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A Problem of Optimal Portfolio Allocation with Transaction Costs

Benoîte de Saporta

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Groupe de Travail Finance, march 2 2006

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- State of the system W_t subject to stochastic dynamics
- Control process π_t
 - value chosen at each time depending on the available information: **adapted**
 - influence the dynamics of W_t
- Performance Criterion $J(W, \pi)$ to be maximised

Value Function

$$V = \sup_{\pi} J(W, \pi)$$

Aim

- Compute or identify the value function
- Find an optimal strategy (if it exists)

Example: Merton's Problem

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Market:	risk-free asset	$dS_t^0 = S_t^0 r dt$
	risky asset	$dS_t = \mu S_t dt + \sigma S_t dB_t$

- Control process $\pi_t \in [0; 1]$: proportion of the wealth invested in the risky asset
- State W_t : wealth process

$$\begin{aligned}\frac{dW_t^\pi}{W_t^\pi} &= \pi_t \frac{dS_t}{S_t} + (1 - \pi_t) \frac{dS_t^0}{S_t^0} \\ &= (\pi_t \mu + (1 - \pi_t)r) dt + \pi_t \sigma dB_t\end{aligned}$$

Example: Merton's Problem

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- Criterion: expected utility of the terminal wealth

$$U(x) = x^\alpha, \quad 0 < \alpha < 1$$

Value Function

$$V(t, x) = \sup_{\pi} \mathbb{E}[U(W_T^{t,x,\pi})]$$

Aim

- Compute or identify the value function
- Find an optimal strategy (if it exists)

Example: Merton's Problem

Hamilton-Jacobi-Bellman Equation

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Hamilton-Jacobi-Bellman (HJB) Equation

$$\frac{\partial \Phi}{\partial t}(t, x) + \sup_{p \in [0;1]} \mathcal{L}^p \Phi(t, x) = 0$$
$$\Phi(T, x) = U(x) = x^\alpha$$

$$\mathcal{L}^p \Phi(t, x) = x(p\mu + (1-p)r) \frac{\partial \Phi}{\partial x}(t, x) + \frac{1}{2} \sigma^2 p^2 x^2 \frac{\partial^2 \Phi}{\partial x^2}(t, x)$$

As $U(W_T^{t,\text{red},\pi}) = U(\text{red} W_T^{t,1,\pi}) = x^\alpha U(W_T^{t,1,\pi})$, we're searching for a factorized solution $\Phi(t, x) = x^\alpha \varphi(t)$

$$0 = \varphi'(t) + \varphi(t) \sup_{p \in [0;1]} \left\{ \alpha(p\mu + (1-p)r) + \frac{\alpha(\alpha-1)}{2} p^2 \sigma^2 \right\}$$

$$1 = \varphi(T)$$

Example: Merton's Problem

Solution of HJB Equation

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Solution

$$\Phi(t, x) = x^\alpha e^{\beta(T-t)}$$

with

$$\begin{aligned}\beta &= \sup_{p \in [0;1]} \left\{ \alpha(p\mu + (1-p)r) + \frac{\alpha(\alpha-1)}{2} p^2 \sigma^2 \right\} \\ &= \alpha r + \frac{\alpha(\mu - r)^2}{2(1-\alpha)\sigma^2}\end{aligned}$$

reached at $p^* = \frac{\mu-r}{(1-\alpha)\sigma^2}$

Example: Merton's Problem

Interpretation of HJB Equation

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Itô formula for Φ between t and T :

$$\Phi(T, W_T^{t,x,\pi}) = \Phi(t, x) + \int_t^T \left(\frac{\partial \Phi}{\partial u} + \mathcal{L}^{\pi_u} \Phi \right)(u, W_u^{t,x,\pi}) du + \text{martingale}$$

$$\begin{aligned} U(W_T^{t,x,\pi}) &\leq \Phi(t, x) + \int_t^T \left(\frac{\partial \Phi}{\partial u} + \sup_{p \in [0;1]} \mathcal{L}^p \Phi \right)(u, W_u^{t,x,\pi}) du \\ &\quad + \text{martingale} \\ &\leq \Phi(t, x) + \text{martingale} \end{aligned}$$

Hence

$$\Phi(t, x) \geq \mathbb{E}[U(W_T^{t,x,\pi})]$$

with equality when $\pi_t = p^*$.

Example: Merton's Problem

Conclusion

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Conclusion

- $V(t, x) = x^\alpha e^{\beta(T-t)}$
- Constant optimal strategy $\pi_t = \frac{\mu - r}{(1-\alpha)\sigma^2}$
- V is a solution of Hamilton Jacobi Bellman equation

Example: Merton's Problem

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- $V(t, x) = x^\alpha e^{\beta(T-t)}$
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Our Problem

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Market:

risk-free asset	$dS_t^0 = S_t^0 r dt$
risky asset	$dS_t = \mu(t) S_t dt + \sigma S_t dB_t$

- the drift alternately takes two different values
 $\mu_1 < 0$ and $\mu_2 > 0$
- transaction costs

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Main approaches to investment

- Fundamental approach
 - fundamental economic principles
- Technical Analysis approach
 - past prices behaviour
- Mathematical Approach
 - mathematical models

