

Lecture 11, Optimal Stopping and Free-bdr Problems

Let $T > 0$, assume we hold a stock S_t , when do we sell it, before T ?



Problem: we can't tell if it's the highest point.

$$\{ S_t = \max_{0 \leq \tau \leq T} S_\tau \} \notin \mathcal{F}_t.$$

More realistically: set a barrier b , once $S_t \geq b$, sell.

Let $T > 0$ and X_t be a stochastic process, an optimal stopping problem is to maximise $\mathbb{E}_{t,x} [g(X_\tau)]$ over some sets of stopping times, where

g is called the gain function. Write

$$V(t, x) = \sup_{\tau \in \mathcal{T}} \mathbb{E}_{t,x} [g(X_\tau)] = \mathbb{E}_{t,x} [g(X_{\tau^*})]$$

value fun. time-dependent!
set of admissible τ 's
optimal strategy (if exists).

- Goal:
- i) Find V . — How much do we get.
 - ii) Find τ^* . — When do we stop.

We consider: i) $T = \infty$, so V is time-independent.

ii) X_t is a time-homogeneous Ito diffusion:

$$dX_t = \mu(X_t) dt + \sigma(X_t) dB_t.$$

Some applications

i). Stochastic analysis.

Let B_t be a 1D-BM. $B_0 = x$. we consider

$$V(x) = \sup_{\tau} \mathbb{E}_x [e^{-r\tau} f(B_\tau)].$$

Consider e.g. $f(x) = x^2$ (HW).

Can't wait forever, when $|B|$ is big enough we stop.

ii). Sequential analysis

Suppose we observe a BM with drift:

$$X_t = \mu t + B_t.$$

where μ is a Bernoulli r.v. $\mathbb{P}(\mu = \mu_0) = p$, $\mathbb{P}(\mu = \mu_1) = 1-p$.

Then to make inference, we test

$$H_0 = \mu = \mu_0, \quad H_1 = \mu = \mu_1.$$

We want to minimize the probability of making an mistake,

\Rightarrow Formulate an optimal stopping problem, can be solved explicitly.

(Shiryaev, 1969). \swarrow inf problem.

★ iii) Mathematical finance.

American put option: Can choose to exercise at any time before expiration T , and get $g(S_\tau) = (K - S_\tau)^+$.
↳ exercising time.

Want: maximize the discounted payoff; thus the price at time t should be.

$$V(t, x) = \sup_{\tau \in [t, T]} \mathbb{E}_{t, x}^Q \left[e^{-r(\tau-t)} (K - S_\tau)^+ \right].$$

we will solve this next time! (w. $T = \infty$).

How to find V and τ^* ?

Example 11.1 (When do we sell the stock?).

Let X_t be a GBM. $dX_t = \mu X_t dt + \sigma X_t dB_t$, $X_0 = x$.

Assume i) Trading is not free: transaction cost $c > 0$.

ii) Rate of discounting $\beta > \mu$: risk-free, tax, inflation ...

Want: Identify a stopping time τ^* s.t.

$$\mathbb{E}_{0, x} \left[e^{-\beta \tau} (X_\tau - c) \right] \text{ is maximised.}$$

($T = \infty$, starting at $t=0 \Rightarrow$ value fun is time-homogeneous!)

$$V(x) = \sup_{\tau} \mathbb{E}_{0,x} [e^{-\beta \tau} (X_{\tau} - a)]$$

Intuition There must be some threshold $b > 0$. s.t.

If $x \geq b$, sell immediately. $\rightarrow \tau^* = 0$
 $b > c$, why?

If $x < b$, wait.
until when?

- Let $\tau^* := \inf \{ t \geq 0 : X_t \geq b \}$, then this first exit time should be the optimal stopping time.
- What is the boundary b ? *Unknown*. "Free-boundary problem".

