin{aligned} \mathbb{V}(S_T) &= A_{T-k} \cdot [\mathbb{E}(S_T)]^2 + 2B_{T-k} \cdot \mathbb{E}(S_T) + C_{T-k} \\ A_{T-k} &= \frac{1}{\mathbb{E}(S_{T-k}^1)} - 1, \quad B_{T-k} = \frac{\mathbb{E}(S_{T-k}^0)}{\mathbb{E}(S_{T-k}^1)}, \\ C_{T-k} &= \frac{[\mathbb{E}(S_{T-k}^0)]^2}{\mathbb{E}(S_{T-k}^1)} + \mathbb{E}((S_{T-k}^0)^2) \end{aligned}$$

- While  $B_{T-k}$  and  $C_{T-k}$  depend on  $(Q_{T-k}^0)_{k=1,\dots,T}$ ,  $A_{T-k}$  does not! (as in the one period model).
- This induces a pure translation of the MVF in the mean-variance space, caused by a lower global MSM surplus  $S_{T-k}^0$

## C. Results for the iid Case

- The basis returns are given in closed form, for instance:
  - Assets only MSM returns

$$R_{T-k}^{0e} = \prod_{i=0}^{k-1} \mathbb{P}_{R_{T-k+i}^{1,\perp}} (R^0),$$

- Local asset excess returns:

$$R_{T-k+i}^{1e} = \mathbb{P}_{R_{T-k+i}^1} (\mathbf{1}) \prod_{j=i+1}^{k-1} \mathbb{P}_{\mathbb{P}_{R_{T-k+j}^{1,\perp}}} (R^0) (\mathbf{1})$$

- The closed form expression for  $Q_{T-k}^{0e}$  is given by:

$$Q_{T-k}^{0e} = \prod_{i=0}^{k-1} Q_{T-k+i}^0 - \sum_{i=0}^{k-1} \left( \mathbb{P}_{R_{T-k+i}^1} (Q^0) \prod_{j=i+1}^{k-1} \mathbb{P}_{\mathbb{P}_{R_{T-k+j}^{1,\perp}}} (R^0) (Q^0) \prod_{j=0}^{i-1} Q_{T-k+j}^0 \right).$$

## D. Myopic vs Dynamic with Exogenous Liabilities

**Question 1:** Given a fixed time horizon, what is the impact of the rebalancing frequency of the portfolio?

**Question 2:** How far off can a myopic investor be, compared to an investor who rebalances his portfolio?

**Answer 1:** Comparing the myopic strategy with the dynamic strategy, we can prove the following:

- The curvature of the MVF is always smaller in the dynamic case. This can be proven by using Jensen's inequality for the expectation operator.
- The vertical shift parameter of the dynamic MVF is always bigger in absolute terms.
- The horizontal shift parameter of the dynamic curve is always smaller.
- The parameters of the frontier converge to a finite value as  $h \rightarrow \infty$ .

This leads to the conclusion, that the MVF of the dynamic investor can only intersect the myopic MVF in the lower, inefficient part of the dynamic MVF.



