


Ergodic singular control for ambiguous compound-Poisson **jump-diffusion** processes

 Abel Guada Azze

 abel.guada@cunef.edu

 CUNEF Universidad, Madrid

 XVII CLAPEM

 Universidad de la República, Montevideo, Uruguay

 6 March, 2025

MOTIVATION

MOTIVATION: CASH MANAGEMENT

- **Context:** A firm **operates continuously** over an **indefinite horizon** and must maintain a cash balance X_t to meet day-to-day obligations.

MOTIVATION: CASH MANAGEMENT

- **Context:** A firm **operates continuously** over an **indefinite horizon** and must maintain a cash balance X_t to meet day-to-day obligations.
- **Cash evolves through:**
 - ▶ High-frequency net inflows/outflows (e.g., many small receipts/payments).
⇒ Diffusion approximation
 - ▶ Rare significant liquidity **shocks** (e.g., tax and supplier payments).
⇒ Downward compound-Poisson jumps

MOTIVATION: CASH MANAGEMENT

- **Context:** A firm **operates continuously** over an **indefinite horizon** and must maintain a cash balance X_t to meet day-to-day obligations.
- **Cash evolves through:**
 - ▶ High-frequency net inflows/outflows (e.g., many small receipts/payments).
⇒ **Diffusion approximation**
 - ▶ Rare significant liquidity **shocks** (e.g., tax and supplier payments).
⇒ **Downward compound-Poisson jumps**
- **Cash can be intervened:**
 - ▶ The firm can add (U_t) and/or remove (D_t) cash **instantaneously at any time**.
⇒ **Singular two-sided controls**
 - ▶ Transfers carry instantaneous **proportional transaction costs**.
⇒ **Until horizon T , cumulative transaction costs amount to $c_U U_T + c_D D_T$**

MOTIVATION: CASH MANAGEMENT

- **Context:** A firm **operates continuously** over an **indefinite horizon** and must maintain a cash balance X_t to meet day-to-day obligations.
- **Cash evolves through:**
 - ▶ High-frequency net inflows/outflows (e.g., many small receipts/payments).
⇒ **Diffusion approximation**
 - ▶ Rare significant liquidity **shocks** (e.g., tax and supplier payments).
⇒ **Downward compound-Poisson jumps**
- **Cash can be intervened:**
 - ▶ The firm can add (U_t) and/or remove (D_t) cash **instantaneously at any time**.
⇒ **Singular two-sided controls**
 - ▶ Transfers carry instantaneous **proportional transaction costs**.
⇒ **Until horizon T , cumulative transaction costs amount to $c_U U_T + c_D D_T$**
- **Holding cash is not free:**
 - ▶ Excess cash entails an **opportunity cost**, and low cash entails a **liquidity cost**.
⇒ **Until horizon T , cumulative cash-level costs amount to $\int_0^T c(X_t) dt$**

MOTIVATION: CASH MANAGEMENT

- **Context:** A firm **operates continuously** over an **indefinite horizon** and must maintain a cash balance X_t to meet day-to-day obligations.
- **Cash evolves through:**
 - ▶ High-frequency net inflows/outflows (e.g., many small receipts/payments).
⇒ **Diffusion approximation**
 - ▶ Rare significant liquidity **shocks** (e.g., tax and supplier payments).
⇒ **Downward compound-Poisson jumps**
- **Cash can be intervened:**
 - ▶ The firm can add (U_t) and/or remove (D_t) cash **instantaneously at any time**.
⇒ **Singular two-sided controls**
 - ▶ Transfers carry instantaneous **proportional transaction costs**.
⇒ **Until horizon T , cumulative transaction costs amount to $c_U U_T + c_D D_T$**
- **Holding cash is not free:**
 - ▶ Excess cash entails an **opportunity cost**, and low cash entails a **liquidity cost**.
⇒ **Until horizon T , cumulative cash-level costs amount to $\int_0^T c(X_t) dt$**
- **Goal:** Minimize **average long-run** (ergodic) cost.
⇒ **Ergodic singular control!**

VISUALIZING THE CASH MANAGEMENT PROBLEM

THE ERGODIC SINGULAR PROBLEM

Ergodic singular control problem

$$\gamma = \inf_{(U,D)} \lim_{T \rightarrow \infty} \frac{1}{T} \mathbb{E}_x^0 \left[\int_0^T c(X_t^{U,D}) dt + c_U U_T + c_D D_T \right]$$

THE ERGODIC SINGULAR PROBLEM

Ergodic singular control problem

$$\gamma = \inf_{(U,D)} \lim_{T \rightarrow \infty} \frac{1}{T} \mathbb{E}_x^0 \left[\int_0^T c(X_t^{U,D}) dt + c_U U_T + c_D D_T \right]$$

- (U, D) : admissible pair of **singular controls** pushing up/down.
- (c_U, c_D) : per-unit cost of pushing up/down.