Aim

Compare the performance of technical analysis and miscalibrated mathematical models

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Market: risk-free asset $dS_t^0 = S_t^0 r dt$
 risky asset $dS_t = \mu(t) S_t dt + \sigma S_t dB_t$

- B standard Brownian motion,
- $\mu(t) \in \{\mu_1, \mu_2\}$ independent of B ,
- **control** process $\pi_t \in \{0, 1\}$ proportion of the wealth invested in the risky asset
- **state** W_t^π wealth when strategy π is applied
- **criterion** expected utility of the terminal wealth

Aim

Maximise the expected utility of the terminal wealth

Technical Analyst strategy

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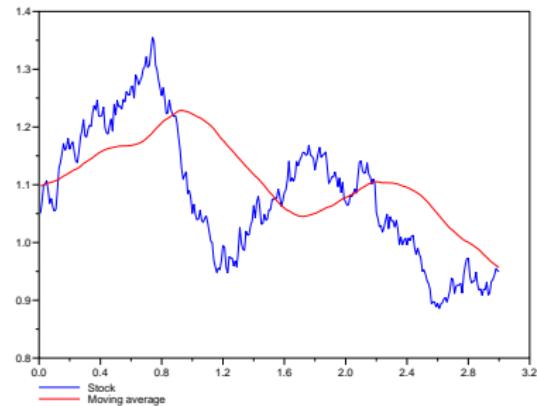
Efficient Strategy

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Moving average

$$M_t^\delta = \frac{1}{\delta} \int_{t-\delta}^t S_u du$$

- If $S_t > M_t^\delta$ buy
- If $S_t < M_t^\delta$ sell



$$\mu_1 = -0.2, \mu_2 = 0.2, \sigma = 0.15, \delta = 0.8.$$

Technical Analyst strategy

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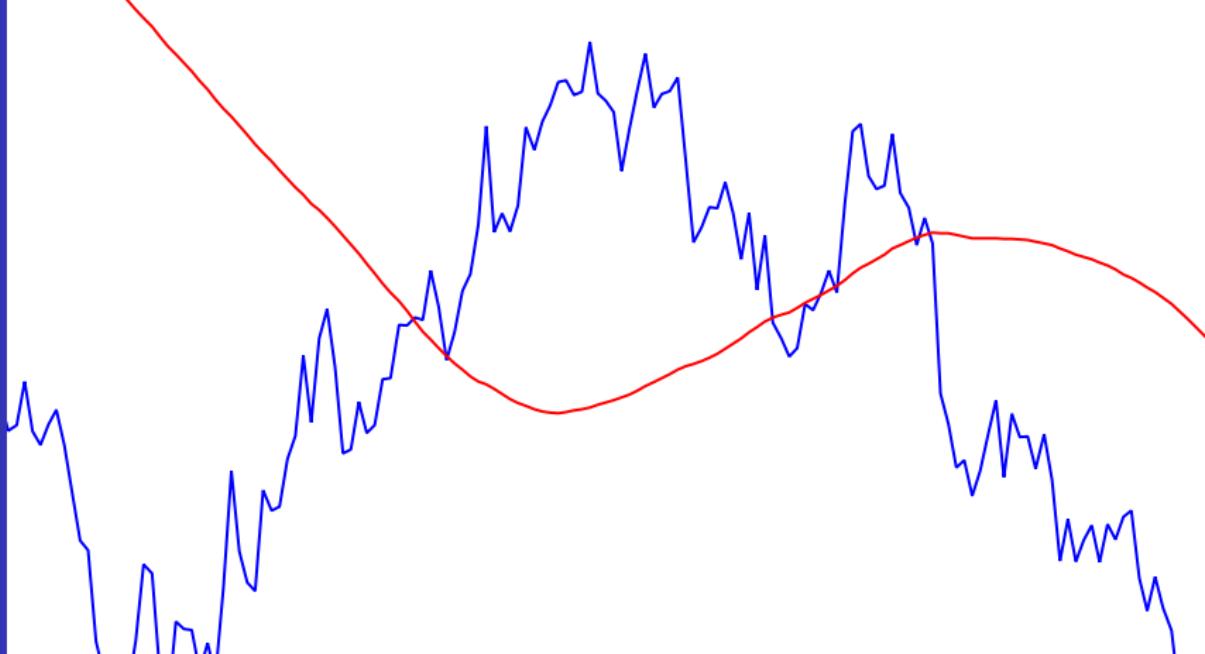
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BLANCHET, DIOP, GIBSON, KAMINSKI, TALAY, TANRÉ (2005)

