Time-Consistent Mean-Variance Portfolio Selection in Discrete and Continuous Time

Christoph Czichowsky

Department of Mathematics ETH Zurich

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Mean-variance portfolio selection in one period

- Harry Markowitz (Portfolio selection, 1952):
	- \triangleright maximise return and minimise risk
	- \blacktriangleright return=expectation
	- \blacktriangleright risk=variance
- **•** Mean-variance portfolio selection with risk aversion $\gamma > 0$ in one period:

$$
U(\vartheta) = E[x + \vartheta^\top \Delta S] - \frac{\gamma}{2} \text{Var}[x + \vartheta^\top \Delta S] = \max_{\vartheta}
$$

• Solution is the so-called **mean-variance efficient strategy**, i.e.

$$
\widetilde{\vartheta}:=\frac{1}{\gamma}\,\text{Cov}[\Delta S|\mathcal{F}_0]^{-1}E[\Delta S|\mathcal{F}_0]=:\widehat{\vartheta}.
$$

Question: How does this extend to multi-period or continuous time?

Markowitz problem:

$$
U(\vartheta) = E\Big[x + \int_0^T \vartheta_u dS_u\Big] - \frac{\gamma}{2} \text{Var}\left[x + \int_0^T \vartheta_u dS_u\right] = \max_{(\vartheta_s)_{0 \le s \le T}}!
$$

- Static: criterion at time 0 determines optimal $\widetilde{\vartheta}$ via $\widetilde{g} = \int_0^T \widetilde{\vartheta} dS$.
- **Question:** more explicit dynamic description of $\widetilde{\vartheta}$ on [0, T] from $\widetilde{\varrho}$?

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- \bullet Dynamic: Use $\widetilde{\vartheta}$ on (0, t] and determine optimal strategy on (t, T] via

$$
U_t(\vartheta) = E\left[x + \int_0^T \vartheta_u dS_u \Big| \mathcal{F}_t\right] - \frac{\gamma}{2} \text{Var}\left[x + \int_0^T \vartheta_u dS_u \Big| \mathcal{F}_t\right] = \max_{\{\vartheta_s\}_{t \leq s \leq T}}
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$$

- **Time inconsistent:** this optimal strategy is different from $\tilde{\vartheta}$ on $(t, T)!$
- Time-consistent mean-variance portfolio selection: \bullet Find a strategy $\widehat{\vartheta}$, which is "optimal" for $U_t(\vartheta)$ and time-consistent.

Previous literature

 \bullet Strotz (1956): "choose the best plan among those that [you] will actually follow." \rightarrow Recursive approach to time inconsistency for a different problem.

In Markovian models: Deterministic functions, HJB PDEs and verification thm.

- **E** Ekeland et al. (2006): game theoretic formulation for different problems.
- Basak and Chabakauri (2007): results for mean-variance portfolio selection. \bullet
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- 2) Rigorous justification of the continuous-time formulation?

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Financial market:

- \mathbb{R}^d -valued semimartingale S wlog. $S = S_0 + M + A \in \mathcal{S}^2(P)$.
- $\Theta = \Theta_S := \{ \vartheta \in L(S) \mid \int \vartheta dS \in S^2(P) \} = L^2(M) \cap L^2(A).$

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Local mean-variance efficiency in discrete time

• Use
$$
x + \vartheta \cdot S := x + \int_0^T \vartheta_u dS_u = x + \sum_{i=1}^T \vartheta_i \Delta S_i
$$
 and suppose $d = 1$.

Definition

A strategy $\hat{\vartheta} \in \Theta$ is locally mean-variance efficient (LMVE) if

$$
U_{k-1}(\widehat{\vartheta})-U_{k-1}(\widehat{\vartheta}+\delta \mathbb{1}_{\{k\}})\geq 0 \qquad P\text{-}a.s.
$$

for all $k = 1, ..., T$ and any $\delta = (\vartheta - \widehat{\vartheta}) \in \Theta$.

