Optimal stopping problems arising from Mathematical Finance

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Outline

- Basic formulation
- Mathematical finance
- Markovian approach
- Free-boundary problem
- American call option
 - Black and Scholes model
 - Diffusion-type models

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where

• the supremum is taken over all stopping times τ \longrightarrow Random variables $\tau(\omega) < \infty$, $\mathbb{P} - a.s.$, such that $\{\omega \mid \tau(\omega) \leq t\} \in \mathcal{F}_t$ for all $t \geq 0$.

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- \exists Markov process $X = (X_t)_{t \geq 0}$ such that $G_t = G(t, X_t)$ for
 - some measurable function G(t,x),
 - $x \in E$,
 - *E* is the state space of *X*.

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Fundamental Theorem of Asset Pricing

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- G(x) is the payoff function of the contingent claim
- written on the underlying risky asset price process X
- *r* is the constant *discount rate* / *interest rate*
- ullet the supremum is taken over all stopping times au of X
- \mathbb{E}_x is taken under some Equivalent Martingale Measure \mathbb{P}_x for X given that $X_0 = x \in E$.

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- \longrightarrow Formulation by [Dynkin '63] 1

¹DYNKIN, E. B. (1963). The optimum choice of the instant for stopping a Markov process. Soviet Mathematical Doklady 4 (627-629)

Solution to the Optimal stopping problem

General theory of Optimal stopping for Markov processes 2

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General theory of Optimal stopping for Markov processes ²



• We have a set of domains C, such that

$$\mathcal{C} = \{x \in E \mid V(x) > G(x)\}$$
 (continuation region)
 $\mathcal{D} = \{x \in E \mid V(x) = G(x)\} = E \setminus \mathcal{C}$ (stopping region)

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Optimal stopping problem

$$V_*(x) = \sup_{\tau} \mathbb{E}_x[e^{-r\tau} G(X_\tau)]$$

is equivalent to the problem of

Finding the smallest superharmonic function \widetilde{V} which dominates G on E.



A traditional way is:

Free-boundary problem ³

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Basic idea: \widetilde{V} and \mathcal{C} (or \mathcal{D}) solve the free-boundary problem:

$$\begin{cases} \mathbb{L}_{X}\widetilde{V} - r\widetilde{V} = 0 & \text{in } \mathcal{C} \\ \widetilde{V}|_{\partial \mathcal{C}} = G|_{\partial \mathcal{C}} & \text{(continuous fit)} \\ \frac{\partial \widetilde{V}}{\partial x}|_{\partial \mathcal{C}} = \frac{\partial G}{\partial x}|_{\partial \mathcal{C}} & \text{(smooth fit)}^{4} \end{cases}$$

where \mathbb{L}_X is the infinitesimal generator of the process X.

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⁴MIKHALEVICH, V. S. (1958). A Bayes test of two hypotheses concerning the mean of a normal process. *Visnik Kiiv. Univ*, 1(101-104).

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The proof is based on the standard verification key arguments of the

- Change-of-variable formula (i.e. Itô's formula and its extensions) from the stochastic calculus
- Ooob's optional sampling theorem from the martingale theory
- etc...

Definition (American Call option)

Contracts which entitle the holder to buy an underlying asset X "on or before" a future expiry or maturity T time for a pre-specified amount called strike price K.

The payoff of a perpetual American Call option is therefore

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• The price is given by the solution of the following optimal stopping problem:

$$V_*(x) = \sup_{\tau} \mathbb{E}_x \left[e^{-r\tau} \left(X_{\tau} - K \right)^+ \right]$$

for all x > 0.

Free-boundary problem for American Call option

 It follows from the general theory of optimal stopping problems for Markov processes, that the optimal stopping time is given by

$$\tau_* = \inf\{t \ge 0 \mid V_*(X_t) = X_t - K\} = \inf\{t \ge 0 \mid X_t \ge b_*\}$$

for some $b_* > 0$ to be specified.