THE ERGODIC SINGULAR PROBLEM

Ergodic singular control problem

$$\gamma = \inf_{(U,D)} \lim_{T \rightarrow \infty} \frac{1}{T} \mathbb{E}_x^0 \left[\int_0^T c(X_t^{U,D}) dt + c_U U_T + c_D D_T \right]$$

- (U, D) : admissible pair of **singular controls** pushing up/down.
- (c_U, c_D) : per-unit cost of pushing up/down.
- $X_t^{U,D}$: controlled process

$$dX_t^{U,D} = b(X_{t-}^{U,D}) dt + \sigma dB_t + d\tilde{Z}_t + dU_t - dD_t,$$

for:

- ▶ $b(\cdot)$: Lipschitz-continuous drift function;
- ▶ σ : volatility;
- ▶ B_t : \mathbb{P}^0 -standard Brownian motion;
- ▶ \tilde{Z}_t : \mathbb{P}^0 -compensated compound-Poisson process (independent from B_t) with intensity $r > 0$ and i.i.d. jumps $\{Y_n\}_{n=1}^\infty$ with jump-size distribution F .

THE ERGODIC SINGULAR PROBLEM

Ergodic singular control problem

$$\gamma = \inf_{(U,D)} \lim_{T \rightarrow \infty} \frac{1}{T} \mathbb{E}_x^0 \left[\int_0^T c(X_t^{U,D}) dt + c_U U_T + c_D D_T \right]$$

- (U, D) : admissible pair of **singular controls** pushing up/down.
- (c_U, c_D) : per-unit cost of pushing up/down.
- $X_t^{U,D}$: controlled process

$$dX_t^{U,D} = b(X_{t-}^{U,D}) dt + \sigma dB_t + d\tilde{Z}_t + dU_t - dD_t,$$

for:

- ▶ $b(\cdot)$: Lipschitz-continuous drift function;
 - ▶ σ : volatility;
 - ▶ B_t : \mathbb{P}^0 -standard Brownian motion;
 - ▶ \tilde{Z}_t : \mathbb{P}^0 -compensated compound-Poisson process (independent from B_t) with intensity $r > 0$ and i.i.d. jumps $\{Y_n\}_{n=1}^{\infty}$ with jump-size distribution F .
- $c(\cdot)$: measurable such that $\mathbb{E}_x^0 \left[\int_0^t |c(X_s^{U,D})| ds \right] < \infty$ for all $x \in \mathbb{R}$.

THE ERGODIC SINGULAR PROBLEM

Ergodic singular control problem

$$\gamma = \inf_{(U,D)} \lim_{T \rightarrow \infty} \frac{1}{T} \mathbb{E}_x^0 \left[\int_0^T c(X_t^{U,D}) dt + c_U U_T + c_D D_T \right]$$

- (U, D) : admissible pair of **singular controls** pushing up/down.
- (c_U, c_D) : per-unit cost of pushing up/down.
- $X_t^{U,D}$: controlled process

$$dX_t^{U,D} = b(X_{t-}^{U,D}) dt + \sigma dB_t + d\tilde{Z}_t + dU_t - dD_t,$$

for:

- ▶ $b(\cdot)$: Lipschitz-continuous drift function;
- ▶ σ : volatility;
- ▶ B_t : \mathbb{P}^0 -standard Brownian motion;
- ▶ \tilde{Z}_t : \mathbb{P}^0 -compensated compound-Poisson process (independent from B_t) with intensity $r > 0$ and i.i.d. jumps $\{Y_n\}_{n=1}^\infty$ with jump-size distribution F .
- $c(\cdot)$: measurable such that $\mathbb{E}_x^0 \left[\int_0^t |c(X_s^{U,D})| ds \right] < \infty$ for all $x \in \mathbb{R}$.
- γ : **ergodic value**.

THE ERGODIC SINGULAR PROBLEM

Ergodic singular control problem

$$\gamma = \inf_{(U,D)} \lim_{T \rightarrow \infty} \frac{1}{T} \mathbb{E}_x^0 \left[\int_0^T c(X_t^{U,D}) dt + c_U U_T + c_D D_T \right]$$

- (U, D) : admissible pair of **singular controls** pushing up/down.
- (c_U, c_D) : per-unit cost of pushing up/down.
- $X_t^{U,D}$: controlled process

$$dX_t^{U,D} = b(X_{t-}^{U,D}) dt + \sigma dB_t + d\tilde{Z}_t + dU_t - dD_t,$$

for:

- ▶ $b(\cdot)$: Lipschitz-continuous drift function;
- ▶ σ : volatility;
- ▶ B_t : \mathbb{P}^0 -standard Brownian motion;
- ▶ \tilde{Z}_t : \mathbb{P}^0 -compensated compound-Poisson process (independent from B_t) with intensity $r > 0$ and i.i.d. jumps $\{Y_n\}_{n=1}^{\infty}$ with jump-size distribution F .
- $c(\cdot)$: measurable such that $\mathbb{E}_x^0 \left[\int_0^t |c(X_s^{U,D})| ds \right] < \infty$ for all $x \in \mathbb{R}$.
- γ : **ergodic value**.
- \mathbb{E}_x^0 : expectation w.r.t. the **benchmark measure** \mathbb{P}_x^0 , indicating $\mathbb{P}_x^0(X_0 = x) = 1$.

