

# Computing Optimal Investment Strategies Under Partial Information and Bounded Shortfall Risk

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# Dynamic Portfolio Optimization

- Financial market** containing risky and risk-free assets  
continuously tradable  
**partial information** on the drift
- Initial capital**  $x_0 > 0$
- Horizon**  $[0, T]$
- Aim** **maximize expected utility** of terminal wealth  
**constrain the risk** of falling short a benchmark
- Problem** find an **optimal investment strategy**

# Financial Market Model

$(\Omega, \mathcal{F} = (\mathcal{F}_t)_{t \in [0, T]}, P)$  filtered probability space

## Stock market

$$dS_t = \text{Diag}(S_t) (\mu_t dt + \sigma_t dW_t)$$

with drift  $\mu_t \in \mathbb{R}^n$  and volatility  $\sigma_t \in \mathbb{R}^{n \times n}$

$$dR_t = \left( \frac{dS_t^1}{S_t^1}, \dots, \frac{dS_t^n}{S_t^n} \right)^\top \quad \text{return process}$$

## Money market

risk-free interest rate  $r_t$

(for simplicity we set  $r_t \equiv 0$  and  $\sigma_t \equiv \sigma$ )

## Martingale density

$$Z_t = \exp \left( - \int_0^t \kappa_s^\top dW_s - \frac{1}{2} \int_0^t \|\kappa_s\|^2 ds \right)$$

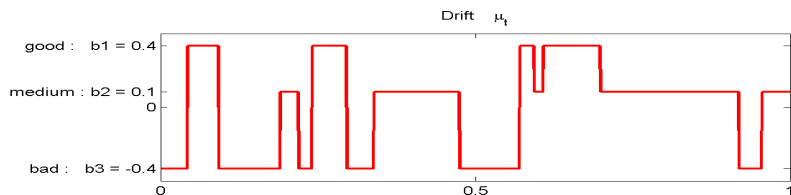
with  $\kappa_t = \sigma^{-1} \mu_t$  market price of risk

## Martingale measure

$$\tilde{P}(A) = E [Z_T 1_A] \quad \text{for } A \in \mathcal{F}_T$$

$\tilde{P} \sim P$  and  $\tilde{W}_t = W_t + \int_0^t \kappa_s ds$  is BM w.r.t.  $\tilde{P}$

# Hidden Markov Model



Returns  $dR_t = \mu(Y_t) dt + \sigma dW_t$  observations

Drift  $\mu(Y_t) = B Y_t$  non-observable (hidden) state  
not  $\mathcal{F}^S$ -adapted

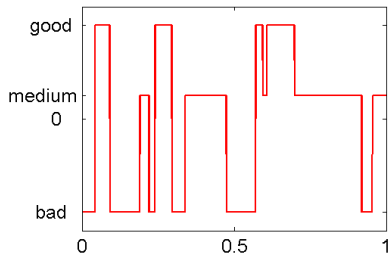
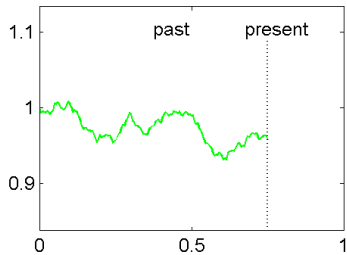
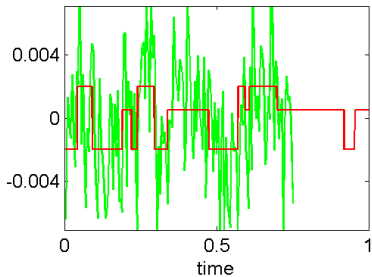
with  $Y_t$  time-continuous homogeneous Markov chain