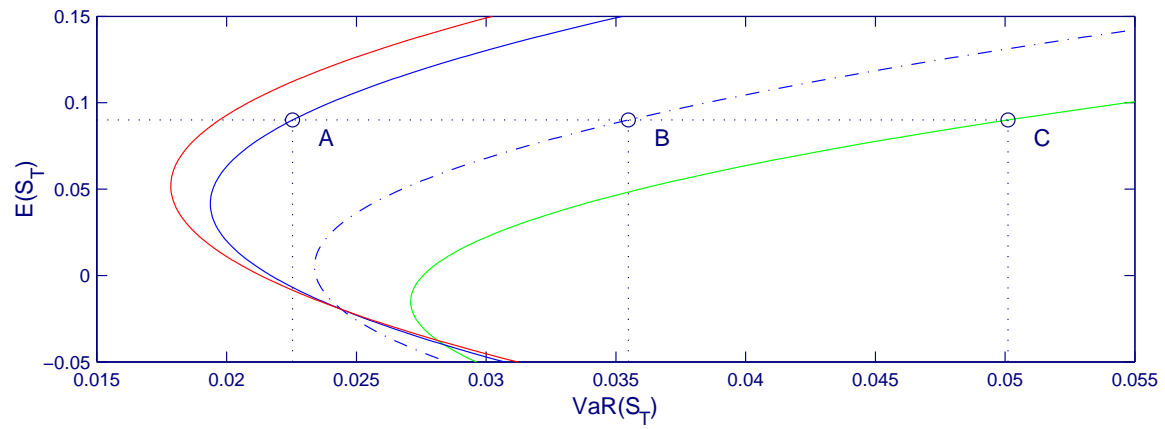
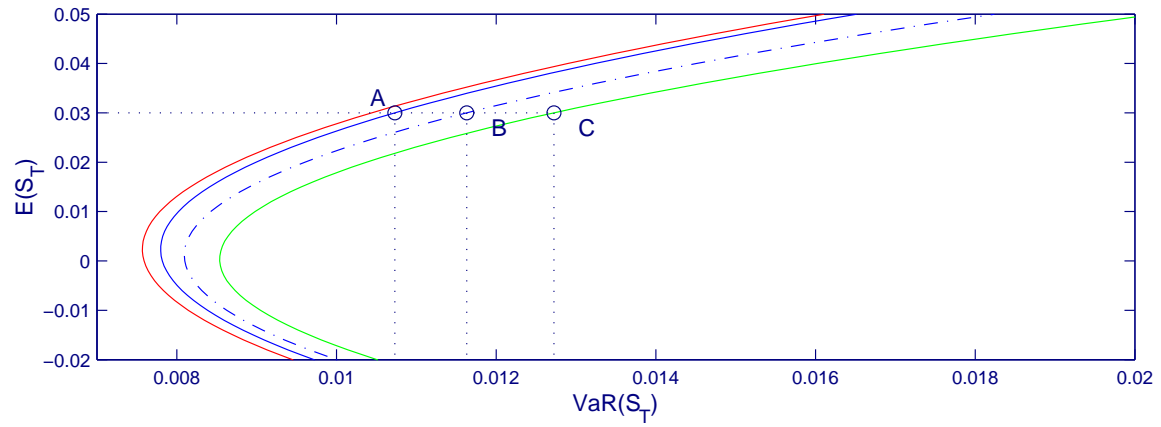


Figure 3: Effects of Time Diversification.

Answer 2: We compare to strategies with a fixed time-horizon of 1 year.

- The first strategy is a myopic one, where the investor does not rebalance his AL portfolio during the whole period.
- The second strategy rebalances the portfolio every month.
- We compare the initial MVFs of both strategies.
- We make a snapshot after 2 month and assume that
  - a) The value of assets to liabilities is 1.1.
  - b) The value of assets to liabilities is 0.95.

The results are plotted in the next figure...

