

Algorithmic Trading: Optimal Control

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Stochastic Optimal Control

- ▶ **Stochastic Optimal Control** is concerned with **maximizing / minimizing** a **performance criteria** where the criteria is affected by **future unknown noise** in the system, as well as the **actions** of the controller / agent

Stochastic Optimal Control

- ▶ We aim to solve the problem

$$H(x) = \sup_{u \in \mathcal{A}} \mathbb{E} \left[\underbrace{G(X_T^{x,u})}_{\text{terminal reward}} + \underbrace{\int_0^T F(s, X_s^{x,u}, u_s) ds}_{\text{running reward/penalty}} \right]$$

where

- ▶ $u = (u_t)_{t \geq 0}$ is the **control process** and the agent chooses it
- ▶ \mathcal{A} is the **admissible set** of controls – e.g., exclude doubling strategies
- ▶ $X^u = (X_t^u)_{t \geq 0}$ is the **controlled process**, which the agent partially controls, and satisfies the SDE

$$dX_t^u = \mu(t, X_t^u, u_t) dt + \sigma(t, X_t^u, u_t) dW_t, \quad X_0 = x$$

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- ▶ Trick is to introduce a larger class of problems indexed by time
- ▶ The **value function** is defined as

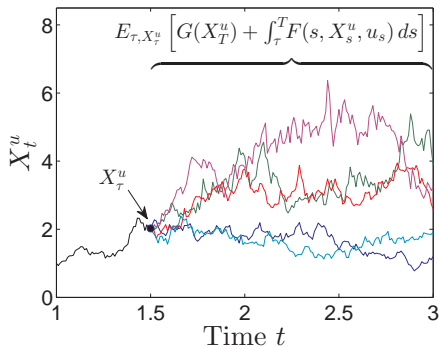
$$H(t, x) = \sup_{u \in \mathcal{A}} \mathbb{E}_{t,x} \left[G(X_T^u) + \int_t^T F(s, X_s^u, u_s) ds \right]$$

where $\mathbb{E}_{t,x}[\cdot]$ means expectation conditional on $X_t = x$.

- ▶ Prove a **dynamic programming principle** and the corresponding **dynamic programming equation**

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We will “flow” an arbitrary admissible control and re-write the value function recursively...



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- Take an **arbitrary admissible control** $u \in \mathcal{A}$

$$\begin{aligned} H^u(t, x) &= \mathbb{E}_{t,x} \left[G(X_T^u) + \int_t^T F(s, X_s^u, u_s) ds \right] \\ &= \mathbb{E}_{t,x} \left[G(X_T^u) + \int_\tau^T F(s, X_s^u, u_s) ds + \int_t^\tau F(s, X_s^u, u_s) ds \right] \\ &= \mathbb{E}_{t,x} \left[\mathbb{E}_{\tau, X_\tau^u} \left[G(X_T^u) + \int_\tau^T F(s, X_s^u, u_s) ds \right] + \int_t^\tau F(s, X_s^u, u_s) ds \right] \\ &\quad \text{(by iterated expectation)} \\ &= \mathbb{E}_{t,x} \left[H^u(\tau, X_\tau^u) + \int_t^\tau F(s, X_s^u, u_s) ds \right] \quad \text{(by defn)} \end{aligned}$$

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- ▶ However, $H(t, x) \geq H^u(t, x)$, with equality if $u = u^*$, hence

$$\begin{aligned} H^u(t, x) &\leq \mathbb{E}_{t,x} \left[H(\tau, X_\tau^u) + \int_t^\tau F(s, X_s^u, u_s) ds \right] \\ &\leq \sup_{u \in \mathcal{A}} \mathbb{E}_{t,x} \left[H(\tau, X_\tau^u) + \int_t^\tau F(s, X_s^u, u_s) ds \right] \end{aligned}$$

and so

$$H(t, x) \leq \sup_{u \in \mathcal{A}} \mathbb{E}_{t,x} \left[H(\tau, X_\tau^u) + \int_t^\tau F(s, X_s^u, u_s) ds \right]$$