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• The equivalent free-boundary problem for $V_*(x)$ and b_* is given by

$$(r-\delta) \times V'(x) + \frac{\sigma^2}{2} \times^2 V''(x) = rV(x)$$
 for $x < b$
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• <u>SOLUTION</u>: The value function $V_*(x)$ and the optimal exercise boundary b_* are given by

$$V_*(x) = \begin{cases} \frac{b_*}{\gamma_1} \left(\frac{x}{b_*}\right)^{\gamma_1} &, & \text{for } 0 < x < b_* \\ x - K &, & \text{for } x \ge b_* \end{cases} , \text{ with } b_* = \frac{\gamma_1 K}{\gamma_1 - 1}$$

Sketch of Verification theorem

• Applying the local time-space formula to the solution $e^{-rt} V(X_t)$ combined with the smooth-fit condition and $P(X_t = b_*) = 0$, we get

$$e^{-rt} V(X_t) = V(x) + \int_0^t e^{-ru} \left((r - \delta) X_u V' + \frac{\sigma^2}{2} X_u^2 V'' - rV \right) (X_u) du + M_t$$

with $M = (M_t)_{t \geq 0}$ defined by

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• It follows from the free-boundary conditions for V with b_* that

$$((r-\delta)xV' + \frac{\sigma^2}{2}x^2V'' - rV)(x) \le 0$$
 and $V(x) \ge (x-K)^+$

hold for all x>0 and thus, for any stopping time τ of X started at x>0 we have

$$e^{-r(t\wedge\tau)}(X_{t\wedge\tau}-K)^+\leq e^{-r(t\wedge\tau)}V(X_{t\wedge\tau})\leq V(x)+M_{t\wedge\tau}$$

Verification theorem

• Taking the expectation with respect to \mathbb{P}_x (using Doob's optional sampling theorem) and letting t go to infinity (using Fatou's lemma), we obtain

$$\mathbb{E}_x \big[e^{-r\tau} \, (X_\tau - K)^+ \big] \leq V(x) \quad \text{for any stopping time τ}, \ \, \forall \, \, x > 0.$$

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- ullet By virtue of the structure of au_* , we observe that
 - If $x \ge b_*$, then we have equality in the above for τ_* instead of τ .

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If x < b*, then using the conditions of the free-boundary problem satisfied by
 V in the continuation region:

$$e^{-r(t\wedge au_*)} V(X_{t\wedge au_*}) = V(x) + M_{t\wedge au_*}$$

Taking the expectation with respect to \mathbb{P}_x (using Doob's optional sampling theorem) and letting t go to infinity (using Lebesgue dominated convergence), we obtain

$$\mathbb{E}_{x}[e^{-r\tau_{*}}(X_{\tau_{*}}-K)^{+}] = \mathbb{E}_{x}[e^{-r\tau_{*}}V(X_{\tau_{*}})] = V(x) \quad \forall \ x < b_{*}$$

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Options on the underlying risky asset

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• Russian option 5 payoff $G(X_t, S_t) = S_t$

⁵SHEPP, L., & SHIRYAEV, A. N. (1993). The Russian option: reduced regret. *The Annals of Applied Probability* **3**(3) (631–640).

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- Russian option ⁵ payoff $G(X_t, S_t) = S_t$
- American lookback option ^{6 7}
 - with fixed strike payoff $G(X_t, S_t) = (S_t K)^+$
 - with floating strike payoff $G(X_t, S_t) = (S_t K X_t)^+$

⁵SHEPP, L., & SHIRYAEV, A. N. (1993). The Russian option: reduced regret. *The Annals of Applied Probability* **3**(3) (631–640).

⁶PEDERSEN, J. L. (2000). Discounted optimal stopping problems for the maximum process. *Journal of Applied Probability* **37**(4) (972–983).

Other optimal stopping problems

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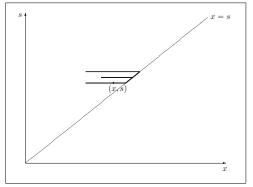


Figure 1. A computer drawing of the state space of the process (X, S)

 (Ω, \mathcal{F}, P) with a standard Brownian motion $B = (B_t)_{t \geq 0}$.

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Asset price process $X = (X_t)_{t \ge 0}$ with dynamics

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- $\delta(s)$, $\sigma(s) > 0$ are continuously differentiable bounded functions on $[0, \infty]^2$.
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Main AIM:

Optimal stopping problem for the valuation of an American Call option:

$$V_*(x,s) = \sup_{\tau} E_{x,s} [e^{-r\tau} (X_{\tau} - K)^+]$$

- supremum over all stopping times au with respect to the natural filtration of X,
- $E_{x,s}$ denotes the expectation under the assumption that (X,S) starts at (x,s).