• Recursive optimisation (Källblad 2008): $\hat{\theta} \in \Theta$ is LMVE if and only if

$$
\widehat{\vartheta}_k = \frac{1}{\gamma} \frac{E[\Delta S_k | \mathcal{F}_{k-1}]}{\text{Var}[\Delta S_k | \mathcal{F}_{k-1}]} - \frac{\text{Cov} \left[\Delta S_k, \sum_{i=k+1}^T \widehat{\vartheta}_i \Delta S_i | \mathcal{F}_{k-1} \right]}{\text{Var} [\Delta S_k | \mathcal{F}_{k-1}]} = \frac{1}{\gamma} \lambda_k - \xi_k(\widehat{\vartheta})
$$

for $k = 1, \ldots, T$.

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\widehat{\vartheta}_{k} = \frac{1}{\gamma} \frac{E[\Delta S_{k} | \mathcal{F}_{k-1}]}{\text{Var}[\Delta S_{k} | \mathcal{F}_{k-1}]} - \frac{\text{Cov} \left[\Delta S_{k}, \sum_{i=k+1}^{T} \widehat{\vartheta}_{i} \Delta S_{i} \Big| \mathcal{F}_{k-1} \right]}{\text{Var}[\Delta S_{k} | \mathcal{F}_{k-1}]} = \frac{1}{\gamma} \lambda_{k} - \xi_{k}(\widehat{\vartheta})
$$
\n
$$
= \frac{1}{\gamma} \frac{\Delta A_{k}}{E \left[(\Delta M_{k})^{2} | \mathcal{F}_{k-1} \right]} - \frac{E \left[\Delta M_{k} E \left[\sum_{i=k+1}^{T} \widehat{\vartheta}_{i} \Delta S_{i} \Big| \mathcal{F}_{k} \right] \Big| \mathcal{F}_{k-1} \right]}{E \left[(\Delta M_{k})^{2} | \mathcal{F}_{k-1} \right]}
$$
\nfor $k = 1, ..., T$.

Structure condition and mean-variance tradeoff process

• S satisfies the **structure condition (SC)**, i.e. there exists a predictable process λ such that

$$
A_k = \sum_{i=1}^k \lambda_i E\left[(\Delta M_i)^2 | \mathcal{F}_{i-1} \right] = \sum_{i=1}^k \lambda_i \Delta \langle M \rangle_i
$$

for $k = 0, \ldots, T$ and the mean-variance tradeoff process (MVT)

$$
K_k := \sum_{i=1}^k \frac{\left(E[\Delta S_i | \mathcal{F}_{i-1}]\right)^2}{Var[\Delta S_i | \mathcal{F}_{i-1}]} = \sum_{i=1}^k \lambda_i^2 \Delta \langle M \rangle_i = \sum_{i=1}^k \lambda_i \Delta A_i
$$

for $k = 0, ..., T$ is finite-valued, i.e. $\lambda \in L^2_{loc}(M)$.

- If the LMVE strategy ϑ exists, then $\lambda \in L^2(M)$, i.e. $K_T \in L^1(P)$.
- **Comments:** 1) SC and MVT also appear naturally in other quadratic optimisation problems in mathematical finance; see Schweizer (2001). 2) No arbitrage condition: $A \ll \langle M \rangle$.

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Expected future gains

• For each $\vartheta \in \Theta$, define the **expected future gains** $Z(\vartheta)$ and the square integrable martingale $Y(\vartheta)$ by

$$
Z_k(\vartheta) := E\left[\sum_{i=k+1}^T \vartheta_i \Delta S_i \middle| \mathcal{F}_k\right] = E\left[\sum_{i=1}^T \vartheta_i \Delta A_i \middle| \mathcal{F}_k\right] - \sum_{i=1}^k \vartheta_i \Delta A_i
$$

$$
=: Y_k(\vartheta) - \sum_{i=1}^k \vartheta_i \Delta A_i
$$

$$
= Y_0(\vartheta) + \sum_{i=1}^k \xi_i(\vartheta) \Delta M_i + L_k(\vartheta) - \sum_{i=1}^k \vartheta_i \Delta A_i
$$

for $k = 0, 1, \ldots, T$ inserting the **GKW decomposition** of $Y(\vartheta)$.

Lemma

The LMVE strategy $\widehat{\vartheta}$ exists if and only if 1) S satisfies (SC) with $\lambda \in L^2(M)$, i.e. $K_T \in L^1(P)$, and 2) $\widehat{\vartheta} = \frac{1}{\gamma} \lambda - \xi(\widehat{\vartheta})$.

