

Optimal Stopping under Model Ambiguity

— A Time-Consistent Equilibrium Approach

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Without ambiguity:

- ▶ choose $\tau \in \mathcal{T}$ to maximize

$$\mathbb{E}^{\mathbb{P}}[e^{-r\tau}g(X_\tau)]. \quad (1)$$

With ambiguity:

- ▶ \mathcal{P} : the set of *plausible* probabilities \mathbb{P} , i.e. *priors*.
- ▶ Worst-case analysis: choose $\tau \in \mathcal{T}$ to maximize

$$\inf_{\mathbb{P} \in \mathcal{P}} \mathbb{E}^{\mathbb{P}}[e^{-r\tau}g(X_\tau)]. \quad (2)$$

- ▶ Best-case analysis: choose $\tau \in \mathcal{T}$ to maximize

$$\sup_{\mathbb{P} \in \mathcal{P}} \mathbb{E}^{\mathbb{P}}[e^{-r\tau}g(X_\tau)]. \quad (3)$$

▶ **Worst-case (or best-case) analysis:**

The dominant approach...

Riedel (2009), Bayraktar & Yao (2011a, 2011b, 2014, 2017), Cheng and Riedel (2013), Ekren et al. (2014), Nutz & Zhang (2015), ...

▶ **What is missing?**

An agent's **ambiguity attitude**.

- ▶ Curley & Yates (1989), Heath & Tversky (1991):
With the same \mathcal{P} , different agents have different levels of ambiguity aversion.

THE α -MAXMIN OBJECTIVE

Motivated by the α -maxmin preference in Ghirardato et al. (2004), we propose to maximize

$$\alpha \inf_{\mathbb{P} \in \mathcal{P}} \mathbb{E}^{\mathbb{P}} [e^{-r\tau} g(X_\tau)] + (1 - \alpha) \sup_{\mathbb{P} \in \mathcal{P}} \mathbb{E}^{\mathbb{P}} [e^{-r\tau} g(X_\tau)].$$

- ▶ *Ambiguity* is captured by \mathcal{P} .
- ▶ *Ambiguity attitude* is captured by $\alpha \in [0, 1]$.
 - ▶ $\alpha = 1$: worst-case analysis (**purely ambiguity-averse**)
 - ▶ $\alpha = 0$: best-case analysis (**purely ambiguity-loving**)

Our goal: optimal stopping under any $\alpha \in [0, 1]$.

TIME INCONSISTENCY

The problem

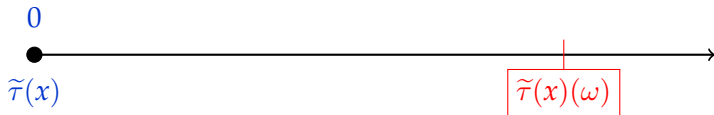
$$\sup_{\tau \in \mathcal{T}} \left(\alpha \inf_{\mathbb{P} \in \mathcal{P}} \mathbb{E}^{\mathbb{P}} [e^{-r\tau} g(X_{\tau})] + (1 - \alpha) \sup_{\mathbb{P} \in \mathcal{P}} \mathbb{E}^{\mathbb{P}} [e^{-r\tau} g(X_{\tau})] \right) \quad (4)$$

is **time-inconsistent!**

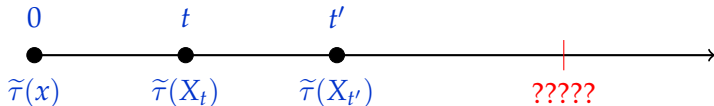
- ▶ An α -maxmin preference induces time inconsistency. (Schröder (2011) and Beissner et al. (2016))

Not meaningful to find an optimal stopping time here.
(unless one dictates his future selves' behavior).

► **Problem Solved. *Feeling Good?***



► **The Reality:**



► **Time Inconsistency:**

- $\tilde{\tau}(x), \tilde{\tau}(X_t), \tilde{\tau}(X_{t'})$ may all be different.

TOWER PROPERTY

- ▶ Time consistency of (1) = tower property of conditional expectations.
- ▶ For (2) and (3),
 - ▶ **Epstein & Schneider (2003)**: tower property holds if
the set of priors is *rectangular*
(**stable under pasting conditional probabilities**).
 \implies Time consistency follows.
- ▶ Nutz and van Handel (2013), Bayraktar and Yao (2014), Ekren et al. (2014), Nutz and Zhang (2015)...
- ▶ Under the α -maxmin preference,
 - ▶ **Schröder (2011), Beissner et al. (2016)**:
Tower property fails, even under “**stable under pasting**”.
 - ▶ Time inconsistency is a genuine challenge for (4).

How to resolve time inconsistency?