Question Can V be characterised by a PDE?

i) When $x \geq b$, $\tau^* = 0$, get immediate payoff.

$$V(x) = x - c.$$

ii) when $x < b$. Since X continuous, X_t will stay below b for $t \in (0, h)$. for h small. Therefore, at time 0:

$$V(x) = e^{-\beta h} \mathbb{E}_{0,x} [\underbrace{V(X_h)}_{\text{r.v.}}]$$

By Feynman-Kac, V "should" solve.

$$\begin{cases} V_t + \mathcal{L}V - \beta V = 0, & \text{on } (0, b) \\ \downarrow \\ = 0 & V(b) = b - c. \\ & \text{discontinuity.} \end{cases}$$

This yields.

$$\left\{ \begin{array}{l} \mu x V_x + \frac{1}{2} \sigma^2 x^2 V_{xx} - \beta V = 0, \quad \text{on } (0, b). \\ V(b) = b - c. \\ V(x) = x - c, \quad \text{on } [b, \infty). \end{array} \right.$$

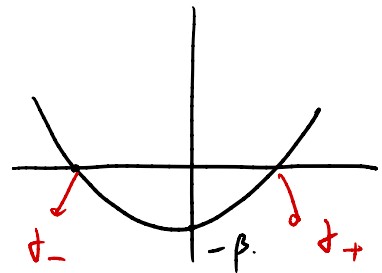
\swarrow
unknown.
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unknown.

Ansatz $V(x) = x^\delta$ seems useful. plug it in:

$$\mu \delta x^{\delta-1} + \frac{1}{2} \delta(\delta-1) \sigma^2 x^{\delta-2} - \beta x^{\delta-1} = 0.$$

$$\Rightarrow \frac{1}{2} \sigma^2 \delta^2 + (\mu - \frac{1}{2} \sigma^2) \delta - \beta = 0$$

$\underbrace{\hspace{2em}}_{>0}$
 $\underbrace{\hspace{2em}}_{<0}$



Denote the roots by $\delta_- < 0 < \delta_+$:

$$\delta_{\pm} = \frac{(\frac{1}{2} \sigma^2 - \mu) \pm \sqrt{(\mu - \frac{1}{2} \sigma^2)^2 + 2\beta \sigma^2}}{\sigma^2}$$

A general solution is then:

$$V(x) = C_1 x^{\delta_+} + C_2 x^{\delta_-}$$

However, as $x \rightarrow 0$, $V(x) \rightarrow 0 \Rightarrow C_2 = 0$.

$$\Rightarrow V(x) = C_1 x^{\delta_+} \quad \text{for } x \in (0, b).$$

$$\text{At } b: V(b) = b - c \Rightarrow C_1 = \frac{b - c}{b^{\delta_+}}$$

$$\Rightarrow V(x) = \begin{cases} \frac{b - c}{b^{\delta_+}} x^{\delta_+}, & x \in (0, b). \\ b - c, & x \geq b. \end{cases}$$

Problem: b is still unknown!

Recall that we used Feynman-Kac on V , implicitly, $V \in C^1$.

Therefore, at the free-boundary b ,

$$\frac{\partial}{\partial x} \left(\frac{b-c}{b^{\delta+}} x^{\delta+} \right) \Big|_{x=b} = 1.$$

derivative from the left

from the right: $x=c$.

"Principle of smooth fit".

$$\Rightarrow b = \frac{c \delta+}{\delta+ - 1}.$$

exercise.

Sanity check: $\delta+ - 1 > 0 \iff \beta > \mu \quad \checkmark$.

Remark i) We wait on $C := (0, b)$. One can verify $V > g$, i.e. waiting gives you higher value. ↳ exercise

We stop on $D := [b, \infty)$, clearly $V = g$, waiting doesn't do you any better.

We define:

Continuation set $C = \{x \in \mathbb{R} : V(x) > g(x)\}$

Stopping set $D = \{x \in \mathbb{R} : V(x) = g(x)\}$

ii). Typically, we consider $\tau^* := \inf \{t \geq 0 : X_t \in D\}$.

Is our guess correct? We usually need to construct and prove a verification thm for the problem we study.

Theorem 11.2. (Verification for OS).