risk-free asset

$$dS_t^0 = S_t^0 r dt,$$

risky asset

$$dS_t = \mu(t) S_t dt + \sigma S_t dB_t,$$

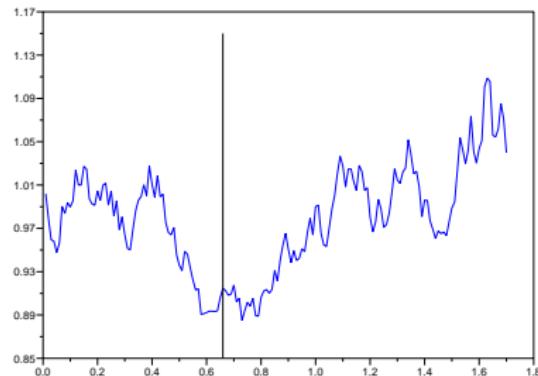
One change of drift

- $\mu(t) = \mu_1$ if $t < \tau$
- $\mu(t) = \mu_2$ if $t \geq \tau$

with $\mathbb{P}(\tau > t) = e^{-\lambda t}$

Strategy

detect τ



$$\mu_1 = -0.2, \mu_2 = 0.2, \sigma = 0.15, \\ \lambda = 2.$$

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- theoretical study of the value function
- theoretical study of detecting the change of drift
- numerical comparisons of strategies
 - well calibrated detection
 - miscalibrated detection
 - moving average

Conclusion

- Moving average strategy can overperform miscalibrated mathematical strategies
- Range of misspecifications for which this is true

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New Model

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- Several changes of drift

$$(\xi_{2n+1}) \text{ iid } \text{Exp}(\lambda_1)$$

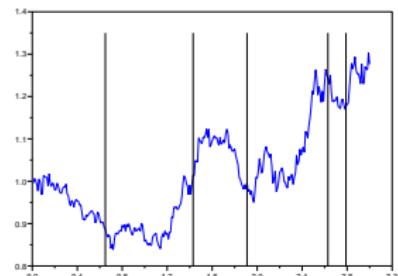
$$(\xi_{2n}) \text{ iid } \text{Exp}(\lambda_2)$$

$$\tau_0 = 0, \tau_n = \xi_1 + \dots + \xi_n$$

$$\mu(t) = \begin{cases} \mu_1 & \text{if } \tau_{2n} \leq t < \tau_{2n+1} \\ \mu_2 & \text{if } \tau_{2n+1} \leq t < \tau_{2n+2} \end{cases}$$

- Transaction costs

- g_{01} buying cost
- g_{10} selling cost



$$\mu_1 = -0.2, \mu_2 = 0.2, \sigma = 0.15, \lambda_1 = \lambda_2 = 2.$$

Admissible strategies

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Control process: $\pi_t \in \{0, 1\}$ proportion of the wealth invested in the risky asset

$$\mathcal{F}_t^S = \sigma(S_u, u \leq t)$$

π_t must be \mathcal{F}_t^S -adapted

Problem

$$\mathcal{F}_t^S \neq \mathcal{F}_t^B = \sigma(B_u, u \leq t)$$

⇒ Change of framework

Filtering Theory

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Optional Projection: $F_t = \mathbb{P}(\mu(t) = \mu_1 | \mathcal{F}_t^S)$

$$\bar{B}_t = \frac{1}{\sigma} \left(\log \frac{S_t}{S_0} - \int_0^t (\mu_1 F_s + \mu_2 (1 - F_s) - \frac{\sigma^2}{2}) ds \right)$$

Martinez, Rubenthaler, Tanré 2005

- \bar{B} is a (\mathcal{F}^S) Brownian motion
- $\mathcal{F}^S = \mathcal{F}^{\bar{B}}$

$$\frac{dS_t}{S_t} = (\mu_1 F_t + \mu_2 (1 - F_t)) dt + \sigma d\bar{B}_t$$

KURTZ, OCONE 1988

$$dF_t = (-\lambda_1 F_t + \lambda_2 (1 - F_t)) dt + \frac{\mu_1 - \mu_2}{\sigma} F_t (1 - F_t) d\bar{B}_t$$

New Framework

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Control process: π_t

State: pair (W_t, F_t)

Dynamics:

$$\frac{dW_t^\pi}{W_{t^-}^\pi} = (\pi_t(\mu_1 F_t + \mu_2(1 - F_t)) + (1 - \pi_t)r) dt + \pi_t \sigma d\bar{B}_t \\ - g_{01}\delta(\Delta\pi_t = 1) - g_{10}\delta(\Delta\pi_t = -1)$$

$$dF_t = (-\lambda_1 F_t + \lambda_2(1 - F_t)) dt + \frac{\mu_1 - \mu_2}{\sigma} F_t(1 - F_t) d\bar{B}_t,$$

Criterion: expected utility of the terminal wealth

Utility: $U(x) = x^\alpha$, $\alpha \in]0, 1[$

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Continuity

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Value Function

$$V^0(t, x, f) = \sup_{\pi} \mathbb{E}[U(W_T^\pi) \mid \pi_{t^-} = 0, W_{t^-}^\pi = x, \mathcal{F}_t = f]$$

$$V^1(t, x, f) = \sup_{\pi} \mathbb{E}[U(W_T^\pi) \mid \pi_{t^-} = 1, W_{t^-}^\pi = x, \mathcal{F}_t = f]$$