MODEL AMBIGUITY

We do not trust the probability measure!

To capture misspecification (in the **drift** and **intensity**) we consider the family $\{\mathbb{P}^{\kappa, \lambda}\}_{(\kappa, \lambda) \in \Lambda}$ whose RN densities w.r.t. \mathbb{P}^0 are given by

$$\frac{\mathbb{P}^{\kappa, \lambda}}{\mathbb{P}^0} \Big|_{\mathcal{F}_t} = \exp \left\{ \int_0^t \kappa_s dB_s - \frac{1}{2} \int_0^t \kappa_s^2 ds \right\} \exp \left\{ \int_0^t \ln \left(\frac{\lambda_s}{r} \right) dN_s - \int_0^t (\lambda_s - r) ds \right\}.$$

¹ Knight (1921). Risk, Uncertainty and Profit. Houghton Mifflin Company

We do not trust the probability measure!

To capture misspecification (in the drift and intensity) we consider the family $\{\mathbb{P}^{\kappa, \lambda}\}_{(\kappa, \lambda) \in \Lambda}$ whose RN densities w.r.t. \mathbb{P}^0 are given by

$$\frac{\mathbb{P}^{\kappa, \lambda}}{\mathbb{P}^0} \Big|_{\mathcal{F}_t} = \exp \left\{ \int_0^t \kappa_s dB_s - \frac{1}{2} \int_0^t \kappa_s^2 ds \right\} \exp \left\{ \int_0^t \ln \left(\frac{\lambda_s}{r} \right) dN_s - \int_0^t (\lambda_s - r) ds \right\}.$$

- $\kappa_t = \kappa(X_t)$ and $\lambda_t = \lambda(X_t)$: state-dependent drift and intensity distortions.
- $\kappa(\cdot)$ and $\lambda(\cdot)$: drift and intensity distortion functions
- Λ : set of admissible distortions:

$$|\kappa(x)| \leq \delta, \quad (1 - \varepsilon)r \leq \lambda(x) \leq r(1 + \varepsilon),$$

for $\delta > 0$ and $0 \leq \varepsilon < 1$.

We do not trust the probability measure!

To capture misspecification (in the **drift** and **intensity**) we consider the family $\{\mathbb{P}^{\kappa, \lambda}\}_{(\kappa, \lambda) \in \Lambda}$ whose RN densities w.r.t. \mathbb{P}^0 are given by

$$\frac{\mathbb{P}^{\kappa, \lambda}}{\mathbb{P}^0} \Big|_{\mathcal{F}_t} = \exp \left\{ \int_0^t \kappa_s dB_s - \frac{1}{2} \int_0^t \kappa_s^2 ds \right\} \exp \left\{ \int_0^t \ln \left(\frac{\lambda_s}{r} \right) dN_s - \int_0^t (\lambda_s - r) ds \right\}.$$

- $\kappa_t = \kappa(X_t)$ and $\lambda_t = \lambda(X_t)$: state-dependent drift and intensity distortions.
- $\kappa(\cdot)$ and $\lambda(\cdot)$: drift and intensity distortion functions
- Λ : set of admissible distortions:

$$|\kappa(x)| \leq \delta, \quad (1 - \varepsilon)r \leq \lambda(x) \leq r(1 + \varepsilon),$$

for $\delta > 0$ and $0 \leq \varepsilon < 1$.

- Under $\mathbb{P}^{\kappa, \lambda}$:

$$dX_t^{U,D} = (b(X_{t-}^{U,D}) + \sigma \kappa(X_{t-}^{U,D}) + (\lambda(X_{t-}^{U,D}) - r) \mathbb{E}[Y]) dt + \sigma dB_t^\kappa + d\tilde{Z}_t^\lambda + dU_t - dD_t,$$

for:

- ▶ B_t^κ : $\mathbb{P}^{\kappa, \lambda}$ -standard Brownian motion;
- ▶ \tilde{Z}_t^λ : $\mathbb{P}^{\kappa, \lambda}$ -compensated “compound-Poisson process” (independent from B_t^κ) with space-dependent intensity $\lambda(X_{t-}^{U,D})$ and i.i.d. jumps $\{Y_n\}_{n=1}^\infty$ with distr. F .