Consistent Planning [Strotz (1955-56)]

- ▶ Take into account future selves' behavior.

Find an *equilibrium* strategy that

once being enforced over time,
no future self would want to deviate from.

- ▶ How to precisely define and find equilibrium strategies?

ITERATIVE APPROACH

Huang & Nguyen-Huu (2018):

Iterative approach for time-inconsistent stopping problems

- ▶ **Equilibrium strategies = fixed points of an operator**
 - ▶ find equilibria easily via fixed-point iterations.
- ▶ Applications:
 - ▶ non-exponential discounting;
 - ▶ probability distortion.

Huang & Nguyen-Huu (2018), Huang & Zhou (2017, 2019),
Huang, Nguyen-Huu, and Zhou (2019)

This talk:

- ▶ Model ambiguity + *ambiguity attitude*
- ▶ A time-inconsistent stopping problem under the α -maxmin preference
- ▶ Iterative approach

THE MODEL

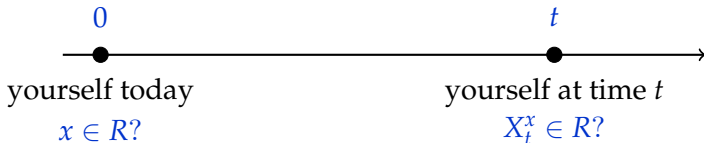
- ▶ $\Omega := C([0, \infty); \mathbb{R}^d)$.
 - ▶ $\Omega^x := \{\omega \in \Omega : \omega_0 = x\}, \forall x \in \mathbb{R}^d$.
- ▶ B : canonical process.
- ▶ $\mathfrak{P}(\Omega)$: the set of probability measures on Ω .
- ▶ For any $x \in \mathbb{R}^d$, let
$$\mathcal{P}(x) \subseteq \{\mathbb{P} \in \mathfrak{P}(\Omega) : \mathbb{P}(\Omega^x) = 1, B \text{ is strong Markov under } \mathbb{P}\}$$
denote the *set of priors* of an agent at $x \in \mathbb{R}^d$.
- ▶ $\mathcal{U}(\mathbb{R}^d)$: the set of *universally measurable* subsets of \mathbb{R}^d .

GAME-THEORETIC APPROACH

- ▶ Focus on *hitting times* to regions in \mathbb{R}^d , i.e.,

$$\tau_R := \inf\{t \geq 0 : B_t \in R\}, \quad R \in \mathcal{U}(\mathbb{R}^d).$$

- ▶ For convenience, we call $R \in \mathcal{U}(\mathbb{R}^d)$ a *stopping policy*.
- ▶ Given a stopping policy $R \in \mathcal{U}(\mathbb{R}^d)$,



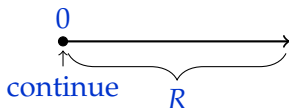
- ▶ Game-theoretic thinking at time 0:
Given that every **future self** will follow R ,
 - ▶ **What is the best stopping strategy at time 0?**

BEST STOPPING STRATEGY

The agent at $x \in \mathbb{R}^d$ can either stop or continue.

- ▶ If she stops, gets $g(x)$ right away.
- ▶ If she continues, she will eventually stop at the moment

$$\rho_R := \inf \{t > 0 : B_t \in R\}.$$



\implies Her α -maxmin expected payoff is then

$$J(x, R) := \boxed{\alpha \inf_{\mathbb{P} \in \mathcal{P}(x)} \mathbb{E}^{\mathbb{P}} [e^{-r\rho_R} g(B_{\rho_R})] + (1 - \alpha) \sup_{\mathbb{P} \in \mathcal{P}(x)} \mathbb{E}^{\mathbb{P}} [e^{-r\rho_R} g(B_{\rho_R})]}.$$

- ▶ The Best stopping policy at time 0 is

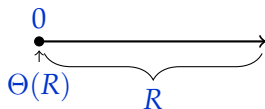
$$\Theta(R) := S_R \cup (I_R \cap R),$$

where

$$S_R := \{x : g(x) > J(x, R)\},$$

$$I_R := \{x : g(x) = J(x, R)\},$$

$$C_R := \{x : g(x) < J(x, R)\}.$$



- ▶ In general, $\Theta(R) \neq R$.
 - ▶ Player 0 wants to follow $\Theta(R)$, instead of R .

EQUILIBRIUM

Definition

$R \in \mathcal{U}(\mathbb{R}^d)$ is called an **equilibrium** if $\Theta(R) = R$.

