

# On the McKean game driven by a spectrally negative Lévy process

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(Joint work in progress with Erik Baurdoux, LSE, UK)

# Framework of a Dynkin game

- Filtered probability space  $(\Omega, \mathcal{F}, \mathbf{F} = (\mathcal{F}_t)_{t \in [0, T]}, \mathbb{P})$ ,  $\mathbf{F}$  satisfying the usual conditions. Let  $\tau$  and  $\sigma$  denote  $\mathbf{F}$ -stopping times
- Two processes  $L$  ('lower payoff') and  $U$  ('upper payoff') with  $L_t \leq U_t, \forall t \in [0, T)$  and  $L_T = U_T$
- A *maximizer* choosing  $\sigma$  (max. expected lower payoff), a *minimizer* choosing  $\tau$  (min. expected upper payoff). Resulting payoff to the maximizer is:

$$R_{\sigma, \tau} := \mathbf{1}_{\{\sigma \leq \tau\}} L_{\sigma} + \mathbf{1}_{\{\tau < \sigma\}} U_{\tau}$$

(i.e. whoever exercises first determines the payoff)

# Specifics McKean game driven by spectr. neg. Lévy process

The McKean game:

- Let  $S = \exp(X)$  for a *spectr. neg. LP*  $X$  and let  $\mathbb{P}_s$  be such that  $\mathbb{P}_s(S_0 = s) = 1$
- Lower payoff given by  $L_t = e^{-rt}(K - S_t)^+$
- Upper payoff given by  $U_t = e^{-rt}((K - S_t)^+ + \delta)$ ,  $t \in [0, T)$  and  $L_T = U_T$ , where  $K, r, \delta > 0$
- Assume  $0 \leq \psi(1) \leq r$  (where  $\psi$  Laplacian of  $X$ )
- Value function:

$$\begin{aligned} v(s) &= \sup_{\sigma} \inf_{\tau} \mathbb{E}_s [\mathbf{1}_{\{\sigma \leq \tau\}} L_{\sigma} + \mathbf{1}_{\{\tau < \sigma\}} U_{\tau}] \\ &= \inf_{\tau} \sup_{\sigma} \mathbb{E}_s [\mathbf{1}_{\{\sigma \leq \tau\}} L_{\sigma} + \mathbf{1}_{\{\tau < \sigma\}} U_{\tau}] \end{aligned}$$

- $v_P$  value function of the American put, i.e.

$$v_P(s) = \sup_{\sigma} \mathbb{E}_s [L_{\sigma}]$$

- If  $\delta \geq v_P(K)$  then game degenerates to Am. put (easy to see)  $\Rightarrow$  assume  $\delta \leq v_P(K)$

### Theorem (Baudoux & Kyprianou (2008))

- *A saddle point  $(\sigma^*, \tau^*)$  exists, where*

$$\sigma^* = \inf\{t > 0 \mid S_t < s_h\} \text{ for some } s_h \in (0, K)$$

$$\tau^* = \inf\{t > 0 \mid S_t \in [K, s_w]\} \text{ for some } s_w \in [K, \infty)$$

- *Expressions for  $s_h$ , for  $v$  on  $(0, K]$  & smooth pasting results.*

- *Question*: when  $s_w = K$  and when  $s_w > K$ ?
- *Answer*: depends on the structure of  $X$
- *Idea of proof*: look at optimal response to the minimizer choosing  $T_K := \inf\{t > 0 \mid S_t = K\}$ , i.e.:

$$v_{max}(s) := \sup_{\sigma} \mathbb{E}_s [R_{\sigma, T_K}] = \mathbb{E}_s [R_{\sigma^*, T_K}]$$

Then:

- if  $\forall s : v_{max}(s) \leq (K - s)^+ + \delta$  then  $s_w = K$  (since  $v \leq v_{max}$ ) and  $v = v_{max}$
  - otherwise  $v \neq v_{max} \Rightarrow s_w > K$
- *Next up*: graphs to illustrate 4 different cases.  
 Red curve =  $v_{max}$ , Green curve =  $v$  when  $s_w > K$