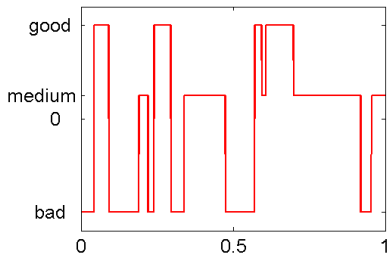
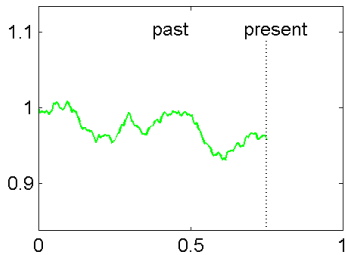
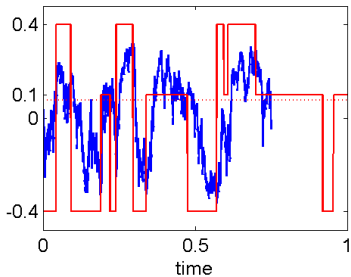
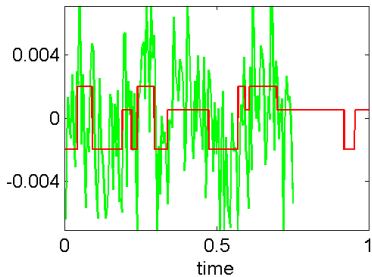
independent of  $W_t$

states of  $Y_t$  are unit vectors in  $\mathbb{R}^d$ :  $e_1, \dots, e_d$

$n \times d$ -matrix  $B = (b^1, \dots, b^d)$ , columns are states of  $\mu_t$

switching between the states is controlled by intensity matrix  $G$

Drift  $\mu_t$ Stock price  $S_t$ Return  $dS_t / S_t = \mu_t dt + \sigma dW_t$ 

Drift  $\mu_t$ Stock price  $S_t$ Filter for  $\mu_t$ Return  $dS_t / S_t = \mu_t dt + \sigma dW_t$ 

# Filter

**Given** observations of prices  $S_u$  (returns  $R_u$ ) for  $u \in [0, t]$

**To find** filter for state  $Y_t: \eta_t = E [Y_t | \mathcal{F}_t^S] \Rightarrow E [\mu_t | \mathcal{F}_t^S] = B\eta_t$

martingale density  $Z_t: \zeta_t = E [Z_t | \mathcal{F}_t^S]$

**Solution** Wonham (1965), Elliot (1993)

unnormalized filter for state  $Y_t: \mathcal{E}_t := \tilde{E} [Z_T^{-1} Y_t | \mathcal{F}_t^S]$

$(\mathcal{E}_t)$  satisfies  $d$ -dimensional linear SDE

$$d\mathcal{E}_t = G^T \mathcal{E}_t dt + \text{Diag}(\mathcal{E}_t) B^T (\sigma \sigma^T)^{-1} \underbrace{dR_t}_{\text{observations}}, \quad \mathcal{E}_0 = E [Y_0]$$

**Filter** for  $Z_t: \zeta_t = (\mathbf{1}_d^T \mathcal{E}_T)^{-1} = \frac{1}{\mathcal{E}_T^1 + \dots + \mathcal{E}_T^d}$

for  $Y_t: \eta_t = \zeta_t \mathcal{E}_t$

# Portfolio

Initial capital  $X_0 = x_0 > 0$

Wealth at time  $t$   $X_t = \underbrace{\pi_t^0}_{\text{bond}} + \underbrace{\pi_t^1}_{\text{stock 1}} + \dots + \underbrace{\pi_t^n}_{\text{stock n}}$   
invested in