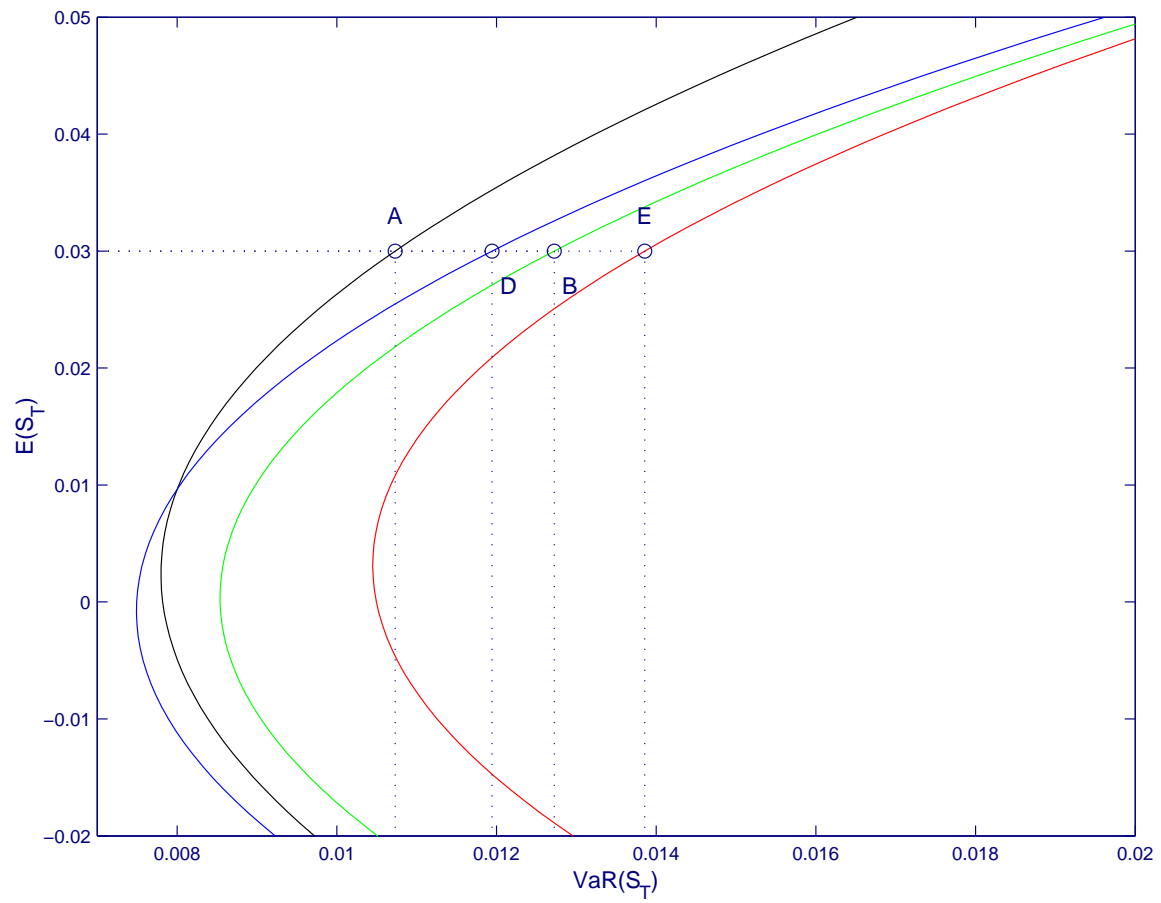


Figure 4: Movement of Dynamic MFVs in Time.

# Multiperiod Model - Endogenous Liabilities

## A. Time $T - 2$

Dropping the term  $\mathcal{C}(z)$ , we can write for  $T - 2$  the unconstrained value function as

$$J(z) \equiv \max_{d \in A(z)} \mathbb{E} \left[ \hat{\gamma}(z_{T-1}) \mathbf{I}'(\hat{D}(z)z + \hat{G}(z)d) - \frac{1}{2}(\hat{D}(z)z + \hat{G}(z)d)' \mathbf{II}'(\hat{D}(z)z + \hat{G}(z)d) \right],$$

where  $\alpha_i \leq 0$  are the Kuhn-Tucker multipliers at  $T - 2$  and

$$\hat{\gamma}(z_{T-1}) = \gamma (1 - \gamma R^{1e}(z_{T-1})).$$

Whenever at time  $T - 1$  a constraint is binding, we can set  $\hat{\gamma}(z_{T-1}) = \gamma$ !

The J-function has the same form as the J-Function at time  $T - 1$ , but with the matrices  $D$  and  $G$  replaced by the state-dependent matrices  $\hat{D}(z)$  and  $\hat{G}(z)$ .

More particularly,

$$\hat{D}(z) \equiv \hat{D}_{T-2}(z) = D_{T-1}^e(z) D_{T-2}, \quad \hat{G}(z) \equiv \hat{G}_{T-2}(z) = D_{T-1}^e(z) G_{T-2}.$$

## B. Time $T - k$

For time  $T - k$ ,  $k = 3, \dots, T$ , we can continue recursively, substituting at each time step  $k$ ,

$$\widehat{D}_{T-k}(z) = D_{T-k+1}^e(z)D_{T-k}, \quad \widehat{G}_{T-k}(z) = D_{T-k+1}^e(z)G_{T-k}.$$