The optimal stopping problem in (X, S)-setting

Main purpose: Solve the optimal stopping problem

$$V_*(x,s) = \sup_{\tau} E_{x,s} \left[e^{-r\tau} \left(X_{\tau} - K \right)^+ \right]$$

General theory of optimal stopping for Markov processes

⇒ Optimal stopping time is

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The free-boundary problem for Call option

From the general theory of optimal stopping problems for Markov processes we formulate the equivalent free-boundary problem

$$(r - \delta(s)) \times \partial_x V(x, s) + \frac{\sigma^2(s)}{2} \times^2 \partial_{xx} V(x, s) = rV(x, s)$$
 for $0 < x < b(s) \wedge s$
 $V(x, s)\big|_{x=0+} = 0$ (natural boundary)
 $V(x, s)\big|_{x=b(s)-} = b(s) - K$ (continuous fit)
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when $b_*(s) \leq s$ holds.

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when $b_*(s) \leq s$ holds.

Otherwise, when $s < b_*(s)$ holds, we have

$$V(x,s)\big|_{x=0+} = 0$$
 (natural boundary)
 $\partial_s V(x,s)\big|_{y=s-} = 0$ (normal reflection)⁸

⁸DUBINS, L., SHEPP, L. A. and SHIRYAEV, A. N. (1993). Optimal stopping rules and maximal inequalities for Bessel processes. *Theory of Probability and Applications* 38(2)

State space of the process (X, S)

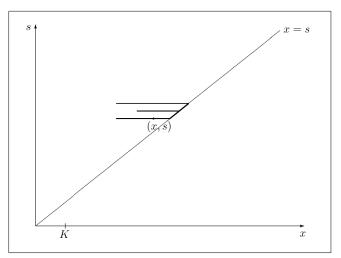


Figure 2. A computer drawing of the state space of the process (X, S)

Free-boundary function for Call option

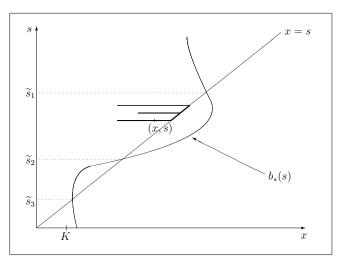


Figure 3. A computer drawing of the state space of the process (X, S) and the boundary function $b_*(s)$.

Solution of the free-boundary problem for Call

The value function takes the form

$$V(x,s;b_*(s)) = \frac{b_*(s)}{\beta_1(s)} \left(\frac{x}{b_*(s)}\right)^{\beta_1(s)} \quad \text{with} \quad b_*(s) = \frac{\beta_1(s)K}{\beta_1(s)-1}$$

for $0 < x < b_*(s) \le s$ and s > K, where

$$\beta_1(s) = \frac{1}{2} - \frac{r - \delta(s)}{\sigma^2(s)} + \sqrt{\left(\frac{1}{2} - \frac{r - \delta(s)}{\sigma^2(s)}\right)^2 + \frac{2r}{\sigma^2(s)}} > 1,$$

Solution of the free-boundary problem for Call

The value function takes the form

$$V(x,s;b_*(s)) = \frac{b_*(s)}{\beta_1(s)} \left(\frac{x}{b_*(s)}\right)^{\beta_1(s)} \quad \text{with} \quad b_*(s) = \frac{\beta_1(s)K}{\beta_1(s)-1}$$

for $0 < x < b_*(s) \le s$ and s > K, where

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and the form

$$V(x,s;\widetilde{s}_{2k-1}) = \exp\left(-\int_{s}^{\widetilde{s}_{2k-1}} \beta_{1}'(q) \ln q \, dq\right) \frac{\widetilde{s}_{2k-1}^{1-\beta_{1}(s_{2k-1})}}{\beta_{1}(\widetilde{s}_{2k-1})} \, x^{\beta_{1}(s)}$$

for $0 < x < s < b_*(s)$.



Free-boundary function for Call option

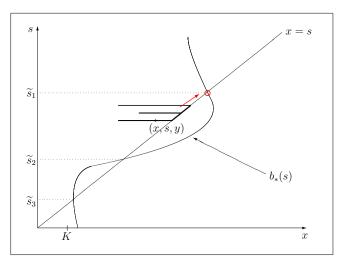


Figure 3. A computer drawing of the state space of the process (X, S) and the boundary function $b_*(s)$.