Continuity

For all $i \in \{0; 1\}$, $0 \leq t \leq \hat{t} \leq T$, $x, \hat{x} > 0$, $0 \leq f, \hat{f} \leq 1$:

$$\begin{aligned} & |V^i(\hat{t}, \hat{x}, \hat{f}) - V^i(t, x, f)| \\ & \leq C(1 + x^{\alpha-1} + \hat{x}^{\alpha-1})(|\hat{x} - x| + x(|\hat{f} - f| + |\hat{t} - t|^{1/2})) \end{aligned}$$

Continuity

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$$V^1(t, x, f) = \sup_{\pi} \mathbb{E}[U(W_T^\pi) \mid \pi_{t^-} = 1, W_{t^-}^\pi = x, \mathcal{F}_t = f]$$

Continuity

For all $i \in \{0; 1\}$, $0 \leq t \leq \hat{t} \leq T$, $x, \hat{x} > 0$, $0 \leq f, \hat{f} \leq 1$:

$$\begin{aligned} & |V^i(\hat{t}, \hat{x}, \hat{f}) - V^i(t, x, f)| \\ & \leq C(1 + x^{\alpha-1} + \hat{x}^{\alpha-1})(|\hat{x} - x| + x(|\hat{f} - f| + |\hat{t} - t|^{1/2})) \end{aligned}$$

Dynamic Programming Principle

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Dynamic Programming Principle

For all $0 \leq s \leq t \leq T$ and x, f, i :

$$V^i(s, x, f) = \sup_{\pi} \mathbb{E}[V^{\pi_{t-}}(t, W_{t-}^{s,x,f,\pi}, F_t^{s,f})]$$

Proof:

$$\begin{aligned} J^i(s, x, f, \pi) &= \mathbb{E}[U(W_T^{s,x,f,\pi})] = \mathbb{E}[\mathbb{E}[U(W_T^{s,x,f,\pi}) | \mathcal{F}_{s,t}]] \\ &= \mathbb{E}[J^{\pi_{t-}}(t, W_{t-}^{s,x,f,\pi}, F_t^{s,f}, \pi)] \\ &\leq \mathbb{E}[V^{\pi_{t-}}(t, W_{t-}^{s,x,f,\pi}, F_t^{s,f})] \end{aligned}$$

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Dynamic Programming Principle

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Proof:

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Dynamic Programming Principle

For all $0 \leq s \leq t \leq T$ and x, f, i :

$$V^i(s, x, f) = \sup_{\pi} \mathbb{E}[V^{\pi_{t^-}}(t, W_{t^-}^{s, x, f, \pi}, F_t^{s, f})]$$

Proof:

$$\begin{aligned} J^i(s, x, f, \pi) &= \mathbb{E}[U(W_T^{s, x, f, \pi})] = \mathbb{E}[\mathbb{E}[U(W_T^{s, x, f, \pi}) | \mathcal{F}_{s,t}]] \\ &= \mathbb{E}[J^{\pi_{t^-}}(t, W_{t^-}^{s, x, f, \pi}, F_t^{s, f}, \pi)] \\ &\leq \mathbb{E}[V^{\pi_{t^-}}(t, W_{t^-}^{s, x, f, \pi}, F_t^{s, f})] \end{aligned}$$

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$$V^i(s, x, f) = \sup_{\pi} \mathbb{E}[V^{\pi_{t^-}}(t, W_{t^-}^{s, x, f, \pi}, F_t^{s, f})]$$

Proof:

$$\begin{aligned} J^i(s, x, f, \pi) &= \mathbb{E}[U(W_T^{s, x, f, \pi})] = \mathbb{E}[\mathbb{E}[U(W_T^{s, x, f, \pi}) | \mathcal{F}_{s,t}]] \\ &= \mathbb{E}[J^{\pi_{t^-}}(t, W_{t^-}^{s, x, f, \pi}, F_t^{s, f}, \pi)] \\ &\leq \mathbb{E}[V^{\pi_{t^-}}(t, W_{t^-}^{s, x, f, \pi}, F_t^{s, f})] \end{aligned}$$

Dynamic Programming Principle

Proof I

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Fix π such that

$$\mathcal{V} = \sup_{\pi} \mathbb{E}[V^{\pi_{t^-}}(t, W_{t^-}^{s,x,f,\pi}, F_t^{s,f})] \leq \varepsilon + \mathbb{E}[V^{\pi_{t^-}}(t, W_{t^-}^{s,x,f,\pi}, F_t^{s,f})]$$

$(\mathcal{B}_p)_{p \in \mathbb{N}}$ partition of $]0; +\infty[\times [0; 1]$ such that
for all i , (x, f) and (\hat{x}, \hat{f}) in \mathcal{B}_p and for all π :

$$|V^i(t, x, f) - V^i(t, \hat{x}, \hat{f})| \leq \varepsilon, \quad |J^i(t, x, f, \pi) - J^i(t, \hat{x}, \hat{f}, \pi)| \leq \varepsilon$$

$$\mathcal{V} \leq \varepsilon + \sum_{p=0}^{\infty} \mathbb{E}[V^{\pi_{t^-}}(t, W_{t^-}^{s,x,f,\pi}, F_t^{s,f}) \mathbf{1}_{(W_{t^-}^{s,x,f,\pi}, F_t^{s,f}) \in \mathcal{B}_p}]$$

fix (x_p, f_p) in \mathcal{B}_p

$$\mathcal{V} \leq \varepsilon + \varepsilon + \sum_{p=0}^{\infty} \mathbb{E}[V^{\pi_{t^-}}(t, x_p, f_p) \mathbf{1}_{(W_{t^-}^{s,x,f,\pi}, F_t^{s,f}) \in \mathcal{B}_p}]$$