Finally, we can write the optimal surplus belonging to the MVF as

$$\begin{aligned} S_T^* &= \mathbf{I}'D_{T-1}^e(z)z_{T-1} + \gamma R_{T-1}^{1e}, \\ &= \mathbf{I}'D_{T-k}^e(z)z_{T-k} + \gamma \sum_{i=0}^{k-1} R_{T-k+i}^{1e}(z), \end{aligned}$$

where  $D_{T-k}^e$  and  $R_{T-k}^{0e}$  have to be determined recursively. As an example, when no constraints are binding, we have

$$D_{T-k}^e(z) = D_{T-k+1}^e(z)\widetilde{D}_{T-k},$$

where  $\tilde{D}_{T-k}$  is the matrix given by

$$\begin{pmatrix} R_{T-k}^0 - \begin{pmatrix} \frac{|\tilde{M}_{Q^1}^{R^1}(R^0)|}{|\tilde{M}_{Q^1}^{R^1}|} \\ \frac{|\tilde{M}_{R^1}^{Q^1}(R^0)|}{|\tilde{M}_{R^1}^{Q^1}|} \\ \frac{|\tilde{M}_{Q^1}^{R^1}|}{|\tilde{M}_{Q^1}^{R^1}|} \end{pmatrix}' \begin{pmatrix} R_{T-k}^1 \\ Q_{T-k}^1 \end{pmatrix} & 0 \\ 0 & Q_{T-k}^0 - \begin{pmatrix} \frac{|\tilde{M}_{Q^1}^{R^1}(Q^0)|}{|\tilde{M}_{Q^1}^{R^1}|} \\ \frac{|\tilde{M}_{R^1}^{Q^1}(Q^0)|}{|\tilde{M}_{R^1}^{Q^1}|} \\ \frac{|\tilde{M}_{Q^1}^{R^1}|}{|\tilde{M}_{Q^1}^{R^1}|} \end{pmatrix}' \begin{pmatrix} Q_{T-k}^1 \\ R_{T-k}^1 \end{pmatrix} \end{pmatrix}.$$

Here, e.g. the matrix  $\tilde{M}_{Q^1}^{R^1}(R^0)$  is given by

$$\begin{pmatrix} \langle [D_{T-k+1}^e]_{11} R_{T-k}^0, [D_{T-k+1}^e]_{11} R_{T-k}^1 \rangle & \langle [D_{T-k+1}^e]_{11} R_{T-k}^1, [D_{T-k+1}^e]_{22} Q_{T-k}^1 \rangle \\ \langle [D_{T-k+1}^e]_{11} R_{T-k}^0, [D_{T-k+1}^e]_{22} Q_{T-k}^1 \rangle & \langle [D_{T-k+1}^e]_{22} Q_{T-k}^1, [D_{T-k+1}^e]_{22} Q_{T-k}^1 \rangle \end{pmatrix},$$

and  $[D_{T-k+1}^e]_{ii}$  is the  $i$ th diagonal element of  $D_{T-k+1}^e$ . The expression for  $R_{T-k}^{1e}$

is given as

$$R_{T-k}^{1e} = \frac{\left| \widetilde{M}_{R^1}^{Q^1}(\mathbf{1}) \right|}{\left| \widetilde{M}_{Q^1}^{R^1} \right|} [D_{T-k+1}^e]_{11} R_{T-k}^1 + \frac{\left| \widetilde{M}_{Q^1}^{R^1}(\mathbf{1}) \right|}{\left| \widetilde{M}_{Q^1}^{R^1} \right|} [D_{T-k+1}^e]_{22} Q_{T-k}^1.$$

Setting  $S_{T-k}^{0e} \equiv \mathbf{I}' D_{T-k}^e(z) z_{T-k}$  and  $S_{T-k}^{1e} \equiv \sum_{i=0}^{k-1} R_{T-k+i}^{1e}(z)$  finishes the proof of Theorem 1. More details can be found in Leippold, Trojani, and Vanini (2002b)

## Conclusion

- We derived analytical solutions for the optimal policies and the MVF implied by a multiperiod mean-variance optimization of AL portfolios under some additional assumptions.
- We discussed some specific issues related to the impact of the rebalancing frequency and the characteristic of the liabilities, whether they are exogenously or endogenously determined.
- [Work in progress:](#)
  - Intertemporal constraints on, for instance, downside risk.
  - Efficient numerical implementation



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