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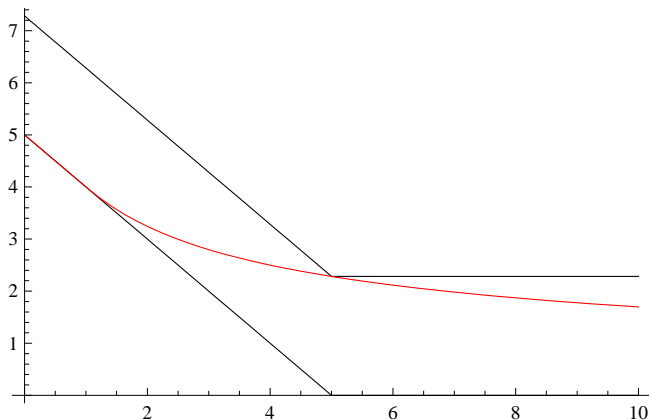
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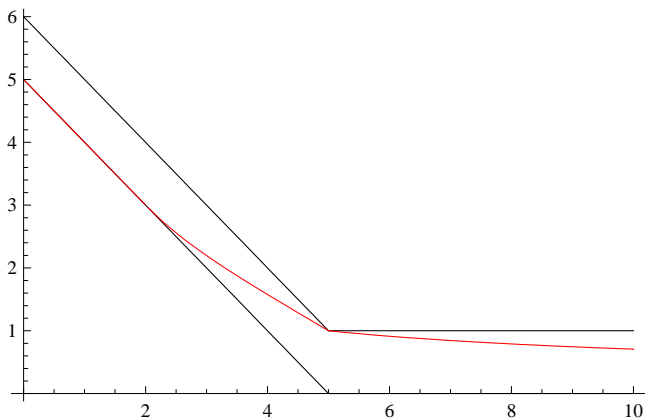
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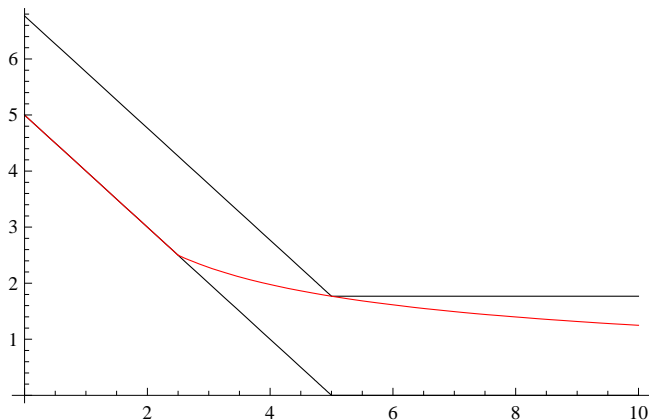
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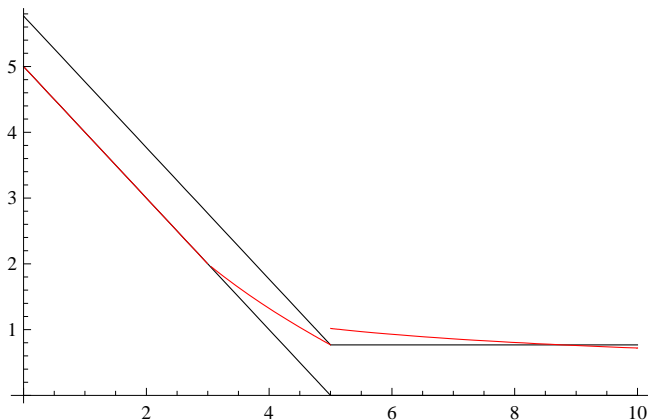
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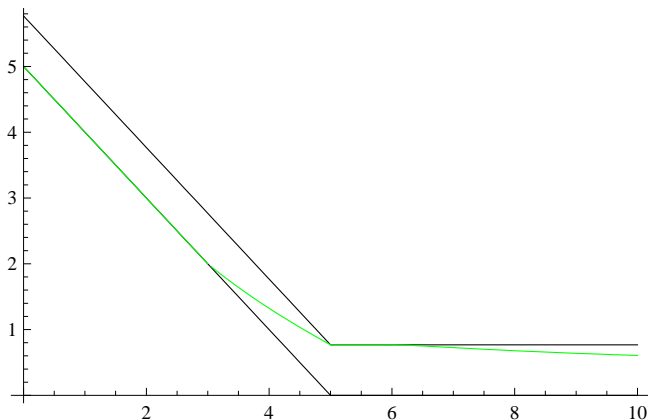
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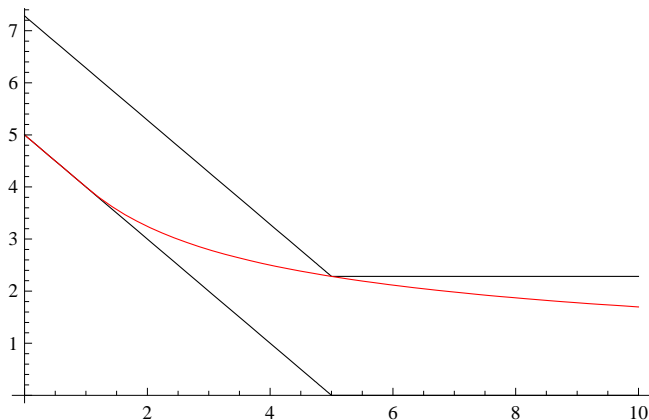
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$\delta < v_P(K)$ , actual value function:



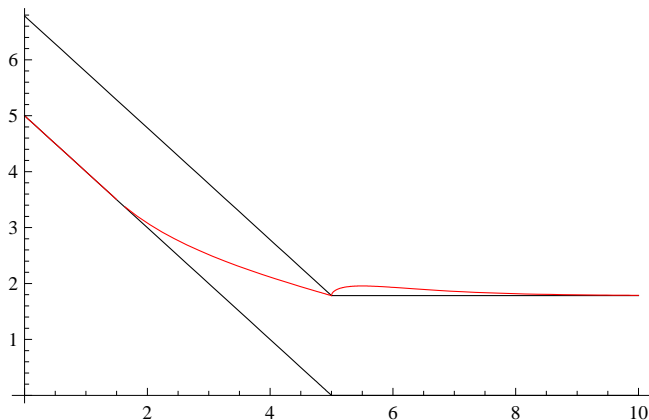
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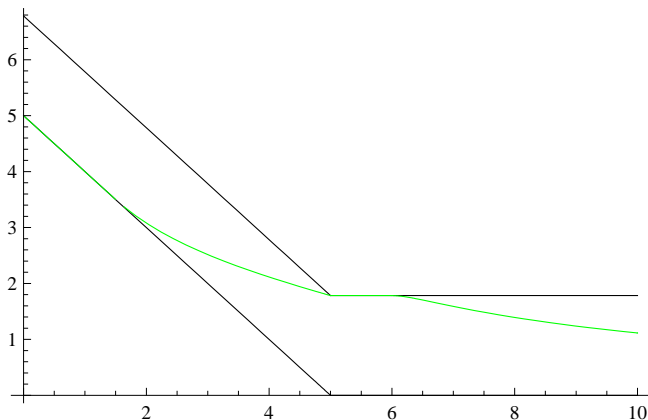
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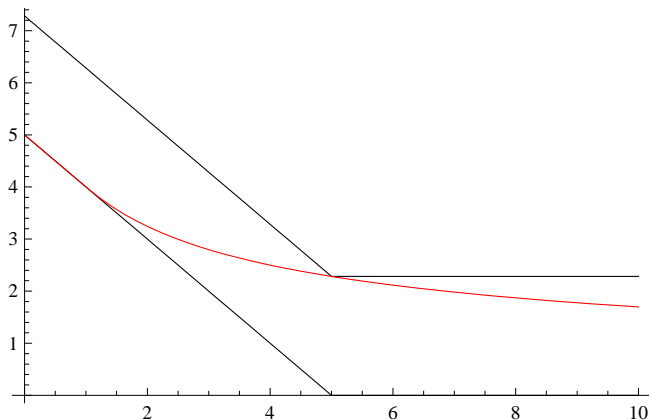
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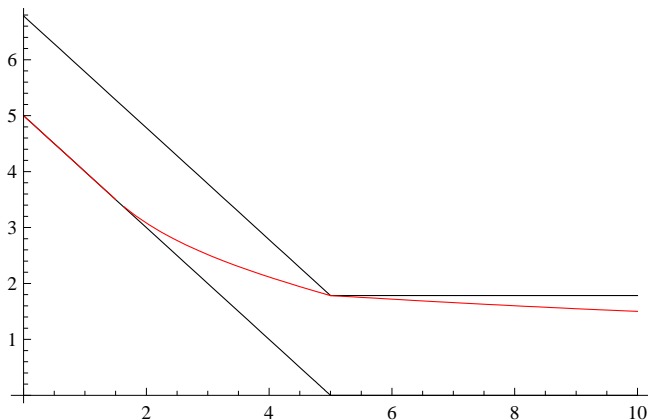
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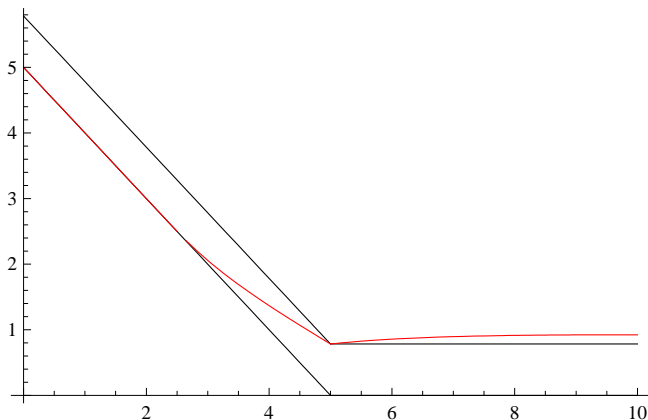
$$\delta = v_P(K) - \varepsilon:$$