**Strategy**  $\pi_t = (\pi_t^1, \dots, \pi_t^n)^\top$

Self financing condition  $\Rightarrow$

## Wealth equation

$$dX_t^\pi = \pi_t^\top (\sigma dW_t + \mu_t dt)$$

$$= \pi_t^\top \sigma d\widetilde{W}_t$$

$$X_0^\pi = x_0$$



# Shortfall Risk

Compare terminal wealth  $X_T$  with benchmark  $q$

e.g.  $q \sim x_0$  initial capital

**Shortfall** if  $X_T < q$

**Risk Measure**  $E_Q [(X_T - q)^-]$  where  $Q \sim P$   
Expected Loss

## Special Cases

- ▶  $Q = \tilde{P}$  **Present** Expected Loss (PEL)  
 $\tilde{E} [(X_T - q)^-]$  option price  
Basak, Shapiro (2001)
- ▶  $Q = P$  **Future** Expected Loss (FEL)  
 $E [(X_T - q)^-]$  average additional capital  
Gabih, Grecksch, W. (2005)

# Optimization Problem

**Wealth equation**  $dX_t^\pi = \pi_t^\top (\mu_t dt + \sigma dW_t), \quad X_0^\pi = x_0$

**Strategy**  $\pi = (\pi_t)_{t \in [0, T]}$

**Admissible strategies**  $\mathcal{A}(x_0) = \{(\pi_t) : \mathcal{F}^S\text{-adapted}$   
integrability conditions,  
 $X_t^\pi \geq 0, \forall t \in [0, T]\}$

**Utility function**  $U : [0, \infty) \rightarrow \mathbb{R} \cup \{-\infty\}$   
strictly increasing, concave (log-, power utility)

## Optimization problem

$$V(x_0) = \sup_{\pi \in \mathcal{A}(x_0)} E[U(X_T^\pi)]$$

$$\text{risk constraint } E_Q[(X_T^\pi - q)^-] \leq \varepsilon$$

# Decomposition of the OP: Full Information

**Dynamic problem**      $V(x_0) = \sup_{\pi \in \mathcal{A}(x_0)} E[U(X_T^\pi)], \quad E_Q[(X_T^\pi - q)^-] \leq \varepsilon$

**Static problem**      $V(x_0) = \sup_{\xi \in \mathcal{B}(x_0)} E[U(\xi)], \quad E_Q[(\xi - q)^-] \leq \varepsilon$

where  $\mathcal{B}(x_0) := \{\xi \geq 0 : \xi \text{ is } \mathcal{F}_T^S\text{-measurable, } \underbrace{\tilde{E}[\xi] \leq x_0}_{\text{budget-constraint}}\}$

terminal wealth generated from initial capital in  $(0, x_0]$

→ **optimal terminal wealth**      $\xi^* = f(Z_T)$

## Representation problem

find a strategy  $\pi \in \mathcal{A}(x_0)$  such that  $\xi^* = X_T^\pi$

→ **optimal strategy**      $\pi_t^*$

# The Case of Partial Information

**Problem** martingale density  $Z_t$  is  $\mathcal{F}$ - but **not  $\mathcal{F}^S$ -adapted**

**Idea** replace  $Z_t$  by its **filter**  $\zeta_t = E [Z_t | \mathcal{F}_t^S]$

**Optimal terminal wealth**  $X_T^* = \xi^* = f(\zeta_T)$

$$V(x_0) = \sup_{\xi \in \mathcal{B}(x_0)} E[U(\xi)]$$

risk constraint  $E_Q [(\xi - q)^-] \leq \varepsilon$

$$\mathcal{B}(x_0) := \{\xi \geq 0 : \xi \text{ is } \mathcal{F}_T^S\text{-measurable, } \tilde{E}[\xi] \leq x_0\}$$

Choose the bound  $\varepsilon$  such that the risk constraint

- ▶ is binding  $\varepsilon \leq \varepsilon_{\max} = E_Q [(X_T^M - q)^-]$   
risk of the Merton portfolio  
(no risk constraint)
- ▶ can be fulfilled  $\varepsilon \geq \varepsilon_{\min} = \dots$  Gabih, Sass, W. (2006)

# Optimal Terminal Wealth

Theorem ( $Q = \tilde{P}$  Present Expected Loss)