Robust Ergodic singular control problem

$$\gamma = \inf_{(U,D) \in \mathcal{A}} \sup_{(\kappa,\lambda) \in \Lambda} J_x(U, D, \kappa, \lambda)$$

- **Robustness:** the controller is ambiguity averse: worst-case scenario.
- **Ergodic cost:** $J_x(U, D, \kappa, \lambda) := \lim_{T \rightarrow \infty} \frac{1}{T} \mathbb{E}_x^{\kappa, \lambda} \left[\int_0^T c(X_t^{U,D}) dt + c_U U_T + c_D D_T \right]$.
- **Solving the problem means finding:**
 - ▶ (U^*, D^*) : optimal controls;
 - ▶ (κ^*, λ^*) optimal ambiguity functions,

such that there is a saddle point at $(U^*, D^*, \kappa^*, \lambda^*)$:

$$\gamma = \inf_{(U,D) \in \mathcal{A}} \sup_{(\kappa,\lambda) \in \Lambda} J_x(U, D, \kappa, \lambda) = \sup_{(\kappa,\lambda) \in \Lambda} \inf_{(U,D) \in \mathcal{A}} J_x(U, D, \kappa, \lambda) = J_x(U^*, D^*, \kappa^*, \lambda^*),$$

STATE OF THE ART: ERGODIC SINGULAR CONTROL

- **Diffusions (no ambiguity)**

Bast literature; early formulations for Brownian dynamics¹, broad settings^{2,3,4}, verification, and vanishing-discount approaches^{5,6,7}.

- **Diffusions with jumps (no ambiguity)**

Scarcer literature and largely intractable; successful attempt⁸ using a Dynkin-game representation and vanishing discount, other works^{9,10,11,12} treating discounted case.

- **Model ambiguity (diffusions)**

Very limited literature; zero-sum game and (max-min) HJB characterization¹³; mean-field (many controllers) extension¹⁴.

- **Model ambiguity with jumps**

We are not aware of results in this category.

¹ Karatzas (1983). A class of singular stochastic control problems. *Advances in Applied Probability*

² Menaldi et al. (1984). Some singular control problem with long term average criterion. In: Springer

³ Weerasinghe (2002). Stationary stochastic control for Itô processes. *Advances in Applied Probability*

⁴ Kunwai et al. (2022). On an Ergodic Two-Sided Singular Control Problem. *Applied Mathematics & Optimization*

⁵ Weerasinghe (2007). An Abelian Limit Approach to a Singular Ergodic Control Problem. *SIAM Journal on Control and Optimization*

⁶ Arapostathis et al. (2011). Ergodic Control of Diffusion Processes. Cambridge University Press

⁷ Menaldi et al. (2013). Singular ergodic control for multidimensional Gaussian–Poisson processes. *Stochastics*

⁸ Mordecki et al. (July 2025). Two sided long-time optimization singular control problems for Lévy processes and Dynkin's games.

⁹ Yamazaki (2017). Inventory Control for Spectrally Positive Lévy Demand Processes. *Mathematics of Operations Research*

¹⁰ Hernández-Hernández et al. (2016). Optimality of Refraction Strategies for Spectrally Negative Lévy Processes. *SIAM Journal on Control and Optimization*

¹¹ Pérez et al. (2020). Optimal Periodic Replenishment Policies for Spectrally Positive Lévy Demand Processes. *SIAM Journal on Control and Optimization*

¹² Noba et al. (2023). On Singular Control for Lévy Processes. *Mathematics of Operations Research*. Publisher: INFORMS

¹³ Cohen et al. (2022). Optimal Ergodic Harvesting under Ambiguity. *SIAM Journal on Control and Optimization*

¹⁴ Ferrari et al. (May 2025). Stationary Mean-Field Games of Singular Control under Knightian Uncertainty.

The HJB equation

The dynamic programming principle suggests the existence of a function $V : \mathbb{R} \rightarrow \mathbb{R}$ and a constant γ satisfying the HJB equation

$$\min \left\{ c_D - V'(x), c_U + V'(x), c(x) + \sup_{(\kappa, \lambda) \in \Lambda} \{ (\mathbb{L}^{\kappa, \lambda} V)(x) \} - \gamma \right\} = 0.$$

- $\mathbb{P}^{\kappa, \lambda}$ -generator of X_t :

$$(\mathbb{L}^{\kappa, \lambda} f)(x) = (b(x) + \sigma \kappa(x) - r \mathbb{E}[Y])f'(x) + \frac{1}{2} \sigma^2 f''(x) + \lambda(x)(\Psi f)(x),$$

for $(\Psi f)(x) = \int_{\mathbb{R}} (f(x+y) - f(x)) dF(y)$.

- **Bang-bang ambiguity regions:** from the HJB, the optimal ambiguity functions (those where $\sup_{(\kappa, \lambda) \in \Lambda} \{ (\mathbb{L}^{\kappa, \lambda} V)(x) \}$ is attained) are given by

$$\kappa^*(x) := \begin{cases} \delta, & V'(x) \geq 0 \\ -\delta, & V'(x) < 0 \end{cases}, \quad \lambda^*(x) := \begin{cases} r(1 + \varepsilon), & (\Psi V)(x) \geq 0 \\ r(1 - \varepsilon), & (\Psi V)(x) < 0 \end{cases}$$

- **Constant-barrier Skorokhod reflections:**

A pair $(U^{\underline{x}}, D^{\bar{x}})$ for constants $\underline{x} < \bar{x}$, such that:

- ▶ $\bar{X} := X_t^{U^{\underline{x}}, D^{\bar{x}}}$ **stays inside the band** $[\underline{x}, \bar{x}]$ for all ambiguity functions:

$$\bar{X} \in [\underline{x}, \bar{x}], \quad \mathbb{P}^{\kappa, \lambda} \text{-a.s. for all } (\kappa, \lambda) \in \Lambda,$$