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- ▶ Take an ε -**optimal control** v^ε – a control that performs better than $H(t, x) - \varepsilon$, but of course not as good as $H(t, x)$, i.e., such that

$$H(t, x) \geq H^{v^\varepsilon}(t, x) \geq H(t, x) - \varepsilon$$

- ▶ Modify the ε -optimal control between t and τ by an arbitrary control u , i.e., define \tilde{v}^ε by

$$\tilde{v}_t^\varepsilon := u_t \mathbb{1}_{t \leq \tau} + v_t^\varepsilon \mathbb{1}_{t > \tau}$$

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- Then,

$$\begin{aligned} H(t, x) &\geq H^{\tilde{v}^\varepsilon}(t, x) \\ &= \mathbb{E}_{t,x} \left[H^{\tilde{v}^\varepsilon}(\tau, X_\tau^{\tilde{v}^\varepsilon}) + \int_t^\tau F(s, X_s^{\tilde{v}^\varepsilon}, \tilde{v}_s^\varepsilon) ds \right] && \text{by iterated expectations} \\ &= \mathbb{E}_{t,x} \left[H^{v^\varepsilon}(\tau, X_\tau^u) + \int_t^\tau F(s, X_s^u, u_s) ds \right] && \text{by the modified strategy} \\ &\geq \mathbb{E}_{t,x} \left[H(\tau, X_\tau^u) - \varepsilon + \int_t^\tau F(s, X_s^u, u_s) ds \right] && \varepsilon\text{-optimal control} \end{aligned}$$

- Hence, taking $\varepsilon \downarrow 0$, and since u is arbitrary

$$H(t, x) \geq \sup_{u \in \mathcal{A}} \mathbb{E}_{t,x} \left[H(\tau, X_\tau^u) + \int_t^\tau F(s, X_s^u, u_s) ds \right]$$

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- ▶ Putting both inequalities together we arrive at the **dynamic programming principle (DPP)**

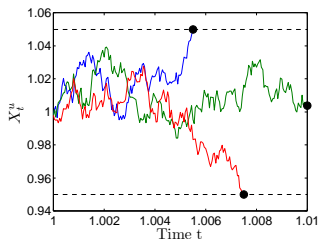
$$H(t, x) = \sup_{u \in \mathcal{A}} \mathbb{E}_{t, x} \left[H(\tau, X_\tau^u) + \int_t^\tau F(s, X_s^u, u_s) ds \right]$$

- ▶ The **dynamic programming equation (DPE)** is the infinitesimal version of this principle

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- Take τ in the DPP to be the following

$$\tau = T \wedge \inf \{s > t : (s - t, |X_s^u - x|) \notin [0, h) \times [0, \epsilon)\}.$$



- that is, either
 1. an amount of time h passes, and the processes has deviated by less than ϵ , or
 2. the process deviates by ϵ and we stop

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- From the **DPP**

$$H(t, x) \geq \sup_{u \in \mathcal{A}} \mathbb{E}_{t,x} \left[H(\tau, X_\tau^u) + \int_t^\tau F(s, X_s^u, u_s) ds \right]$$

- Hence, for a **constant strategy** v on the interval $[t, \tau]$

$$H(t, x) \geq \mathbb{E}_{t,x} \left[H(\tau, X_\tau^v) + \int_t^\tau F(s, X_s^v, v) ds \right]$$

- Applying **Itô's lemma** to the value function

$$\begin{aligned} H(\tau, X_\tau^v) &= H(t, X_t) + \int_t^\tau (\partial_t + \mathcal{L}_s^v) H(s, X_s^v) ds \\ &\quad + \int_t^\tau \partial_x H(s, X_s^v) \sigma(s, X_s^v, v) dW_s, \end{aligned}$$

where the generator is

$$\mathcal{L}_t^v = \mu(t, x, v) \partial_x + \frac{1}{2} \sigma^2(t, x, v) \partial_{xx}$$