For  $\varepsilon \in (\varepsilon_{\min}, \varepsilon_{\max})$  the PEL-optimal terminal wealth is

$$\xi^* = f(\zeta_T; y_1^*, y_2^*)$$

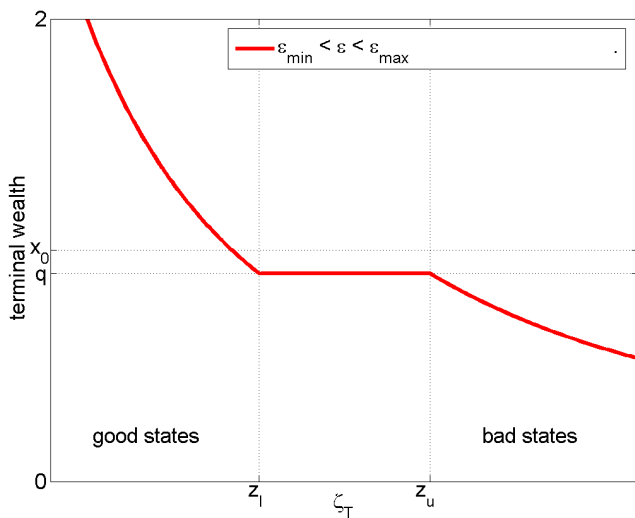
$$\text{where } f(z; y_1, y_2) = \begin{cases} I(y_1 z) & \text{for } z \in (0, z_l] \\ q & \text{for } z \in (z_l, z_u] \\ I((y_1 - y_2)z) & \text{for } z \in (z_u, \infty). \end{cases}$$

$$I = (U')^{-1}, \quad z_l = \frac{U'(q)}{y_1} \quad \text{and} \quad z_u = \frac{U'(q)}{y_1 - y_2}.$$

The real numbers  $y_1^*, y_2^* > 0$  uniquely solve the equations

$$\begin{aligned} \tilde{E}[f(\zeta_T; y_1, y_2)] &= x_0 \\ E_Q[(f(\zeta_T; y_1, y_2) - q)^-] &= \varepsilon. \end{aligned}$$

# Optimal Terminal Wealth: The Function $f(z; y_1, y_2)$



# Computation of the Lagrange Multipliers $y_1, y_2$

System of nonlinear equations

$$\tilde{E}[f(\zeta_T; y_1, y_2)] = x_0$$

$$E_Q[(f(\zeta_T; y_1, y_2) - q)^-] = \varepsilon$$

**Existence & Uniqueness:** Gabih, Sass, W. (2006)

**Solution** requires

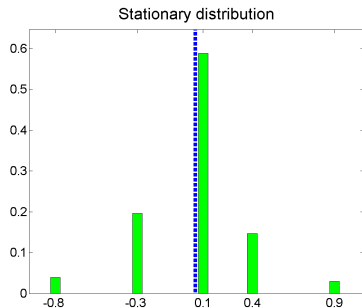
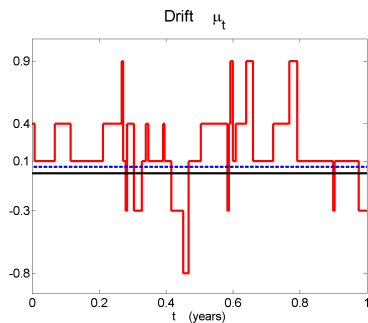
- ▶ Approximation of the expectation in on the left-hand sides using Monte-Carlo simulation
- ▶ Numerical methods for solving nonlinear equations