- ▶ **The Skorokhod complementary condition holds:**

(it's never optimal pushing simultaneously from both sides)

$$\int_0^\infty (\bar{X} - \underline{x}) dU_t^{\underline{x}} = \int_0^\infty (\bar{x} - \bar{X}) dD_t^{\bar{x}} = 0.$$

Guess (CB-SR)

The optimal controls are among constant-barriers Skorokhod reflections:

$$(U^*, D^*) = (U^{\underline{x}}, D^{\bar{x}})$$

We use the CB-SR guess to **build and verify a candidate solution.**

Verification theorem

Suppose that $V : \mathbb{R} \rightarrow \mathbb{R}$ and $\gamma \in \mathbb{R}$ solve the free-boundary problem

$$\left\{ \begin{array}{ll} \mathbb{L}^*V(x) + c(x) = \gamma & x \in [\underline{x}, \bar{x}], & (1a) \\ \mathbb{L}^*V(x) + c(x) \geq \gamma & x \notin [\underline{x}, \bar{x}], & (1b) \\ -c_U \leq V'(x) \leq c_D, & x \in (\underline{x}, \bar{x}), & (1c) \\ V'(x) = -c_U, & x \leq \underline{x}, & (1d) \\ V'(x) = c_D, & x \geq \bar{x}, & (1e) \\ V \in C^2(\mathbb{R}), & & (1f) \end{array} \right.$$

for $\mathbb{L}^* = \mathbb{L}^{\kappa^*, \lambda^*}$. Then,

$$\gamma = \inf_{(U,D) \in \mathcal{A}} \sup_{(\kappa, \lambda) \in \Lambda} J_x(U, D, \kappa, \lambda) = \sup_{(\kappa, \lambda) \in \Lambda} \inf_{(U,D) \in \mathcal{A}} J_x(U, D, \kappa, \lambda) = J_x(U^*, D^*, \kappa^*, \lambda^*),$$

NEGATIVE JUMPS

Bang-bang singletons

If

- **FBP:** V satisfies the FBP (1a)-(1e);
- **Convexity guess:** V' increases in the inaction region $[\underline{x}, \bar{x}]$; (later verified)
- **Negative jumps:** $\text{supp}(F) = (-\infty, 0]$ and $\mathbb{P}(Y < 0) > 0$.

Then, κ^* and λ^* take the form

$$\kappa^*(x) = \begin{cases} \delta, & x \geq x^\kappa \\ -\delta, & x < x^\kappa \end{cases}, \quad \lambda^*(x) = \begin{cases} r(1 + \varepsilon), & x \leq x^\lambda \\ r(1 - \varepsilon), & x > x^\lambda \end{cases}, \quad (2)$$

for unique values x^κ and x^λ such that one of these regimes hold

$$\text{Regime 1: } \underline{x} < x^\kappa < x^\lambda < \bar{x}, \quad \text{Regime 2: } \underline{x} < x^\kappa < \bar{x} \leq x^\lambda.$$

Moreover,

$$\text{Regime 1} \iff \mathbb{E}[V(\bar{x} + Y) - V(\bar{x})] < 0.$$

VISUALIZING THE ROBUST ERGODIC SINGULAR PROBLEM (REGIME 1)

GUESSING AND VERIFYING A SOLUTION: BUILDING THE CANDIDATE SOLUTION

Building a candidate solution

- **Fix boundaries:** Take $\underline{x}_*, \bar{x}_*, x_*^\kappa, x_*^\lambda \in \mathbb{R}$.
- **Define ambiguous functions:** κ^* and λ^* as in (2).
- **Free-boundary problem:** For $\mathbb{L}^* = \mathbb{L}^{\kappa^*, \lambda^*}$, assume that H_* solves the FBP

$$\left\{ \begin{array}{ll} (\mathbb{L}^* H_*)(x) + c'(x) = 0 & x \in (\underline{x}_*, \bar{x}_*) \setminus \{x_*^\kappa, x_*^\lambda\}, \quad (3a) \\ H_*(x) = -c_U, & x \leq \underline{x}_*, \quad (3b) \\ H_*(x) = c_D, & x \geq \bar{x}_*, \quad (3c) \\ H_*(x_*^\kappa) = 0, & (3d) \\ \int_{-\infty}^0 H_*(x_*^\lambda + y) F(y) dy = 0, & (3e) \\ H_* \in C^1(\mathbb{R}), & (3f) \\ H_* \in C^3((\underline{x}_*, \bar{x}_*) \setminus \{x_*^\kappa, x_*^\lambda\}), & (3g) \end{array} \right.$$

Candidate solution

We define our candidate solution V_* and γ_* such that

$$V_*' = H_*, \quad \gamma_* := -c_U(b - \delta\sigma - r\mathbb{E}[Y]) - c_U(1 + \varepsilon)r\mathbb{E}[Y] + c(\underline{x}_*).$$

Assumptions

- **Negative jumps:** $\text{supp}(F) = (-\infty, 0]$ and $\mathbb{P}(Y < 0) > 0$.
- **Constant drift:** $b(\cdot) \equiv b$.
- **Regular convex cost:** $c : \mathbb{R} \rightarrow \mathbb{R}$ is $C^2(\mathbb{R})$, convex everywhere and strictly convex in $[\underline{x}_*, \bar{x}_*]$, and it attains its unique minimum at $x = 0$.
- **Convexity of V_* at ambiguity thresholds:** $H'_*(x_*^\kappa) > 0, H'_*(x_*^\lambda) > 0$.