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Global description of $\xi(\widehat{\vartheta})$ via FS decomposition

• Combining both representations we obtain

$$
\sum_{i=1}^{T} \widehat{\vartheta}_{i} \Delta A_{i} = \sum_{i=1}^{T} \left(\frac{1}{\gamma} \lambda_{i} - \xi_{i}(\widehat{\vartheta}) \right) \Delta A_{i}
$$
\n
$$
= Y_{0}(\widehat{\vartheta}) + \sum_{i=1}^{T} \xi_{i}(\widehat{\vartheta}) \Delta M_{i} + L_{T}(\widehat{\vartheta})
$$
\n
$$
\frac{1}{\gamma} K_{T} = \frac{1}{\gamma} \sum_{i=1}^{T} \lambda_{i} \Delta A_{i} = Y_{0}(\widehat{\vartheta}) + \sum_{i=1}^{T} \xi_{i}(\widehat{\vartheta}) \Delta S_{i} + L_{T}(\widehat{\vartheta}) \tag{1}
$$

- [\(1\)](#page-14-0) is almost the Föllmer–Schweizer (FS) decomposition of $\frac{1}{\gamma}K_{\mathcal{T}}.$
- The integrand $\xi(\widehat{\vartheta})=:\frac{1}{\gamma}\widehat{\xi}$ in the FS decomposition yields the locally risk-minimising strategy for the contingent claim $\frac{1}{\gamma}K_{\mathcal{T}}.$
- **Global description:** $\hat{\theta} \in \Theta$ exists iff [\(1\)](#page-14-0) and $\hat{\theta} = \frac{1}{\gamma}(\lambda \hat{\xi})$.

Continuous time setting

- \bullet Increasing, integrable, predictable process B called "operational time" such that: $A = a \cdot B$, $\langle M, M \rangle = \widetilde{c}^M \cdot B$ and $a = \widetilde{c}^M \lambda + \eta$ with $\eta \in \text{Ker}(\widetilde{c}^M)$.
- **•** S satisfies the **structure condition (SC)**, if $\eta = 0$, i.e.

$$
\mathcal{A}=\int d\langle M\rangle \lambda,
$$

and the mean-variance tradeoff process (MVT)

$$
K_t := \int_0^t \lambda_u^{\top} d \langle M \rangle_u \lambda_u = \int_0^t \lambda_u d A_u < +\infty.
$$

Expected future gains $Z(\vartheta)$ and GKW decomposition of $Y(\vartheta)$ \bullet

$$
Z_t(\vartheta) := E\left[\int_t^T \vartheta_u dS_u \Big| \mathcal{F}_t\right] = E\left[\int_0^T \vartheta_u dA_u \Big| \mathcal{F}_t\right] - \int_0^t \vartheta_u dA_u
$$

$$
=: Y_t(\vartheta) - \int_0^t \vartheta_u dA_u
$$

$$
= Y_0(\vartheta) + \int_0^t \xi_u(\vartheta) dM_u + L_t(\vartheta) - \int_0^t \vartheta_u dA_u
$$

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Local mean-variance efficiency in continuous time

IDEA: Combine *recursive optimisation* with a *limiting argument*.

Definition

A strategy $\hat{\theta} \in \Theta$ is locally mean-variance efficient (in continuous time) if

$$
\varliminf_{n\to\infty}u^{\Pi_n}[\widehat{\vartheta},\delta]:=\varliminf_{n\to\infty}\sum_{t_i,t_{i+1}\in\Pi_n}\frac{U_{t_i}(\widehat{\vartheta})-U_{t_i}(\widehat{\vartheta}+\delta1\!\!1_{(t_i,t_{i+1}]})}{E[B_{t_{i+1}}-B_{t_i}|\mathcal{F}_{t_i}]}\mathbb{1}_{(t_i,t_{i+1}]}\geq 0\quad \text{P}\otimes B\text{-a.e.}
$$

for any increasing sequence (Π_n) of partitions such that $|\Pi_n| \to 0$ and any $\delta \in \Theta$.