# Example: Parameter of the Financial Market

$n = 1$  stock with volatility  $\sigma = 0.25$

HMM for the drift  $\mu_t$  with  $d = 5$  states



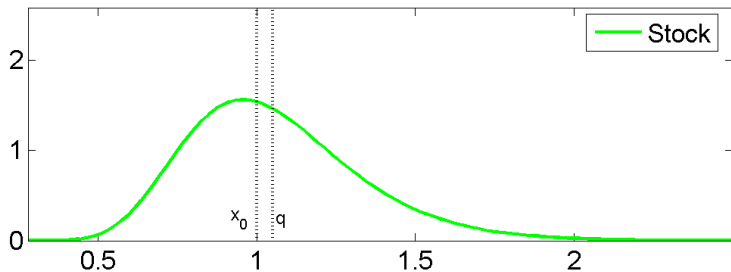
Ergodic mean  $\bar{\mu} \approx 0.054$

# Example: Parameter for the Portfolio Optimization

<b>Horizon</b>	$T = 1$ year
<b>Utility function</b>	$U(x) = 2x^{1/2} - 2$ (power utility)
<b>Initial capital</b>	$x_0 = 1$
<b>Benchmark</b>	$q = 1.05$ ( $q > x_0$ , portfolio insurance impossible)
<b>Risk measure</b>	Present Expected Loss $\tilde{E}[(X_T - q)^-]$
<b>Bound</b>	$\varepsilon = 0.1$
<b>Minimal Risk</b>	$\varepsilon_{\min} = q - x_0 = 0.05$
<b>Maximal Risk</b>	$\varepsilon_{\max} \approx 0.248$ (i.e. $\varepsilon \approx 40\%$ of $\varepsilon_{\max}$ )

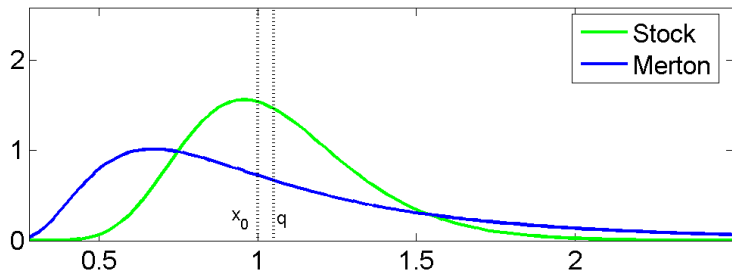
Monte-Carlo simulation:  $N = 10^7$  realizations of  $X_T^* = f(\zeta_T; y_1^*, y_2^*)$

# Distribution of Terminal Wealth



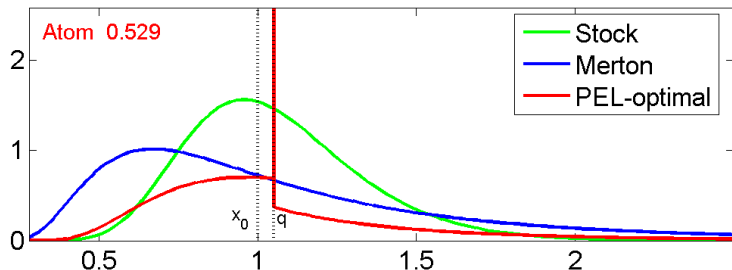
	Expected Utility	Risk
Stock	0.040	0.129

# Distribution of Terminal Wealth



	Expected Utility	Risk
Stock	0.040	0.129
Merton	0.076	$0.248 = \varepsilon_{max}$

# Distribution of Terminal Wealth



	Expected Utility	Risk
Stock	0.040	0.129
Merton	0.076	$0.248 = \varepsilon_{max}$
PEL-optimal	0.049	$0.100 = \varepsilon$

# Optimal Strategy

## Clark formula

Let  $D_t \xi$  be the Malliavin derivative of the  $\mathcal{F}_T^S$ -measurable r.v.  $\xi \in D_{1,1}$ , then it holds

$$\xi = \tilde{E}[\xi] + \int_0^T \tilde{E}[(D_t \xi)^\tau | \mathcal{F}_t^S] d\tilde{W}_t$$