Verification theorem

Under the previous assumptions, the pair V_* and γ_* solve the FBP (1a)-(1f).

EXAMPLE:

NEGATIVE-EXPONENTIAL JUMP-SIZES AND PARABOLIC COST

Running assumptions

- **Negative-exponential jumps:** $F(y) = e^{\mu(y \wedge 0)}$, for all $y \in \mathbb{R}$, and for some $\mu > 0$.
- **Parabolic cost:** $c(x) = x^2$.
- **Constant drift:** $b(\cdot) \equiv b$.
- **Regime 1:** $\underline{x}_* < x_*^\kappa < x_*^\lambda < \bar{x}_*$. Regime 2 follows identical arguments.

Sufficient condition for Regime 1

The following condition suffice for Regime 1:

$$b \geq 0, \quad \varepsilon < c_D / (c_D + c_U), \quad (r/\mu)^{1/3} > \max\{2K_3/K_1, (2K_2/K_1)^{1/3}\},$$

for

$$K_1 := c_D - (c_D + c_U)\varepsilon, \quad K_2 := c_U(b + \delta\sigma) + 3((c_D + c_U)\sigma^2/4)^{2/3},$$

$$K_3 := 3((c_D + c_U)/4)^{2/3}(2(1 + \varepsilon)/\mu)^{2/3}.$$

Candidate FBP

Under the running assumptions, there exists constants $\{a_{i,j}\}_{i,j=1}^3$ and $\{b_j\}_{j=0}^2$ such that, for

$$I_1^* = (\underline{x}_*, x_*^k), \quad I_2^* := (x_*^k, x_*^\lambda), \quad I_3^* := (x_*^\lambda, \bar{x}_*),$$

$H_* = V_*$ and γ_* solves, in the inaction region, the piece-wise constant-coefficient ODE

$$0 = a_{i,3}H_*''(x) + a_{i,2}H_*'(x) + a_{i,1}H_*(x) + \mu x^2 + 2x - \mu\gamma_*, \quad x \in I_i^*,$$

while outside the inaction region,

$$H_* \equiv -c_U, \quad \text{on } (-\infty, \underline{x}_*], \quad H_* \equiv c_D, \quad \text{on } [\bar{x}_*, \infty),$$

and the smooth-fit condition holds:

$$H_* \in C^1(\mathbb{R}).$$

$$0 = a_{i,3}H_*''(x) + a_{i,2}H_*'(x) + a_{i,1}H_*(x) + \mu x^2 + 2x - \mu\gamma_*$$



ODE solution

$H \equiv H_i$ on each I_i^* , with

$$H_i(x) = \mathbf{c}_i^- e^{-\rho_i^- x} + \mathbf{c}_i^+ e^{-\rho_i^+ x} + p_i(x, \gamma_*),$$

for the unknown coefficients \mathbf{c}_i^\pm and the x -polynomial $p_i(x, \gamma_*)$ such that:

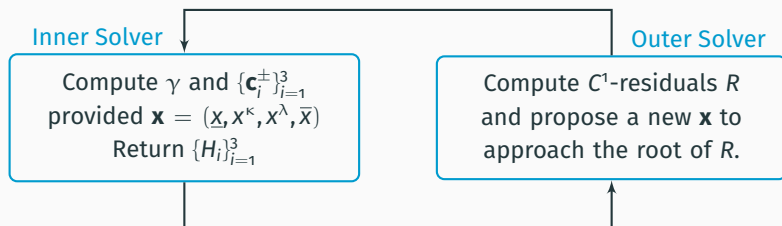
- $p_i(x)$ is of second degree iff $a_{i,1} \neq 0$;
- $p_i(x)$ is of third degree iff $a_{i,1} = 0$.

Exponents ρ_i^\pm and polynomials' coefficients p_i are known and explicit from parameters. The polynomials p_i are linear in γ .

We need to find 11 unknowns

- **Ergodic value:** γ_* .
- **6 ODEs' constants:** $\{\mathbf{c}_i^\pm\}_{i=1}^3$.
- **2 ambiguity thresholds:** x_*^κ and x_*^λ .
- **2 reflection barriers:** \underline{x}_* and \bar{x}_* .