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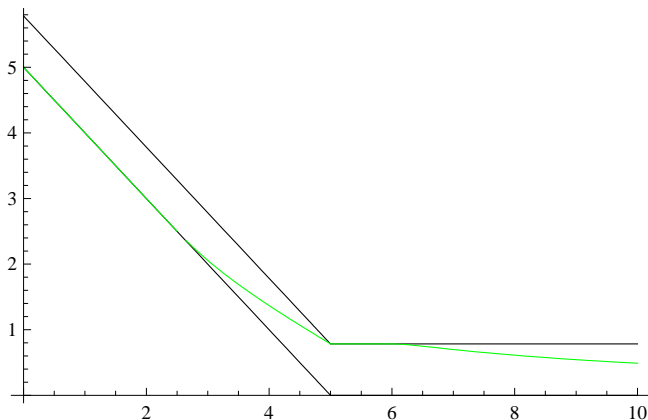
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## With Gaussian part (Baurdoux & KvS)

$\delta \ll v_P(K)$ , actual value function:



### Theorem (Baurdoux & KvS (in progress))

*If  $X$  has a positive Gaussian component and Lévy measure  $\Pi \neq 0$ , a threshold  $\delta_0 \in (0, v_P(K))$  exists s.t.*

- *if  $\delta \in (0, \delta_0)$  then  $s_w > K$*
- *if  $\delta \in [\delta_0, v_P(K)]$  then  $s_w = K$ .*

Some additional results:

- $\delta \mapsto s_w(\delta)$  is cts. and decreasing
- $\lim_{\delta \uparrow \delta_0} s_w(\delta) = K$ ,  $\lim_{\delta \downarrow 0} s_w(\delta) = K - \log(\inf \text{supp}(\Pi))$
- expressions for  $s_w$  and  $v$  (more explicit in jump-diffusion case)

# Finite expiry case ( $T < \infty$ ), some examples

Numerical method: *Canadization* (Carr (1998)).