Main Result of the Call option in the (X,S)-setting

Theorem

The value function of the opt. stop. prob. for Call option is given by

$$V_{*}(x,s) = \begin{cases} V(x,s;b_{*}(s)), & \text{if } 0 < x < b_{*}(s) \leq s \\ V(x,s;\tilde{s}), & \text{if } 0 < x \leq s < b_{*}(s) \\ x - K, & \text{if } b_{*}(s) \leq x \leq s \end{cases}$$

and the optimal stopping time is

$$\tau_* = \inf\{t \geq 0 \,|\, X_t \geq b_*(S_t)\}$$

where

$$b_*(s) = \frac{\beta_1(s)K}{\beta_1(s) - 1}$$

 (Ω, \mathcal{F}, P) with a standard Brownian motion $B = (B_t)_{t \geq 0}$.

 (Ω, \mathcal{F}, P) with a standard Brownian motion $B = (B_t)_{t \geq 0}$.

Asset price process $X = (X_t)_{t \ge 0}$ with dynamics

$$dX_t = (r - \delta(S_t, Y_t)) X_t dt + \sigma(S_t, Y_t) X_t dB_t \quad (X_0 = x > 0)$$

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where $\delta(s,y)$, $\sigma(s,y) > 0$ are continuously differentiable bounded functions on $[0,\infty]^2$.

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- running maximum process $S = (S_t)_{t \geq 0}$ defined by

$$S_t = \max_{0 \le u \le t} X_u \vee s$$

- running maximum drawdown process $Y = (Y_t)_{t \ge 0}$ defined by

$$Y_t = \max_{0 \le u \le t} (S_u - X_u) \vee y$$

 (Ω, \mathcal{F}, P) with a standard Brownian motion $B = (B_t)_{t \geq 0}$.

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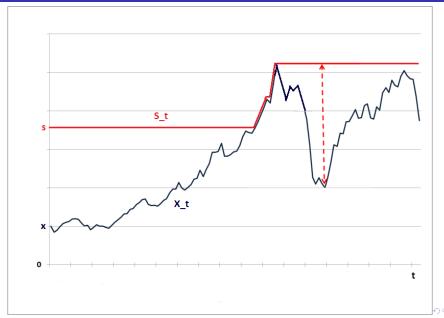
- running maximum drawdown process $Y = (Y_t)_{t \ge 0}$ defined by

$$Y_t = \max_{0 \le u \le t} (S_u - X_u) \vee y$$

• 3-D state space of the process (X, S, Y): $0 < s - y \le x \le s$.



Processes X, S and Y



State space of the process (X, S, Y)

Recall that: $0 < s - y \le x \le s$.

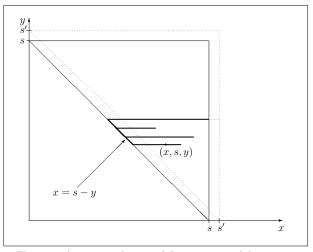


Figure 1. A computer drawing of the state space of the process (X, S, Y), for some s fixed, which increases to s'.

Optimal stopping problem for the process (X, S, Y)

Main AIM:

Optimal stopping problem for the valuation of an American Call option for the process (X, S, Y): ⁹

$$V_*(x, s, y) = \sup_{\tau} E_{x, s, y} [e^{-r\tau} (X_{\tau} - K)^+]$$

- supremum over all stopping times τ wrt to the natural filtration of X
- $E_{x,s,y}$ denotes the expectation under the assumption that (X, S, Y) starts at (x, s, y).

⁹GAPEEV P.V. & RODOSTHENOUS N. (2014) Optimal stopping problems in diffusion-type models with running maxima and drawdowns. *Journal of Applied_Probability* **51**(3) (799–817) 0.000

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• Other problems for the drawdown process:

Pospisil, Vecer, and Hadjiliadis (2009), Mijatovic and Pistorius (2012), Zhang (2015), among others.