Strategy: two-stages solvers



Inner Solver

- Obtain γ by solving $\mathbb{L}^{\kappa, \lambda} V(\underline{x}+) = \mathbf{o}$:

$$\gamma = \frac{1}{2} \sigma^2 H'_1(\underline{x}, \gamma) + c_U(\delta \sigma - b - \epsilon r \mathbb{E}[Y]) + c(\underline{x}).$$

- Obtain $\{\mathbf{c}_i^{\pm}\}_{i=1}^3$ by solving the $H \in C^0(\mathbb{R})$ linear system

$$\begin{cases} H_1(\underline{x}) = -c_U, & H_3(\bar{x}) = c_D, & H_1(x^\kappa) = H_2(x^\kappa), & H_2(x^\lambda) = H_3(x^\lambda), \\ H_1(x^\kappa) = \mathbf{o}, & \int_{-\infty}^{\mathbf{o}} H(x^\lambda + y) \mu e^{\mu y} dy = \mathbf{o}. \end{cases}$$

Outer Solver

Repeat 1-2 until $R \approx (\mathbf{o}, \mathbf{o}, \mathbf{o}, \mathbf{o})$.

- Given $\mathbf{x} = (\underline{x}, x^\kappa, x^\lambda, \bar{x})$, use Inner Solver to compute C^1 residuals:

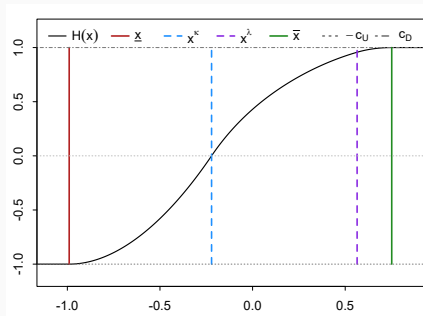
$$R = (H'_1(\underline{x}), H'_3(\bar{x}), H'_1(x^\kappa) - H'_2(x^\kappa), H'_2(x^\lambda) - H'_3(x^\lambda)).$$

- Propose a new \mathbf{x} using a root-search updating algorithm (Broyden-Newton).

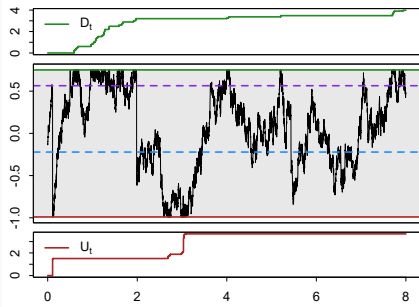
NUMERICS

BOUNDARIES, CONTROLLED PROCESS AND CONTROL POLICIES.

Baseline parameters: $b = 0$, $\delta = 1$, $r = 1$, $\varepsilon = 0.5$, $\sigma = 1$, and $\mu = 1$.



(a) H_* with reflecting barriers and ambiguity thresholds



(b) Reflected path \bar{X}

Figure 1: Image (a): numerical computation of the function H_* with the corresponding ambiguity thresholds x^K and x^λ , and the reflecting barriers \underline{x} and \bar{x} . **Image (b)** controlled process \bar{X}_t . We used $c_U = 1$, $c_D = 1$.

COMPARATIVE STATISTICS. VARYING THE DRIFT

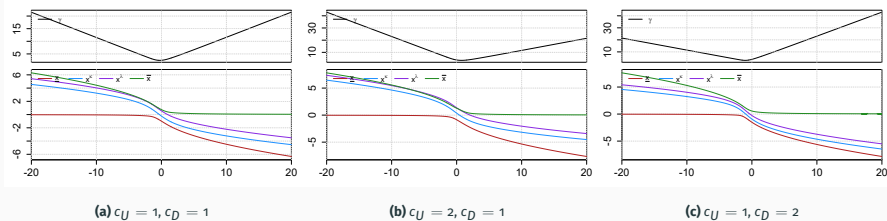


Figure 2: Reflecting barriers and ambiguity thresholds as b varies.

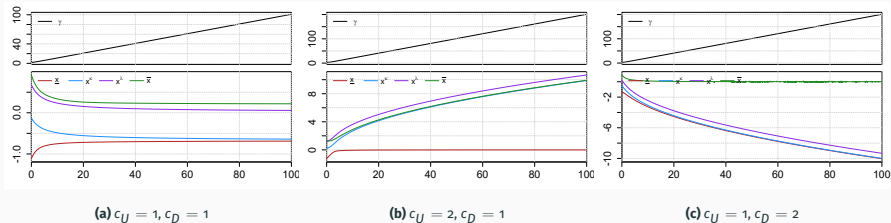


Figure 3: Reflecting barriers and ambiguity thresholds as δ varies.

COMPARATIVE STATISTICS. VARYING THE INTENSITY

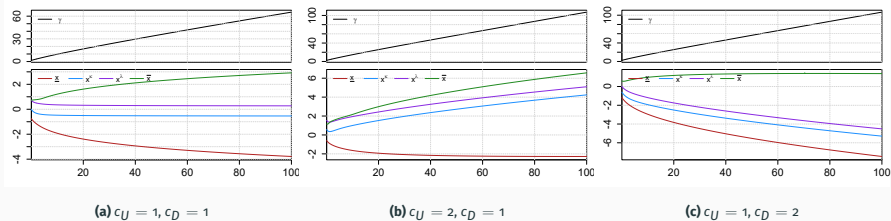


Figure 4: Reflecting barriers and ambiguity thresholds as r varies.