Inspired by the concept of local risk-minimisation (LRM); Schweizer (88, 08).

$$
\lim_{n\to\infty} u^{\Pi_n}[\widehat{\vartheta},\delta] = \left(\gamma(\xi(\widehat{\vartheta}) + \widehat{\vartheta}) - \lambda + \frac{\gamma}{2}\delta\right)^{\top} c^M \delta - \delta^{\top} \eta \qquad P \otimes B\text{-a.e.}
$$

● Remarks: 1) Convergence without any additional assumptions, i.e. boundedness assumptions on δ and continuity of A. 2) Generalises also results for LRM.

The LMVE strategy $\widehat{\vartheta}$ in continuous time

Theorem

1) The LMVE strategy $\widehat{\vartheta} \in \Theta$ exists if and only if

- i) S satisfies (SC) with $\lambda \in L^2(M)$, i.e. $K_\mathcal{T} \in L^1(\mathcal{P})$.
- ii) $\hat{\vartheta}=\frac{1}{\gamma}\lambda-\xi(\hat{\vartheta})$, i.e. $\hat{J}(\hat{\vartheta})=\hat{\vartheta}$, where $\hat{J}(\psi):=\frac{1}{\gamma}\lambda-\xi(\psi)$ for $\psi\in\Theta$ and $\xi(\psi)$ is the integrand in the GKW decomposition of $\int_0^T\psi_u dA_u.$

2) If K is bounded and continuous, $\widehat{J}(\cdot)$ is a contraction on $(\Theta, \|\cdot\|_{\beta,\infty})$ where

$$
\|\vartheta\|_{\beta,\infty} := \Big\|\Big(\int_0^T \frac{1}{\mathcal{E}(-\beta K)_u} \vartheta_u^{\top} d\langle M \rangle_u \vartheta_u\Big)^{\frac{1}{2}}\Big\|_{L^2(P)} \sim \|\vartheta\|_{L^2(M)} + \|\vartheta\|_{L^2(A)}.
$$

In particular, the LMVE strategy $\widehat{\vartheta}$ is given as the limit $\widehat{\vartheta} = \lim_{n \to \infty} \vartheta^n$ in $(\Theta, \|\. \|_{\beta,\infty})$, where $\vartheta^{n+1} = \widehat{J}(\vartheta^n)$ for $n \geq 1$, for any $\vartheta^0 = \vartheta \in \Theta$.

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Global description of $\xi(\widehat{\vartheta})$ via FS decomposition

Theorem

The LMVE strategy $\widehat{\vartheta} \in \Theta$ exists if and only if S satisfies (SC) and the MVT process $K_{\mathcal{T}}\in L^1(P)$ and can be written as

$$
K_{T} = \widehat{K}_{0} + \int_{0}^{T} \widehat{\xi} dS + \widehat{L}_{T}
$$
 (2)

with $K_0 \in L^2(\mathcal{F}_0)$, $\xi \in L^2(M)$ such that $\xi - \lambda \in L^2(A)$ and $L \in \mathcal{M}_0^2(P)$ strongly orthogonal to M. In that case, $\widehat{\vartheta} = \frac{1}{\gamma} \big(\lambda - \widehat{\xi} \big)$, $\xi(\widehat{\vartheta}) = \frac{1}{\gamma} \widehat{\xi}$ and $U(\widehat{\vartheta}) = \ldots$ [\(2\)](#page-18-1).

If the minimal martingale measure exists, i.e. $\frac{d\widehat{P}}{dP} := \mathcal{E}(-\lambda \cdot M)_T \in L^2(P)$ and strictly positive, and $K_\mathcal{T} \in L^2(P)$, then

$$
Z_t(\widehat{\vartheta}) = \frac{1}{\gamma} \left(\widehat{K}_0 + \int_0^t \widehat{\xi} dS + \widehat{L}_t - K_t \right) = \frac{1}{\gamma} \widehat{E}[K_T - K_t | \mathcal{F}_t],
$$

and $\hat{\xi}$ is related to the GKW of $K_{\mathcal{T}}$ under \hat{P} ; see Choulli et al. (2010).

Application in concrete models: 1) λ λ λ , 2) K, 3) $\mathcal{E}(-\lambda \cdot M)$ a[nd](#page-22-0) [4\)](#page-0-0) $\widehat{\xi}$ [.](#page-22-0)..