For the Malliavin derivative of the optimal terminal wealth

$$\xi = f(\zeta_T) = f(\zeta_T; y_1^*, y_2^*) \quad \text{it holds} \quad D_t f(\zeta_T) = f'(\zeta_T) D_t \zeta_T$$

$$\text{Wealth equation} \quad \xi = x_0 + \int_0^T (\pi_t^*)^\tau \sigma d\tilde{W}_t$$

Comparison of coefficients  $\Rightarrow$  optimal strategy is

$$\pi_t^* = \sigma^{-\tau} \tilde{E} \left[ f'(\zeta_T) D_t \zeta_T | \mathcal{F}_t^S \right]$$

# Computation of the Optimal Strategy

$$\pi_t^* = \sigma^{-\tau} \tilde{E} \left[ f'(\zeta_T) D_t \zeta_T \mid \mathcal{F}_t^S \right] \quad (*)$$

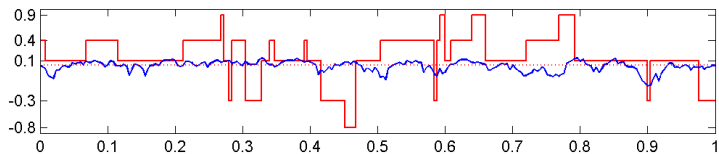
requires

- ▶ Numerical solution of SDE's for the Malliavin Derivative  $D_t \zeta_T$
- ▶ Approximation of the conditional expectation in (\*) using Monte-Carlo simulation

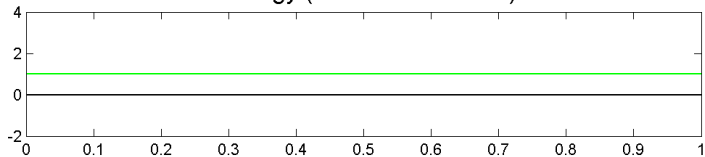
Generate  $N = 10^3$  realizations of  $\zeta_T$  and  $D_t \zeta_T$

⋮

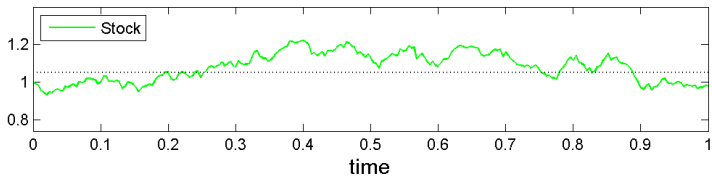
### Drift and Filter



### Strategy (Fraction of Wealth)

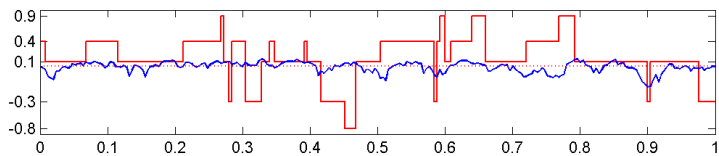


### Wealth

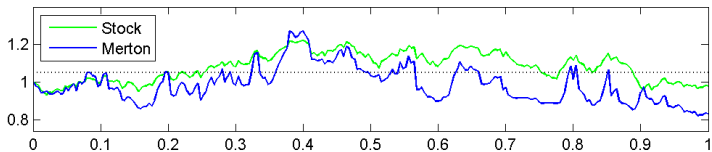
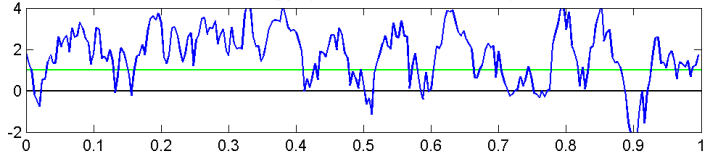