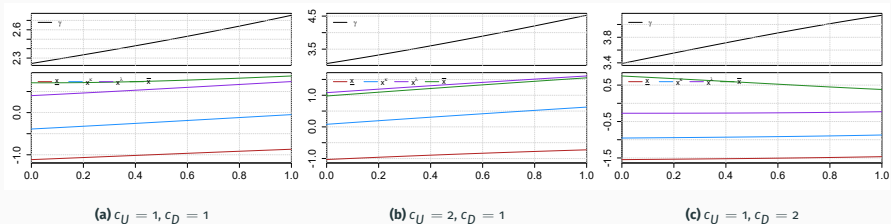


Figure 5: Reflecting barriers and ambiguity thresholds as ϵ varies.

COMPARATIVE STATISTICS- VARYING THE DIFFUSION AND THE JUMP-SIZE MEAN

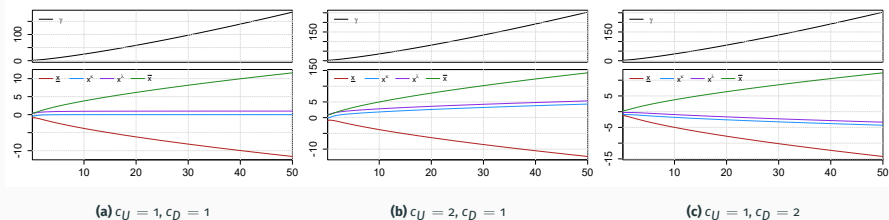


Figure 6: Reflecting barriers and ambiguity thresholds as σ varies.

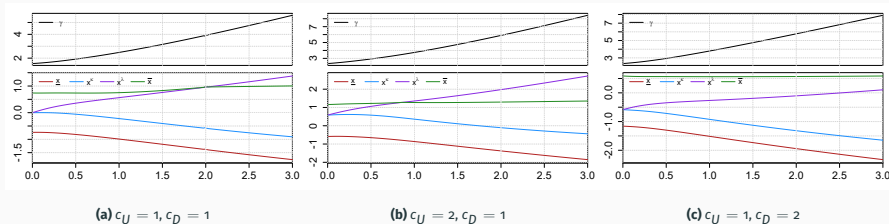


Figure 7: Reflecting barriers and ambiguity thresholds as $-\mathbb{E}[Y] = 1/\mu$ varies.

COMPARATIVE STATISTICS. VARYING THE COSTS

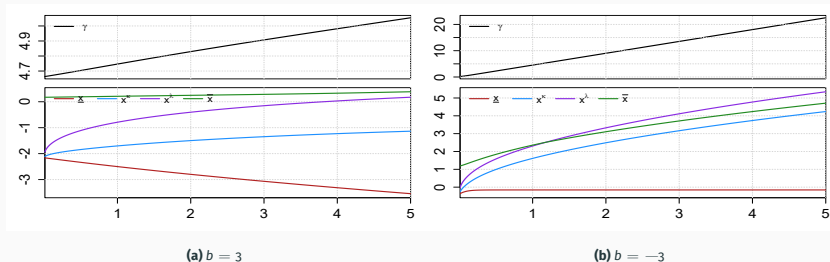


Figure 8: Reflecting barriers and ambiguity thresholds as c_U varies.

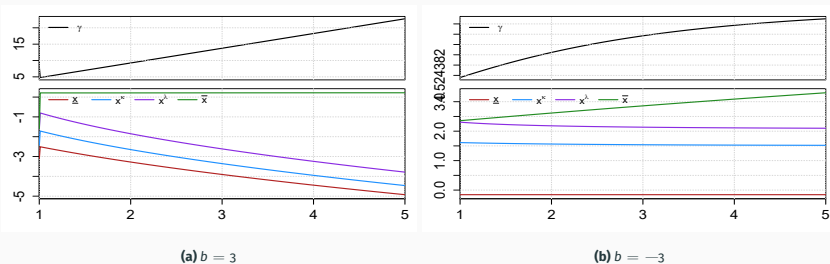


Figure 9: Reflecting barriers and ambiguity thresholds as c_D varies.

CONTROLLER-ADVERSARY: THE ESSENTIAL FEEDBACK LOOP

Controller (robust reflection)

- **Tolerate self-correcting moves:** relax where drift/jumps tend to pull back.
- **Block persistent risk:** tighten where worst-case dynamics can keep pushing into costly states.
- **Buffer the expensive action:** tilt the band toward the cheaper intervention.
- **Frequency vs exposure:** widen when volatility/shocks make reflection too frequent; shrink when adverse trends make exposure too costly.

Adversary (least-favorable switching)

- **Make damage stick:** select drift/intensity regimes that sustain deviations from low-cost regions.
- **Push toward costly reflection:** create one-sided pressure toward the boundary tied to the expensive control.
- **Remove helpful corrections:** downweight dynamics that would self-correct expensive excursions.

Gracias!