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- We then have,

$$H(t, x) \geq \mathbb{E}_{t,x} \left[H(t, X_t) + \int_t^\tau (\partial_t + \mathcal{L}_s^v) H(s, X_s^v) ds + \int_t^\tau \partial_x H(s, X_s^v) \sigma(s, X_s^v, v) dW_s + \int_t^\tau F(s, X_s^v, v) ds \right]$$

- Since $|X_t^v - x| < \epsilon$ on $[t, \tau]$, the stochastic integral is indeed a martingale and we have

$$H(t, x) \geq \mathbb{E}_{t,x} \left[H(t, x) + \int_t^\tau \left\{ (\partial_t + \mathcal{L}_s^v) H(s, X_s^v) + F(s, X_s^v, v) \right\} ds \right]$$

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► Then,

$$\begin{aligned} 0 &\geq \lim_{h \downarrow 0} \mathbb{E}_{t,x} \left[\frac{1}{h} \int_t^\tau \left\{ (\partial_t + \mathcal{L}_s^v) H(s, X_s^v) + F(s, X_s^v, v) \right\} ds \right] \\ &= (\partial_t + \mathcal{L}_t^v) H(t, x) + F(t, x, v) \end{aligned}$$

which follows b/c

- (i) as $h \searrow 0$, $\tau = t + h$ a.s. since the process will not hit the barrier of ϵ in extremely short periods of time,
- (ii) the condition that $|X_\tau^v - x| \leq \epsilon$, which implies that if the process does hit the barrier it is bounded,
- (iii) the Mean-Value Theorem allows us to write $\lim_{h \downarrow 0} \frac{1}{h} \int_t^{t+h} \omega_s ds = \omega_t$, and
- (iv) the process starts at $X_t^v = x$.

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- ▶ This inequality holds for every constant v , and therefore

$$\partial_t H(t, x) + \sup_v \{ \mathcal{L}_t^v H(t, x) + F(t, x, v) \} \leq 0$$

Stochastic Optimal Control

- ▶ Next, we show the opposite inequality, i.e., that

$$\partial_t H(t, x) + \sup_v \{ \mathcal{L}_t^v H(t, x) + F(t, x, v) \} \geq 0$$

- ▶ We do this by **contradiction**... assume $\exists (t_0, x_0)$ s.t.,

$$\partial_t H(t_0, x_0) + \sup_v \{ \mathcal{L}_t^v H(t_0, x_0) + F(t_0, x_0, v) \} < 0 \quad (1)$$

- ▶ Then, define a modification φ of the value function H via

$$\varphi(t, x) = H(t, x) + \epsilon \left((t - t_0)^2 + (x - x_0)^4 \right)$$

this lies above the value function, but equals it at (t_0, x_0) , and is differentiable enough to apply Itô's lemma

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- Note also that

$$\begin{aligned}\partial_t \varphi(t_0, x_0) &= \partial_t H(t_0, x_0), \\ \partial_x \varphi(t_0, x_0) &= \partial_x H(t_0, x_0), \\ \partial_{xx} \varphi(t_0, x_0) &= \partial_{xx} H(t_0, x_0)\end{aligned}$$

- Therefore, from (1), we have that

$$\partial_t \varphi(t_0, x_0) + \sup_v \{ \mathcal{L}_t^v \varphi(t_0, x_0) + F(t_0, x_0, v) \} < 0$$

and if the Hamiltonian $\sup_v \{ \mathcal{L}_t^v H(t, x) + F(t, x, v) \}$ is continuous, then

$$\exists \text{ a neighbourhood } \mathcal{N}_r = (t_0 - r, t_0 + r) \times (x_0 - r, x_0 + r)$$

s.t.