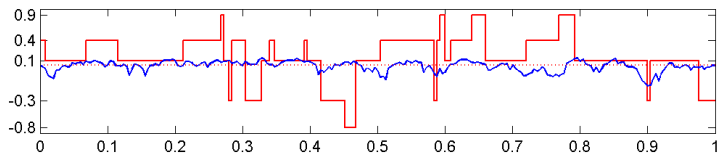
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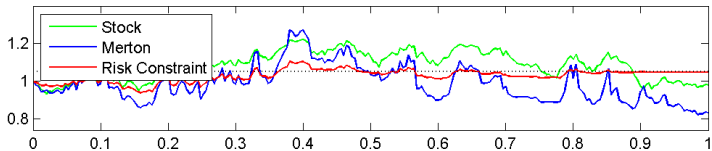
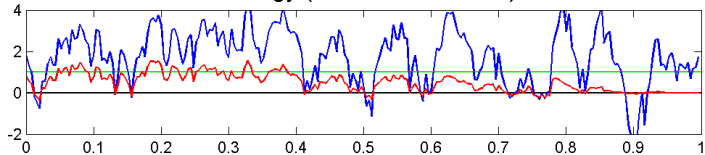
### Strategy (Fraction of Wealth)



### Drift and Filter



### Strategy (Fraction of Wealth)



# Trading in Discrete Time

## (A) Actual wealth

$$X_t^* = x_0 + \int_0^t (\pi_s^*)^\top dR_s \approx x_0 + \sum_{0 \leq s_j < t} (\hat{\pi}_{s_j})^\top \Delta R_{s_j} =: X_t^A$$

where  $\hat{\pi}_{s_j}$  is the Monte-Carlo approximation of  $\pi_{s_j}^*$

## (B) Theoretical optimal wealth

$$X_t^* = \tilde{E} \left[ X_T^* | \mathcal{F}_t^S \right] = \tilde{E} \left[ f(\zeta_T; y_1^*, y_2^*) | \mathcal{E}_t \right]$$

Given the unnormalized filter at time  $t$  is  $\mathcal{E}_t = x$  we find

$$X_t^* = \tilde{E} \left[ f(\zeta_T^{t,x}; y_1^*, y_2^*) \right] \approx X_t^B \quad (\text{by Monte-Carlo approximation})$$

Can be evaluated without computing the optimal strategy

If  $X_t^A \neq X_t^*$  then trading according to  $\pi^*$  is no longer optimal

# Updating of the Optimal Strategy

If we observe a "critical deviation"  $|X_t^A - X_t^B| > \delta$

we set up a new optimization problem with

Horizon  $T - t$

Initial capital  $X_t^A$  (actual wealth)

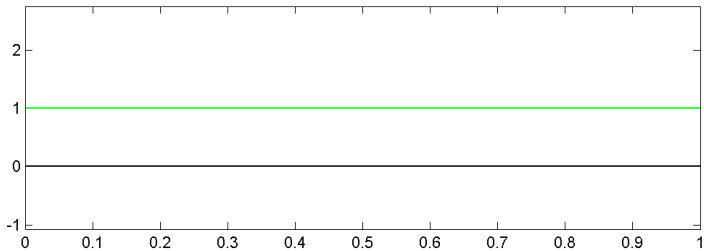
Risk bound Expected Loss of  $X_T^*$  at time  $t$  given  $\mathcal{F}_t^S$

$$\varepsilon_t^* = E_Q [(X_T^* - q)^- | \mathcal{F}_t^S] \quad (= \text{option price for PEL})$$

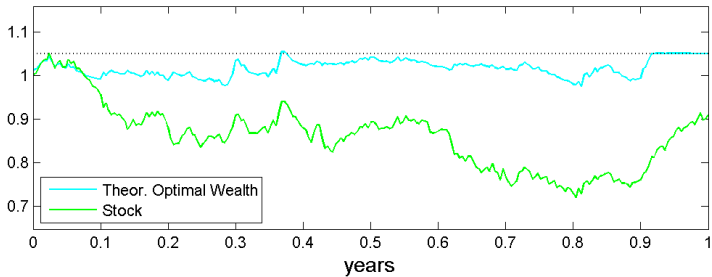
Compute new Lagrange multipliers  $y_1^t, y_2^t$  by solving

$$\begin{aligned} \tilde{E} [f(\zeta_T; y_1, y_2) | \mathcal{F}_t^S] &= X_t^A \\ E_Q [(f(\zeta_T; y_1, y_2) - q)^- | \mathcal{F}_t^S] &= \varepsilon_t^*. \end{aligned}$$