⁹GAPEEV P.V. & RODOSTHENOUS N. (2014) Optimal stopping problems in diffusion-type models with running maxima and drawdowns. *Journal of Applied_Probability* **51**(3) (799–817) \(\) \(\) \(\)

The optimal stopping problem in the (X,S,Y)-setting

$$V_*(x, s, y) = \sup_{\tau} E_{x, s, y} \left[e^{-r\tau} \left(X_{\tau} - K \right)^+ \right]$$

General theory of optimal stopping for Markov processes

 \Rightarrow Optimal stopping time is

$$\tau_* = \inf\{t \ge 0 \mid V_*(X_t, S_t, Y_t) = X_t - K\}$$

The optimal stopping problem in the (X,S,Y)-setting

$$V_*(x, s, y) = \sup_{\tau} E_{x, s, y} \left[e^{-r\tau} \left(X_{\tau} - K \right)^+ \right]$$

General theory of optimal stopping for Markov processes

⇒ Optimal stopping time is

$$au_* = \inf\{t \geq 0 \mid V_*(X_t, S_t, Y_t) = X_t - K\}$$
 ψ
 $au_* = \inf\{t \geq 0 \mid X_t \geq b_*(S_t, Y_t)\}$

for some function $K < b_*(s, y)$ to be determined.

The optimal stopping problem in the (X,S,Y)-setting

$$V_*(x, s, y) = \sup_{\tau} E_{x, s, y} \left[e^{-r\tau} \left(X_{\tau} - K \right)^+ \right]$$

General theory of optimal stopping for Markov processes

⇒ Optimal stopping time is

$$\tau_* = \inf\{t \ge 0 \mid V_*(X_t, S_t, Y_t) = X_t - K\}$$

$$\downarrow \downarrow$$

$$\tau_* = \inf\{t \ge 0 \mid X_t \ge b_*(S_t, Y_t)\}$$

for some function $K < b_*(s, y)$ to be determined.

The free-boundary problem

From the general theory of optimal stopping problems for Markov processes we formulate the equivalent free-boundary problem

$$(r - \delta(s, y)) \times \partial_x V(x, s, y) + \frac{\sigma^2(s, y)}{2} \times^2 \partial_{xx} V(x, s, y) = rV(x, s, y)$$
 such that $s - y < x < b(s, y) \wedge s$
$$\partial_y V(x, s, y)\big|_{x = (s - y) +} = 0$$
 (normal reflection)
$$V(x, s, y)\big|_{x = b(s, y) -} = b(s, y) - K$$
 (continuous fit)
$$\partial_x V(x, s, y)\big|_{x = b(s, y) -} = 1$$
 (smooth fit)

when $b_*(s, y) \leq s$ holds.

The free-boundary problem

From the general theory of optimal stopping problems for Markov processes we formulate the equivalent free-boundary problem

$$(r-\delta(s,y)) \times \partial_x V(x,s,y) + \frac{\sigma^2(s,y)}{2} \times^2 \partial_{xx} V(x,s,y) = rV(x,s,y)$$
 such that $s-y < x < b(s,y) \wedge s$
$$\partial_y V(x,s,y)\big|_{x=(s-y)+} = 0 \qquad \qquad (normal\ reflection)$$

$$V(x,s,y)\big|_{x=b(s,y)-} = b(s,y) - K \qquad (continuous\ fit)$$

$$\partial_x V(x,s,y)\big|_{x=b(s,y)-} = 1 \qquad (smooth\ fit)$$
 when $b_*(s,y) \leq s$ holds.

*()3 / =

Otherwise, when
$$s < b_*(s,y)$$
 holds, we have
$$\left. \partial_y V(x,s,y) \right|_{x=(s-y)+} = 0 \qquad \qquad (normal\ reflection)$$
 $\left. \partial_s V(x,s,y) \right|_{y=s-} = 0 \qquad \qquad (normal\ reflection)$

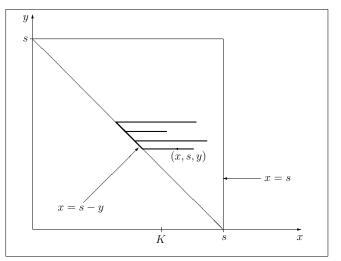


Figure 2. A computer drawing of the state space of the process (X,S,Y), for some s fixed.