Consider the problem $V(x) = \sup_{\tau} \mathbb{E}[g(X_{\tau})]$. Assume $\hat{V} \in C^2$, and \hat{V} solves the free boundary problem.

$$\begin{cases} \Delta \hat{V} \leq 0. \\ \hat{V} \geq G \end{cases} \quad (*)$$

for all x , and there exists C and s_0 .

$$\hat{V} > g \text{ in } C, \quad \hat{V} = g \text{ in } D.$$

then $\hat{V} \equiv V$. and $\tau^* := \inf \{t \geq 0 : X_t \in D\}$ is an optimal strategy.

Remark. i) Moreover, assume g smooth in a neighbourhood of ∂C and ∂C is "nice", then $(*)$ splits into:

$$\begin{cases} \Delta \hat{V} = 0, \text{ in } C. \\ \hat{V} = g, \text{ in } D \\ \hat{V}|_{\partial C} = g|_{\partial C}. \quad (\text{"Smooth fit"}). \end{cases}$$

ii) In our cases, usually it's possible to solve it explicitly by "guess and verify".

Prf. (Sketch) Let $\theta = \tau \wedge T$, Apply Ito on \hat{V} :

$$\hat{V}(X_{\theta}) = \hat{V}(x) + \int_0^{\theta} \Delta \hat{V}(X_s) ds + \int_0^{\theta} \dots dB_s.$$

Since $\Delta \hat{V} \leq 0$,

$$\mathbb{E}[\hat{V}(X_0) - \hat{V}(x)] \leq 0 + \underbrace{\mathbb{E}\left[\int_0^0 \dots dB_s\right]}_0 = 0.$$

$$\Rightarrow \hat{V}(x) \geq \mathbb{E}[\hat{V}(X_0)] \geq \mathbb{E}[\hat{g}(X_0)]$$

↑
 $\hat{V} \geq g.$

If $\tau < \infty$ a.s., take T up to ∞ : $\lim_{T \rightarrow \infty} \tau \wedge T = \tau.$

$$\Rightarrow \hat{V}(x) \geq \mathbb{E}[g(X_\tau)] \text{ for all } \tau < \infty \text{ a.s.}$$

$$\Rightarrow \hat{V}(x) \geq \sup_{\tau} \mathbb{E}[g(X_\tau)].$$

The other direction: take $\tau = \tau^*$.

$$\mathbb{E}[\hat{V}(X_{\tau^*})] = \hat{V}(x) + \mathbb{E}\left[\int_0^{\tau^*} \underbrace{(\mathcal{L}\hat{V})(X_s)}_{= 0 \text{ in } D} ds\right] = \hat{V}(x).$$

$$\hat{V}(x) = \mathbb{E}[\hat{V}(X_{\tau^*})] = \mathbb{E}[g(X_{\tau^*})] \leq \sup_{\tau} \mathbb{E}[g(X_\tau)].$$

↑
 $\hat{V} = g$ in D

It follows that τ^* is an optimal strategy. □

Remark. Other formulations of OS:

i). With discounting.

$$V(x) = \sup_{\tau} \mathbb{E}_x \left[e^{-\lambda \tau} g(X_\tau) + \int_0^{\tau} e^{-\lambda s} h(X_s) ds \right]$$

where the discounting process $\lambda_t \in \mathbb{R}_+$ is

$$\lambda_t = \int_0^t \pi(x_s) ds.$$

then V solves.

$$\begin{cases} \Delta V(x) + h(x) - \pi(x) V(x) = 0 & \text{in } C \\ V = g & \text{in } D \\ \text{Smooth fit} \end{cases}$$

(check with 11.1).

ii). The "inf" equivalence.

Sometimes we want to minimize the cost, the cost function

$$V(x) = \inf_{\tau} E[g(x_{\tau})].$$

the method of "guess and verify" is entirely parallel

Applications $\begin{matrix} \xrightarrow{\hspace{1cm}} \\ \xleftarrow{\hspace{1cm}} \end{matrix}$ optimal stopping $\begin{matrix} \xrightarrow{\hspace{1cm}} \\ \xleftarrow{\hspace{1cm}} \end{matrix}$ Free-bdr problems