- ▶ **Trivial Equilibrium:** Consider $R := \mathbb{R}^d$. Then $I_R = \mathbb{R}^d$, so $\Theta(R) = S_R \cup (I_R \cap R) = R$.
- ▶ **In general,** given any $R \in \mathcal{B}(\mathbb{R}^d)$, carry out iteration:
$$R \longrightarrow \Theta(R) \longrightarrow \Theta^2(R) \longrightarrow \cdots \longrightarrow \text{“equilibrium”??}$$
- ▶ **To show:**
(i) $R_* := \lim_{n \rightarrow \infty} \Theta^n(R)$ converges (ii) $\Theta(R_*) = R_*$.

STRONG FORMULATION

- ▶ $d = 1$ and \mathbb{P}_0 : the Wiener measure.
- ▶ $I = (\ell, r)$, for some given $-\infty \leq \ell < r \leq \infty$. Consider

$$X_t^{x,b,\sigma} = x + \int_0^t b(X_s^{x,b,\sigma}) ds + \int_0^t \sigma(X_s^{x,b,\sigma}) dB_s, \quad \mathbb{P}_0\text{-a.s.} \quad (5)$$

- ▶ \mathcal{L} : the set of all $b, \sigma : I \rightarrow \mathbb{R}$ that are
 - (i) Lipschitz, grows linearly;
 - (ii) $\sigma^2 > 0$ on I .
- ▶ \mathcal{A} : the set of all *set-valued* maps $\Pi : I \rightarrow 2^{\mathcal{L}}$.
- ▶ \mathcal{A}^∞ : the set of all *set-valued* maps $\Pi : I \rightarrow 2^{\mathcal{L}}$ satisfying:
for any $x \in I$, $\exists K > 0$ such that for any $(b, \sigma) \in \Pi(x)$,

$$\begin{aligned} |b(u) - b(v)| + |\sigma(u) - \sigma(v)| &\leq K|u - v| \\ |b(u)| + |\sigma(u)| &\leq K(1 + |u|), \quad \forall u, v \in I. \end{aligned}$$

STRONG FORMULATION

- ▶ For each $x \in I$ and $(b, \sigma) \in \mathfrak{L}$, define

$$\mathbb{P}_{b,\sigma}^x := \mathbb{P}_0 \circ (X^{x,b,\sigma})^{-1} \in \mathfrak{P}(\Omega). \quad (6)$$

- ▶ Given $\Pi \in \mathcal{A}$, we introduce

$$\mathcal{P}(x) := \{\mathbb{P}_{b,\sigma}^x : (b, \sigma) \in \Pi(x)\}, \quad \forall x \in I. \quad (7)$$

Lemma

Given $x \in I$ and $(b, \sigma) \in \mathfrak{L}$, $X^{x,b,\sigma}$ is a regular diffusion, i.e.,

$$\text{for any } x \in I, \quad \mathbb{P}_0(T_y^x < \infty) > 0, \quad \forall y \in I.$$

This implies

$$\begin{aligned} T_{(\ell,x)}^x &= T_{(x,r)}^x = 0 \quad \mathbb{P}_0\text{-a.s.} \\ \implies \rho_{(\ell,x)} &= \rho_{(x,r)} = 0 \quad \mathbb{P}_{b,\sigma}^x\text{-a.s.} \end{aligned}$$

Proposition

For any $R \in \mathcal{U}(I)$, $\boxed{R \subseteq \Theta(R)}$. Then,

$$R_* := \lim_{n \rightarrow \infty} \Theta^n(R) = \bigcup_{n \in \mathbb{N}} \Theta^n(R).$$

CONVERGENCE IN CAPACITY

Standard SDE estimate + proof of Kolmogorov's criterion:

Lemma

For any $\Pi \in \mathcal{A}^\infty$, $\mathcal{P}(x)$ is relatively compact for all $x \in I$.

Introduce $R_n := \Theta^n(R)$, $\rho^n := \rho_{R_n}$, $\rho^* := \rho_{R^*}$.

Lemma

For any $\Pi \in \mathcal{A}^\infty$ and $\varepsilon > 0$,

$$\lim_{n \rightarrow \infty} \sup_{\mathbb{P} \in \mathcal{P}(x)} \mathbb{P} (|\rho^n - \rho^*| \geq \varepsilon) = 0, \quad (8)$$

$$\lim_{n \rightarrow \infty} \sup_{\mathbb{P} \in \mathcal{P}(x)} \mathbb{P} (|B_{\rho^n} - B_{\rho^*}| \mathbf{1}_{\{\rho^n < \infty\}} \geq \varepsilon) = 0. \quad (9)$$

Relies crucially on relative compactness of $\mathcal{P}(x)$.