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fix (x_p, f_p) in \mathcal{B}_p

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For fixed p, i , let $\pi^{p,i}$ be a strategy on $[t, T]$ such that

$$V^i(t, x_p, f_p) \leq \varepsilon + J^i(t, x_p, f_p, \pi^{p,i})$$

$$\begin{aligned} \mathcal{V} &\leq \varepsilon + \varepsilon + \varepsilon + \sum_{p=0}^{\infty} \mathbb{E}[J^{\pi_{t^-}}(t, x_p, f_p, \pi^{p,\pi_{t^-}}) \mathbf{1}_{(W_{t^-}^{s,x,f,\pi}, F_t^{s,f}) \in \mathcal{B}_p}] \\ &\leq \varepsilon + 2\varepsilon + \varepsilon \\ &\quad + \sum_{p=0}^{\infty} \mathbb{E}[J^{\pi_{t^-}}(t, W_{t^-}^{s,x,f,\pi}, F_t^{s,f}, \pi^{p,\pi_{t^-}}) \mathbf{1}_{(W_{t^-}^{s,x,f,\pi}, F_t^{s,f}) \in \mathcal{B}_p}] \end{aligned}$$

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$$\mathcal{V} \leq \varepsilon + \varepsilon + \sum_{p=0}^{\infty} \mathbb{E}[V^{\pi_{t^-}}(t, x_p, f_p) \mathbf{1}_{(W_{t^-}^{s,x,f,\pi}, F_t^{s,f}) \in \mathcal{B}_p}]$$

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$$\mathcal{V} \leq \varepsilon + \varepsilon + \sum_{p=0}^{\infty} \mathbb{E}[V^{\pi_{t^-}}(t, x_p, f_p) \mathbf{1}_{(W_{t^-}^{s,x,f,\pi}, F_t^{s,f}) \in \mathcal{B}_p}]$$

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$$\mathcal{V} \leq 4\varepsilon + \sum_{p=0}^{\infty} \mathbb{E}[J^{\pi_{t^-}}(t, W_{t^-}^{s,x,f,\pi}, F_t^{s,f}, \pi^{p,\pi_{t^-}}) \mathbf{1}_{(W_{t^-}^{s,x,f,\pi}, F_t^{s,f}) \in \mathcal{B}_p}]$$

Combining step

$$\hat{\pi}_u = \begin{cases} \pi_u & \text{if } s \leq u < t \\ \pi_u^{p,\hat{\pi}_{t^-}} & \text{if } u \geq t, \text{ and } (W_{t^-}^{s,x,f,\hat{\pi}}, F_t^{s,f}) \in \mathcal{B}_p \end{cases}$$

$$\begin{aligned} \mathcal{V} &\leq 4\varepsilon + \mathbb{E}[J^{\hat{\pi}_{t^-}}(t, W_{t^-}^{s,x,f,\hat{\pi}}, F_t^{s,f}, \hat{\pi})] \\ &= 4\varepsilon + \mathbb{E}[\mathbb{E}[U(W_T^{s,x,f,\hat{\pi}}) | \mathcal{F}_{s,t}]] \\ &= 4\varepsilon + \mathbb{E}[U(W_T^{s,x,f,\hat{\pi}})] \\ &\leq 4\varepsilon + \mathbb{E}[V^i(s, x, f)] \end{aligned}$$

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$$\mathcal{V} \leq 4\varepsilon + \sum_{p=0}^{\infty} \mathbb{E}[J^{\pi_{t^-}}(t, W_{t^-}^{s,x,f,\pi}, F_t^{s,f}, \pi^{p,\pi_{t^-}}) \mathbf{1}_{(W_{t^-}^{s,x,f,\pi}, F_t^{s,f}) \in \mathcal{B}_p}]$$

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$$\begin{aligned} \mathcal{V} &\leq 4\varepsilon + \mathbb{E}[J^{\hat{\pi}_{t^-}}(t, W_{t^-}^{s,x,f,\hat{\pi}}, F_t^{s,f}, \hat{\pi})] \\ &= 4\varepsilon + \mathbb{E}[\mathbb{E}[U(W_T^{s,x,f,\hat{\pi}}) | \mathcal{F}_{s,t}]] \\ &= 4\varepsilon + \mathbb{E}[U(W_T^{s,x,f,\hat{\pi}})] \\ &\leq 4\varepsilon + \mathbb{E}[V^i(s, x, f)] \end{aligned}$$

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$$\mathcal{V} \leq 4\varepsilon + \sum_{p=0}^{\infty} \mathbb{E}[J^{\pi_{t^-}}(t, W_{t^-}^{s,x,f,\pi}, F_t^{s,f}, \pi^{p,\hat{\pi}_{t^-}}) \mathbf{1}_{(W_{t^-}^{s,x,f,\pi}, F_t^{s,f}) \in \mathcal{B}_p}]$$