Discretisation of the financial market

• Let $(\Pi_n)_{n\in\mathbb{N}}$ increasing such that $|\Pi_n|\to 0$ and $S=S_0+M+A$.

Discretisation of processes

 $S_t^n := S_{t_i}, M_t^n := M_{t_i}$ and $A_t^n := A_{t_i}$ for $t \in [t_i, t_{i+1})$ and all $t_i \in \Pi_n$.

Discretisation of filtration

$$
\text{•} \ \mathcal{F}^n_{t_i} := \mathcal{F}_{t_i} \text{ for } t \in [t_i, t_{i+1}) \text{ and all } t_i \in \Pi_n \text{ and } \mathbb{F}^n := (\mathcal{F}^n_t)_{0 \leq t \leq T}.
$$

Canonical decomposition of $S^n=S_0+\bar{M}^n+\bar{A}^n\in\mathcal{S}^2(P,\mathbb{F}^n)$

$$
\bullet \ \ \bar{A}_t^n := \sum_{k=1}^i E[\Delta A_{t_k}^n | \mathcal{F}_{t_{k-1}}] = A_t^n - \mathbf{M_t}^{\mathbf{A},n}
$$

 $\bar{M}_t^n := M_t^n + \mathsf{M_t^{A,n}}$ for $t \in [t_i, t_{i+1})$

where the **"discretisation error"** is given by the \mathbb{F}^n -martingale

$$
\mathbf{M_t^{A,n}} := \sum_{k=1}^i (\Delta A_{t_k}^n - E[\Delta A_{t_k}^n | \mathcal{F}_{t_{k-1}}]) \text{ for } t \in [t_i, t_{i+1}).
$$

Convergence of solutions ϑ^n

- **•** Due to time inconsistency usual abstract arguments don't work.
- Work with **global description** directly to show

$$
\widehat{\vartheta}^n=\frac{1}{\gamma}\big(\lambda^n-\widehat{\xi}^n\big)\stackrel{L^2(M)}{\longrightarrow}\widehat{\vartheta}=\frac{1}{\gamma}\big(\lambda-\widehat{\xi}\big),\quad \text{as}\;|\Pi^n|\to 0.
$$

Discrete- and continuous-time FS decomposition

$$
K_T^n = \widehat{K}_0^n + \sum_{t_i \in \Pi_n} \widehat{\xi}_{t_i}^n \Delta S_{t_i}^n + \widehat{L}_T^n \text{ and } K_T = \widehat{K}_0 + \int_0^T \widehat{\xi}_u dS_u + \widehat{L}_T.
$$

• For this we establish

1)
$$
\lambda^{n} = \sum_{t_{i}, t_{i+1} \in \Pi_{n}} \frac{\Delta \bar{A}_{t_{i+1}}^{n}}{E[(\Delta \bar{M}_{t_{i+1}}^{n})^{2} | \mathcal{F}_{t_{i}}]} \mathbb{1}_{(t_{i}, t_{i+1}]} \xrightarrow{L^{2}(M)} \lambda
$$

\n2)
$$
K_{T}^{n} = \sum_{t_{i}, t_{i+1} \in \Pi_{n}} \lambda_{t_{i+1}}^{n} \Delta \bar{A}_{t_{i+1}}^{n} \xrightarrow{L^{2}(P)} K_{T} = \int_{0}^{T} \lambda_{u} dA_{u}
$$

\n3) 2),
$$
|\Pi_{n}| \to 0 \text{ implies } \hat{\xi}^{n} \xrightarrow{L^{2}(M)} \hat{\xi}.
$$

- Problem to control the "discretisation error" MA,n.
- **Simple sufficient con[d](#page-19-0)itio[n](#page-19-0):** $K = \int \frac{dK}{dt} dt$ $K = \int \frac{dK}{dt} dt$ $K = \int \frac{dK}{dt} dt$ a[nd](#page-22-0) $\frac{dK}{dt}$ [u](#page-21-0)n[ifo](#page-20-0)r[ml](#page-0-0)[y b](#page-22-0)[ou](#page-0-0)nd[ed](#page-0-0)[.](#page-22-0) Ω

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Thank you for your attention!

http://www.math.ethz.ch/∼czichowc

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