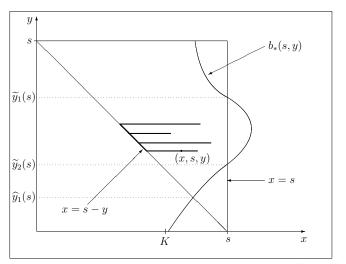


Figure 3. A computer drawing of the state space of the process (X,S,Y), for some s fixed, and the boundary function $b_*(s,y)$.

$$V(x, s, y; b_*(s, y)) = \sum_{i=1}^{2} \frac{(\gamma_{3-i}(s, y) - 1)b_*(s, y) - \gamma_{3-i}(s, y)K}{(\gamma_{3-i}(s, y) - \gamma_i(s, y))b_*(s, y)^{\gamma_i(s, y)}} x^{\gamma_i(s, y)}$$

for $0 < s - y \le x < b_*(s, y) \le s$ and s > K, for

$$\gamma_{i}(s,y) = \frac{1}{2} - \frac{r - \delta(s,y)}{\sigma^{2}(s,y)} - (-1)^{i} \sqrt{\left(\frac{1}{2} - \frac{r - \delta(s,y)}{\sigma^{2}(s,y)}\right)^{2} + \frac{2r}{\sigma^{2}(s,y)}}$$

and 0 < y < s.

$$V(x, s, y; b_*(s, y)) = \sum_{i=1}^{2} \frac{(\gamma_{3-i}(s, y) - 1)b_*(s, y) - \gamma_{3-i}(s, y)K}{(\gamma_{3-i}(s, y) - \gamma_i(s, y))b_*(s, y)^{\gamma_i(s, y)}} x^{\gamma_i(s, y)}$$

for $0 < s - y \le x < b_*(s, y) \le s$ and s > K, for

$$\gamma_{i}(s,y) = \frac{1}{2} - \frac{r - \delta(s,y)}{\sigma^{2}(s,y)} - (-1)^{i} \sqrt{\left(\frac{1}{2} - \frac{r - \delta(s,y)}{\sigma^{2}(s,y)}\right)^{2} + \frac{2r}{\sigma^{2}(s,y)}}$$

and 0 < y < s. The boundary $b_*(s, y)$ solves

$$\partial_{y}b(s,y) = \sum_{i=1}^{2} \frac{((\gamma_{3-i}(s,y)-1)b(s,y)-\gamma_{3-i}(s,y)K)b(s,y)}{(\gamma_{i}(s,y)-1)(\gamma_{3-i}(s,y)-1)b(s,y)-\gamma_{i}(s,y)\gamma_{3-i}(s,y)K} \,\partial_{y}\gamma_{i}(s,y)$$

$$\times \left(\frac{1}{\gamma_{3-i}(s,y) - \gamma_{i}(s,y)} + \frac{((s-y)/b(s,y))^{\gamma_{i}(s,y)} \ln{((s-y)/b(s,y))}}{((s-y)/b(s,y))^{\gamma_{i}(s,y)} - ((s-y)/b(s,y))^{\gamma_{3-i}(s,y)}} \right)$$

$$V(x, s, y; b_*(s, y)) = \sum_{i=1}^{2} \frac{(\gamma_{3-i}(s, y) - 1)b_*(s, y) - \gamma_{3-i}(s, y)K}{(\gamma_{3-i}(s, y) - \gamma_i(s, y))b_*(s, y)^{\gamma_i(s, y)}} x^{\gamma_i(s, y)}$$

for $0 < s - y \le x < b_*(s, y) \le s$ and s > K, for

$$\gamma_i(s,y) = \frac{1}{2} - \frac{r - \delta(s,y)}{\sigma^2(s,y)} - (-1)^i \sqrt{\left(\frac{1}{2} - \frac{r - \delta(s,y)}{\sigma^2(s,y)}\right)^2 + \frac{2r}{\sigma^2(s,y)}}$$

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$$\partial_{y}b(s,y) = \sum_{i=1}^{2} \frac{((\gamma_{3-i}(s,y)-1)b(s,y)-\gamma_{3-i}(s,y)K)b(s,y)}{(\gamma_{i}(s,y)-1)(\gamma_{3-i}(s,y)-1)b(s,y)-\gamma_{i}(s,y)\gamma_{3-i}(s,y)K} \,\partial_{y}\gamma_{i}(s,y) \\
\times \left(\frac{1}{\gamma_{3-i}(s,y)-\gamma_{i}(s,y)} + \frac{((s-y)/b(s,y))^{\gamma_{i}(s,y)} \ln((s-y)/b(s,y))}{((s-y)/b(s,y))^{\gamma_{i}(s,y)} - ((s-y)/b(s,y))^{\gamma_{3-i}(s,y)}}\right)$$

The starting condition is given by

$$b_*(s,s-)=b_*(s)=rac{eta_1(s)K}{eta_1(s)-1}$$
 as $y\uparrow s$.