$$\partial_t \varphi(t, x) + \sup_v \{ \mathcal{L}_t^v \varphi(t, x) + F(t, x, v) \} < 0 \quad (2)$$

for all $(t, x) \in \mathcal{N}_r$

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- Define

$$\eta = \max_{(t,x) \in \partial \mathcal{N}_r} (\varphi - H)(t, x) > 0$$

- Take an arbitrary control $u \in \mathcal{A}$ and define the stopping time

$$\tau = \inf\{s > t_0 : X_s^u \notin \mathcal{N}_r\}$$

- Since X is continuous

$$X_\tau^u \in \partial \mathcal{N}_r$$

and therefore

$$\varphi(\tau, X_\tau^u) \geq \eta + H(\tau, X_\tau^u) \quad (3)$$

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- Apply Itô's lemma to φ to find

$$\varphi(\tau, X_\tau^u) = \varphi(t_0, x_0) + \int_{t_0}^{\tau} (\partial_t + \mathcal{L}_t^u) \varphi(s, X_s^u) ds + \int_{t_0}^{\tau} \partial_x \varphi(s, X_s^u) \sigma(s, X_s^u, u) dW_s$$

- Therefore,

$$\begin{aligned} V(t_0, x_0) &= \varphi(t_0, x_0) \\ &= \mathbb{E}_{t_0, x_0} \left[\varphi(\tau, X_\tau^u) - \int_{t_0}^{\tau} (\partial_t + \mathcal{L}_t^u) \varphi(s, X_s^u) ds \right] \end{aligned}$$

- From (2), for $(t, x) \in \mathcal{N}_r$

$$\psi(t, x) = \partial_t \varphi(t, x) + \sup_v \{ \mathcal{L}_t^v \varphi(t, x) + F(t, x, v) \} < 0$$

so that

$$\partial_t \varphi(t, x) + \mathcal{L}_t^u \varphi(t, x) + F(t, x, u_t) \leq \psi(t, x) < 0$$

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- Therefore, we have

$$\begin{aligned} V(t_0, x_0) &= \varphi(t_0, x_0) \\ &= \mathbb{E}_{t_0, x_0} \left[\varphi(\tau, X_\tau^u) - \int_{t_0}^{\tau} (\partial_t + \mathcal{L}_t^u) \varphi(s, X_s^u) ds \right] \\ &\geq \mathbb{E}_{t_0, x_0} \left[\varphi(\tau, X_\tau^u) + \int_{t_0}^{\tau} (F(s, X_s^u, u_s) - \psi(s, X_s^u)) ds \right] \end{aligned}$$

- On boundary \mathcal{N}_r , φ dominates H , i.e. from (3), we have

$$\begin{aligned} V(t_0, x_0) &\geq \mathbb{E}_{t_0, x_0} \left[\eta + H(\tau, X_\tau^u) + \int_{t_0}^{\tau} (F(s, X_s^u, u_s) - \psi(s, X_s^u)) ds \right] \quad (\text{since } \psi < 0) \\ &\geq \eta + \mathbb{E}_{t_0, x_0} \left[H(\tau, X_\tau^u) + \int_{t_0}^{\tau} F(s, X_s^u, u_s) ds \right] \\ &> \mathbb{E}_{t_0, x_0} \left[H(\tau, X_\tau^u) + \int_{t_0}^{\tau} F(s, X_s^u, u_s) ds \right] \end{aligned}$$

This **violates the DPP!**

Stochastic Optimal Control

Hence we obtain the **Dynamic Programming Equation** (DPE), aka **Hamilton-Jacobi-Bellman** (HJB) equation

$$\partial_t H(t, x) + \sup_v \{ \mathcal{L}_t^v H(t, x) + F(t, x, v) \} = 0$$

- ▶ This is a **non-linear PDE** that the **value function must satisfy**
- ▶ It is not clear that if we solve this PDE, the solution is the value function!
- ▶ This requires a **verification theorem**