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$$\begin{cases} \min \left\{ -\frac{\partial \varphi^0}{\partial t} - \mathcal{L}^0 \varphi^0; \varphi^0(t, x, f) - \varphi^1(t, x(1 - g_{01}), f) \right\} = 0 \\ \min \left\{ -\frac{\partial \varphi^1}{\partial t} - \mathcal{L}^1 \varphi^1; \varphi^1(t, x, f) - \varphi^0(t, x(1 - g_{10}), f) \right\} = 0 \end{cases}$$

$$\begin{aligned} \mathcal{L}^0 \varphi(t, x, f) &= xr \frac{\partial \varphi}{\partial x}(t, x, f) + (-\lambda_1 f + \lambda_2(1 - f)) \frac{\partial \varphi}{\partial f}(t, x, f) + \\ &\quad \frac{1}{2} \left(\frac{\mu_1 - \mu_2}{\sigma} \right)^2 f^2 (1 - f)^2 \frac{\partial^2 \varphi}{\partial f^2}(t, x, f) \end{aligned}$$

$$\begin{aligned} \mathcal{L}^1 \varphi(t, x, f) &= x(\mu_1 f + \mu_2(1 - f)) \frac{\partial \varphi}{\partial x}(t, x, f) + \frac{1}{2} x^2 \sigma^2 \frac{\partial^2 \varphi}{\partial x^2}(t, x, f) \\ &\quad + (-\lambda_1 f + \lambda_2(1 - f)) \frac{\partial \varphi}{\partial f}(t, x, f) + x(\mu_1 - \mu_2)f(1 - f) \frac{\partial^2 \varphi}{\partial x \partial f}(t, x, f) \\ &\quad + \frac{1}{2} \left(\frac{\mu_1 - \mu_2}{\sigma} \right)^2 f^2 (1 - f)^2 \frac{\partial^2 \varphi}{\partial f^2}(t, x, f) \end{aligned}$$

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$$(P) \quad F(t, x, v(t, x), D_t v(t, x), Dv(t, x), D^2 v(t, x)) = 0$$

Definition

- v is a **viscosity sub-solution** of (P) if

$$F(\bar{t}, \bar{x}, v(\bar{t}, \bar{x}), D_t \varphi(\bar{t}, \bar{x}), D\varphi(\bar{t}, \bar{x}), D^2 \varphi(\bar{t}, \bar{x})) \leq 0$$

for all (\bar{t}, \bar{x}) and all functions $\varphi \in C^{1,2}$ such that (\bar{t}, \bar{x}) is a local maximum of $v - \varphi$

- v is a **viscosity super-solution** of (P) if

$$F(\bar{t}, \bar{x}, v(\bar{t}, \bar{x}), D_t \varphi(\bar{t}, \bar{x}), D\varphi(\bar{t}, \bar{x}), D^2 \varphi(\bar{t}, \bar{x})) \geq 0$$

for all (\bar{t}, \bar{x}) and all functions $\varphi \in C^{1,2}$ such that (\bar{t}, \bar{x}) is a local minimum of $v - \varphi$

Identification of the Value Function

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\mathcal{V}_α : set of continuous functions φ on $[0; T] \times [0; +\infty[\times [0; 1]$ satisfying $\varphi(t, 0, f) = 0$ and

$$\sup_{[0;T] \times]0;+\infty[^2 \times [0;1]^2} \frac{|\varphi(t, x, f) - \varphi(t, \hat{x}, \hat{f})|}{(1 + x^{\alpha-1} + \hat{x}^{\alpha-1})(|x - \hat{x}| + |f - \hat{f}|)} < \infty.$$

Theorem

(V^0, V^1) is the unique viscosity solution of HJB equation in $\mathcal{V}_\alpha \times \mathcal{V}_\alpha$ satisfying

$$V^0(T, x, f) = V^1(T, x, f) = U(x) = x^\alpha$$

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Dependence on x :

$$V^i(t, x, f) = \sup_{\pi} \mathbb{E}[U(W_T^{t,x,f,\pi})] = x^\alpha V^i(t, 1, f)$$

Numerical Scheme

- $\hat{V}^0(T, f) = \hat{V}^1(T, f) = 1$
- With the PDE part in HJB, compute $\overline{V^0}(t, \cdot)$ et $\overline{V^1}(t, \cdot)$ from $\hat{V}^0(t + dt, \cdot)$ and $\hat{V}^1(t + dt, \cdot)$
- Comparison
 - if $\overline{V^0}(t, f) \geq (1 - g_{01})^\alpha \overline{V^1}(t, f)$, set $\hat{V}^0(t, f) = \overline{V^0}(t, f)$
otherwise $\hat{V}^0(t, f) = (1 - g_{01})^\alpha \overline{V^1}(t, f)$
 - if $\overline{V^1}(t, f) \geq (1 - g_{10})^\alpha \overline{V^0}(t, f)$, set $\hat{V}^1(t, f) = \overline{V^1}(t, f)$
otherwise $\hat{V}^1(t, f) = (1 - g_{10})^\alpha \overline{V^0}(t, f)$