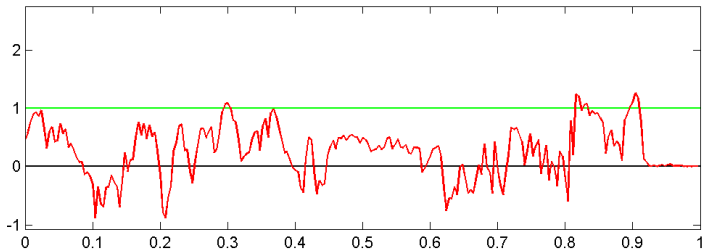
Strategy (Fraction of Wealth)



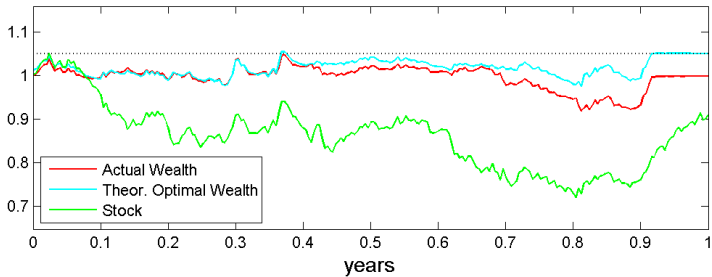
Wealth



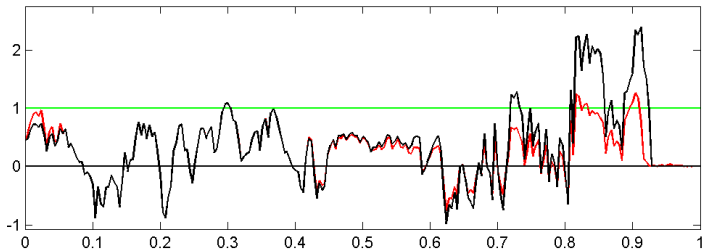
Strategy (Fraction of Wealth)



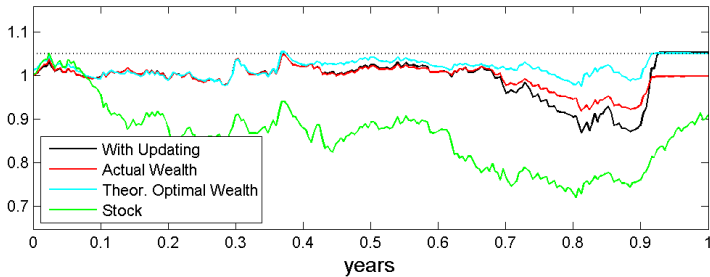
Wealth



Strategy (Fraction of Wealth)



Wealth



# Conclusion

- ▶ Dynamic portfolio optimization under risk constraint

$$V(x_0) = \sup_{\pi \in \mathcal{A}(x_0)} E[U(X_T^\pi)], \quad E_Q[(X_T^\pi - q)^-] \leq \varepsilon$$

- ▶ Partial information on the drift (Hidden Markov Model)

$X_T^*$  as a function of  $\zeta_T$ , the filter for the martingale density  $Z_T$

$\pi_t^*$  depends on Malliavin derivative  $D_t X_T^*$

- ▶ Optimal strategies can be computed using Monte-Carlo simulations
- ▶ For references see [www.fh-zwickau.de/~raw](http://www.fh-zwickau.de/~raw)