- discretise time stochastically, i.e. take grid points with distance iid exponentials  $(\xi_i^{(n)})_{i \geq 1}$  with param.  $n$ , indep. of  $S$
- define:  $v_k^{(n)}$  = value of game with expiry date  $\sum_{i=0}^k \xi_i^{(n)}$
- convergence: if  $(k(n))$  s.t.  $k(n)/n \rightarrow T$  as  $n \rightarrow \infty$ , then

$$\sum_{i=0}^{k(n)} \xi_i^{(n)} \xrightarrow{\text{a.s.}} T \implies v_{k(n)}^{(n)}(s) \rightarrow v(T, s)$$

- algorithm for computing  $(v_k^{(n)})_{k \geq 0}$ : set  $v_0^{(n)}(s) := (K - s)^+$ . For  $k \geq 1$ ,  $v_k^{(n)}$  is a game with payoff:

$$\mathbf{1}_{\{\sigma \wedge \tau < \xi_n\}} R_{\sigma, \tau} + \mathbf{1}_{\{\sigma \wedge \tau \geq \xi_n\}} e^{-r\xi_n} v_{k-1}^{(n)}(S_{\xi_n}).$$

$\xi_n$  indep. exp.  $\implies$  perpetual game with terminal payoff, can be solved.

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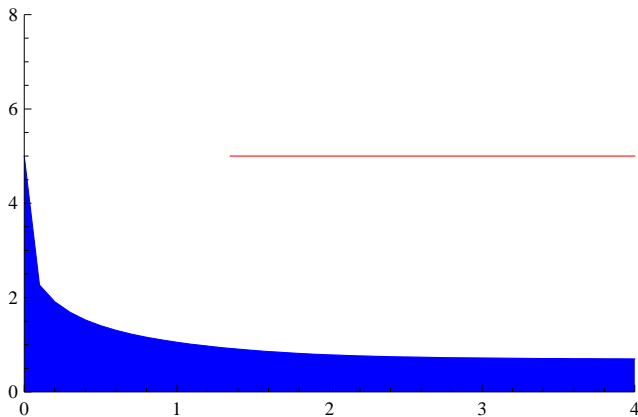
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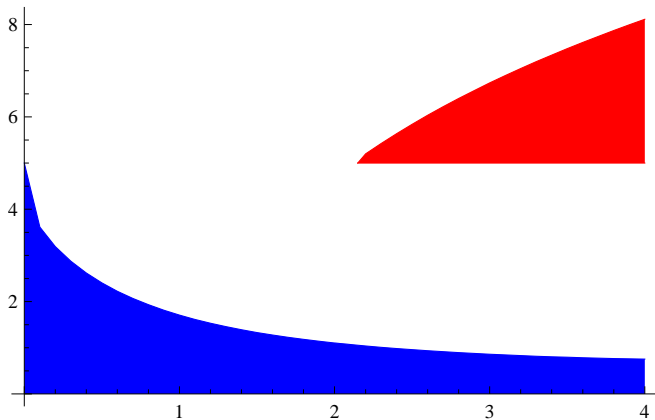
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## Brownian motion:

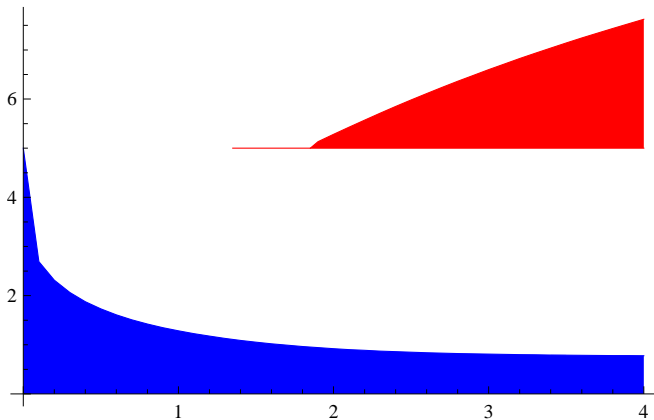




Jump-diffusion without Gaussian part:



Jump-diffusion with Gaussian part:



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