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Dependence on x :

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- Comparison
 - if $\overline{V^0}(t, f) \geq (1 - g_{01})^\alpha \overline{V^1}(t, f)$, set $\hat{V}^0(t, f) = \overline{V^0}(t, f)$
otherwise $\hat{V}^0(t, f) = (1 - g_{01})^\alpha \overline{V^1}(t, f)$
 - if $\overline{V^1}(t, f) \geq (1 - g_{10})^\alpha \overline{V^0}(t, f)$, set $\hat{V}^1(t, f) = \overline{V^1}(t, f)$
otherwise $\hat{V}^1(t, f) = (1 - g_{10})^\alpha \overline{V^0}(t, f)$

Discretization Scheme

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Dependence on x :

$$V^i(t, x, f) = \sup_{\pi} \mathbb{E}[U(W_T^{t,x,f,\pi})] = x^\alpha V^i(t, 1, f)$$

Numerical Scheme

- $\hat{V}^0(T, f) = \hat{V}^1(T, f) = 1$
- With the PDE part in HJB, compute $\overline{V^0}(t, \cdot)$ et $\overline{V^1}(t, \cdot)$ from $\hat{V}^0(t + dt, \cdot)$ and $\hat{V}^1(t + dt, \cdot)$
- Comparison
 - if $\overline{V^0}(t, f) \geq (1 - g_{01})^\alpha \overline{V^1}(t, f)$, set $\hat{V}^0(t, f) = \overline{V^0}(t, f)$
otherwise $\hat{V}^0(t, f) = (1 - g_{01})^\alpha \overline{V^1}(t, f)$
 - if $\overline{V^1}(t, f) \geq (1 - g_{10})^\alpha \overline{V^0}(t, f)$, set $\hat{V}^1(t, f) = \overline{V^1}(t, f)$
otherwise $\hat{V}^1(t, f) = (1 - g_{10})^\alpha \overline{V^0}(t, f)$

Value Function V^0

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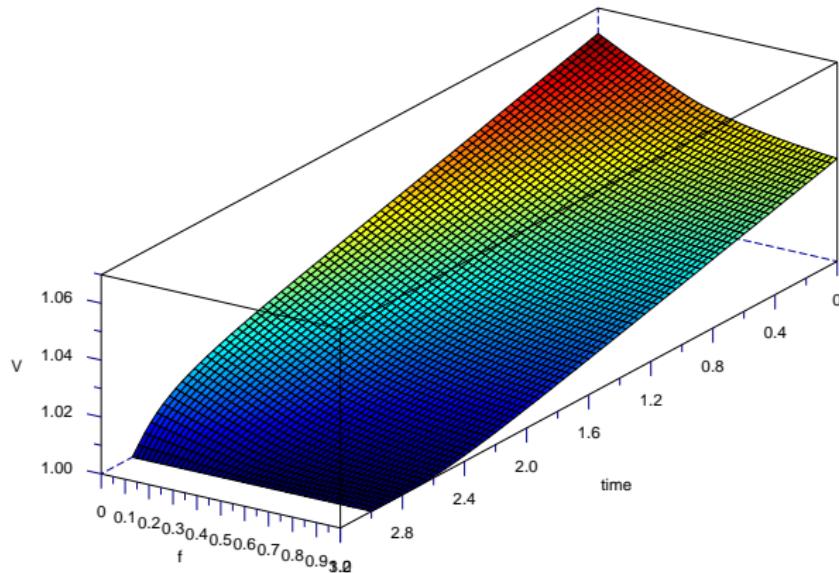
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Parameters: $T = 3$, $\mu_2 = -\mu_1 = 0.2$, $\lambda_1 = \lambda_2 = 2$, $\sigma = 0.15$,
 $g_{01} = g_{10} = 0.001$



Value Function V^0

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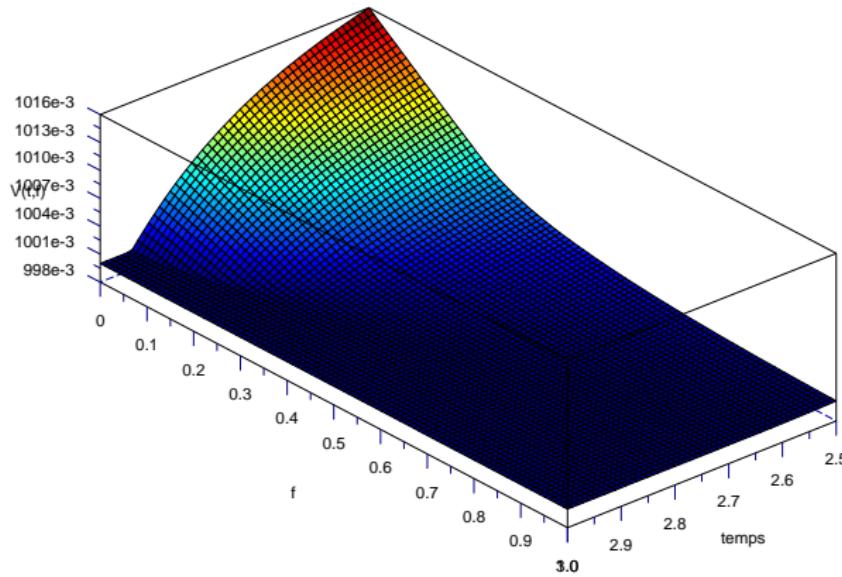
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Transaction costs $g_{01} = g_{10} = 0.01$
Zoom between $t = 2.5$ and $t = 3 = T$