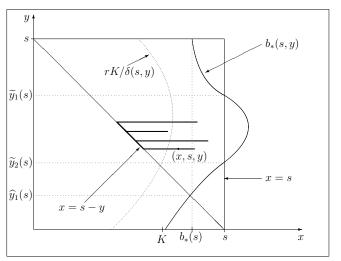


Figure 4. A computer drawing of the state space of the process (X, S, Y), for some s fixed, and the boundary function $b_*(s, y)$.

$$V(x, s, y) = C_1(s, y) x^{\gamma_1(s, y)} + C_2(s, y) x^{\gamma_2(s, y)}$$

for $0 < s - y \le x \le s < b_*(s, y)$, where $C_1(s, y)$ and $C_2(s, y)$ solve

$$\sum_{i=1}^{2} \left(\partial_{s} C_{i}(s,y) s^{\gamma_{i}(s,y)} + C_{i}(s,y) \partial_{s} \gamma_{i}(s,y) s^{\gamma_{i}(s,y)} \ln s \right) = 0$$

$$\sum_{i=1}^{2} \left(\partial_{y} C_{i}(s,y) (s-y)^{\gamma_{i}(s,y)} + C_{i}(s,y) \partial_{y} \gamma_{i}(s,y) (s-y)^{\gamma_{i}(s,y)} \ln(s-y) \right) = 0$$

for each 0 < y < s, with boundary conditions...

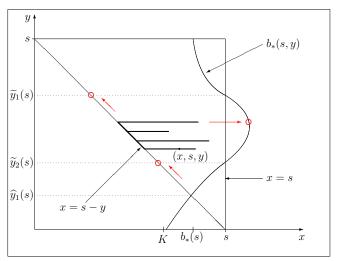


Figure 6. A computer drawing of the state space of the process (X, S, Y), for some s fixed, and the boundary function $b_*(s, y)$.

with boundary conditions...

$$\begin{split} C_{1}(s,\widetilde{y}_{2l}(s)+) & ((s-\widetilde{y}_{2l}(s))-)^{\gamma_{1}(s,\widetilde{y}_{2l}(s)+)} \\ & + C_{2}(s,\widetilde{y}_{2l}(s)+) \left((s-\widetilde{y}_{2l}(s))-)^{\gamma_{2}(s,\widetilde{y}_{2l}(s)+)} \right. \\ & = V(s-\widetilde{y}_{2l}(s),s,\widetilde{y}_{2l}(s);b(s,\widetilde{y}_{2l}(s))) \;, \\ C_{1}(s,\widetilde{y}_{2l-1}(s)) & (s-\widetilde{y}_{2l-1}(s))^{\gamma_{1}(s,\widetilde{y}_{2l-1}(s))} \\ & + C_{2}(s,\widetilde{y}_{2l-1}(s)) & (s-\widetilde{y}_{2l-1}(s))^{\gamma_{2}(s,\widetilde{y}_{2l-1}(s))} \\ & = V((s-\widetilde{y}_{2l-1}(s))-,s,\widetilde{y}_{2l-1}(s)+;b_{*}(s,\widetilde{y}_{2l-1}(s)+)) \end{split}$$

and

$$C_{1}(\overline{s}(y)-,y)(\overline{s}(y)-)^{\gamma_{1}(\overline{s}(y)-,y)}+C_{2}(\overline{s}(y)-,y)(\overline{s}(y)-)^{\gamma_{2}(\overline{s}(y)-,y)}$$

$$=V(\overline{s}(y),\overline{s}(y),y;\ b_{*}(\overline{s}(y),y))\equiv \overline{s}(y)-K$$

Main Result of the Call option in (X,S,Y)-setting

Proposition 2. The value function of the opt. stop. prob. for Call option is

$$V_*(x, s, y) = \begin{cases} V(x, s, y; b_*(s, y)), & \text{if } s - y \le x < b_*(s, y) \le s \\ V(x, s, y), & \text{if } s - y \le x \le s < b_*(s, y) \\ x - K, & \text{if } b_*(s, y) \le x \le s \end{cases}$$

and the optimal stopping time is

$$\tau_* = \inf\{t \geq 0 \,|\, X_t \geq b_*(S_t, Y_t)\}$$

where $b_*(s, y)$ is the minimal solution of an ODE with starting condition $b_*(s, s-) = b_*(s)$, such that it stays strictly above the surface $x = K \vee (rK/\delta(s, y))$.

Thank you!