THE MAIN RESULT

Theorem

Fix $\Pi \in \mathcal{A}^\infty$. Let $g : \bar{I} \rightarrow \mathbb{R}$ be continuous and

$$\lim_{t \rightarrow \infty} e^{-rt} g(X_t^{x,b,\sigma}) = 0 \quad \mathbb{P}_0\text{-a.s.}, \quad \forall x \in I, (b, \sigma) \in \Pi(x).$$

Then, for any $R \in \mathcal{U}(I)$, R_* is an equilibrium, i.e.

$$\Theta(R_*) = R_*.$$

Consequently,

$$\mathcal{E} = \left\{ \lim_{n \rightarrow \infty} \Theta^n(R) : R \in \mathcal{U}(I) \right\}.$$

REAL OPTIONS VALUATION

- ▶ Applies financial option pricing techniques to corporate investment decision making.
 - ▶ Use **risk-neutral pricing** to evaluate the right, but not the obligation, to undertake a business plan.
- ▶ Suffers *model ambiguity* more severely than pricing a financial option...
 - ▶ ... as the underlying can be neither tradable nor fully observable.
- ▶ This leads to
 - ▶ a set of *plausible* risk-neutral measures
⇒ an interval of *plausible* values of a real option.
 - ▶ How to deal with these multiple values?
Unclear in the literature....

EXAMPLE

- ▶ $g(x) = (K - x)^+$ for a given $K > 0$.
- ▶ The underlying is a GBM:

$$X_t^{x,b,\sigma} = x + \int_0^t b X_s^{x,b,\sigma} ds + \int_0^t \sigma X_s^{x,b,\sigma} dB_s, \quad \mathbb{P}_0\text{-a.s.},$$

for some *unknown* $b \in \mathbb{R}$ and $\sigma > 0$.

- ▶ Riskfree rate $r > 0$ is known.
- ▶ $\sigma \in [\underline{\sigma}, \bar{\sigma}]$, for given $0 < \underline{\sigma} < \bar{\sigma}$ (**uncertain volatility model**)

⇒ The α -maxmin objective:

$$J(x, R) = \alpha \inf_{\sigma \in [\underline{\sigma}, \bar{\sigma}]} \mathbb{E}^{\mathbb{P}_0} \left[e^{-rT_R} (K - X_{T_R}^{x,r,\sigma})^+ \right] \\ + (1 - \alpha) \sup_{\sigma \in [\underline{\sigma}, \bar{\sigma}]} \mathbb{E}^{\mathbb{P}_0} \left[e^{-rT_R} (K - X_{T_R}^{x,r,\sigma})^+ \right].$$

Define

$$m_1 := \frac{2r}{\underline{\sigma}^2}, \quad m_2 := \frac{2r}{\bar{\sigma}^2}.$$

$$a^* := \frac{m_1 \alpha + m_2 (1 - \alpha)}{1 + m_1 \alpha + m_2 (1 - \alpha)} K \in (0, K).$$

Proposition

- ▶ $\mathcal{E} = \{(0, a] : a^* \leq a \leq K\}$.
- ▶ $(0, a^*]$ is *optimal* among \mathcal{E} .

“Optimal” in the sense that for any $a^* \leq a \leq K$,

$$J(x, (0, a^*]) \geq J(x, (0, a]) \quad \forall x \in (0, \infty).$$

- ▶ *Optimal equilibrium* in Huang & Zhou (2019, 2017).

OBSERVATIONS

- ▶ a^* is increasing in $\alpha \in [0, 1]$.
 - ▶ The *larger* α , the *larger* the optimal equilibrium $(0, a^*)$.
 - ▶ The *more risk-averse*,
the *more eager to stop*—to exit the uncertain environment.
- ▶ When $\underline{\sigma} = \bar{\sigma} = \sigma$ (**no ambiguity**),

$$a^* = \frac{2r/\sigma^2}{1 + 2r/\sigma^2}K.$$

This is exactly the optimal stopping threshold for

$$\sup_{\tau \in \mathcal{T}} \mathbb{E}^{\mathbb{P}_0} [e^{-r\tau} (K - X_{\tau}^{x,r,\sigma})^+]$$

(Theorem 2.7.2 in Karatzas and Shreve (1998)).

SUMMARY

- ▶ Resolves the time-inconsistent stopping problem under the α -maxmin preference
 - ▶ Allow us to go beyond worst-case (best-case) analysis.
- ▶ Focuses on **ambiguity aversion**, a *cause of time inconsistency* only slightly discussed in the literature.
 - ▶ relative to *non-exponential discounting, probability distortion,...*
- ▶ Provides a new approach for real options valuation
 - ▶ α -maxmin preference + equilibrium approach
- ▶ A new measurable projection theorem.
 - ▶ *does not* require specific Borel structure.

THANK YOU!!

- ▶ *“Optimal Stopping under Model Ambiguity: a Time-Consistent Equilibrium Approach”*
(H. and X. Yu), Available @ arXiv:1906.01232.