Value Function V^0

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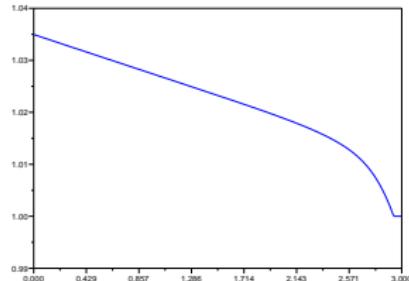
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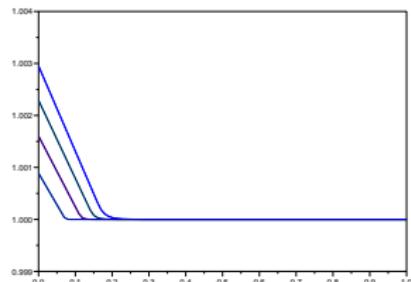
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Section at $f = 0.05$



Sections at $t = 2.90,$
 $t = 2.91, t = 2.92, t = 2.93,$



Efficient Strategy

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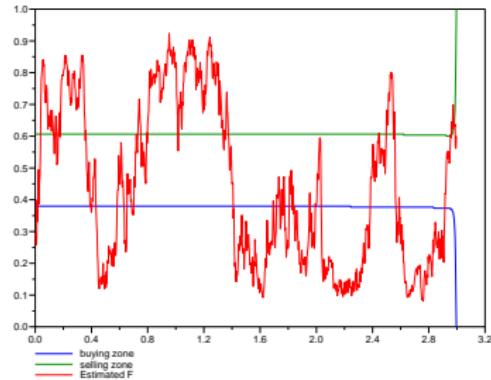
- Compute \hat{V}^0, \hat{V}^1
- Estimate \hat{F}_t from the stock
- Compare $\hat{V}^0(t, \hat{F}_t)$ et $\hat{V}^1(t, \hat{F}_t)$:

- buy if

$$\hat{V}^0(t, \hat{F}_t) = (1 - g_{01})^\alpha \hat{V}^1(t, \hat{F}_t)$$

- sell if

$$\hat{V}^1(t, \hat{F}_t) = (1 - g_{10})^\alpha \hat{V}^0(t, \hat{F}_t)$$



$$\begin{aligned}\mu_1 &= -0.2, \mu_2 = 0.2, \sigma = 0.15, \\ \lambda_1 &= 2, \lambda_2 = 2, T = 3\end{aligned}$$

Efficient Strategy vs Value Function

- Computation of the value function:

- time discretization step 10^{-6}
- space discretization step 10^{-3}

- 10^5 Monte Carlo simulations of the Efficient Strategy

F_0	0	0.1	0.2	0.3	0.4	0.5
\hat{V}^0	1.061	1.057	1.053	1.049	1.045	1.043
Strategy	1.061	1.056	1.052	1.049	1.045	1.043

F_0	0.6	0.7	0.8	0.9	1
\hat{V}^0	1.041	1.039	1.038	1.037	1.036
Strategy	1.040	1.039	1.038	1.037	1.036

Miscalibrated Efficient Strategy vs Moving Average

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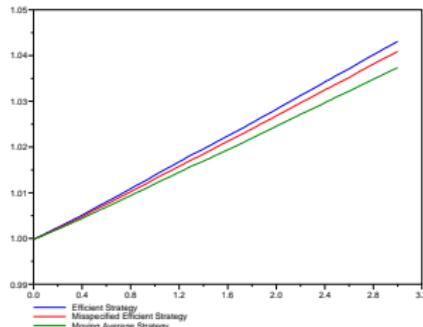
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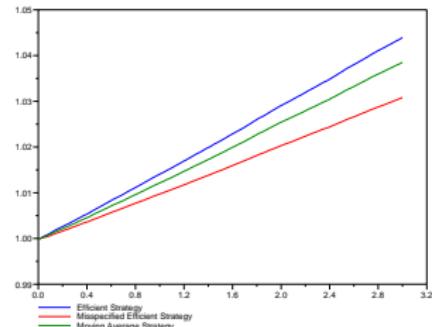
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miscalibrated parameters:
 $\mu_1 = -1.8, \mu_2 = 1.8, \sigma = 0.15,$
 $\lambda_1 = 4, \lambda_2 = 4$



miscalibrated parameters:
 $\mu_1 = -1.8, \mu_2 = 1.8, \sigma = 0.25,$
 $\lambda_1 = 4, \lambda_2 = 4$

Real parameters: $\mu_1 = -0.2, \mu_2 = 0.2, \sigma = 0.15, \lambda_1 = 2,$
 $\lambda_2 = 2, T = 3, \delta = 0.8$
100000 Monte Carlo Simulations