

# Weak Reflection Principle and Static Hedging of Barrier Options

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# Static vs Dynamic

- In Finance, hedging is a process of offsetting the risks arising from holding a financial instrument by trading (buying and selling) other instruments.
- When the market is complete, the price of a derivative contract can always be replicated by dynamically trading the underlying asset  $S$ .
  - Such **dynamic hedging** strategies have certain drawbacks, in particular, due to the presence of transaction costs.
- Whether the market is complete or not, there sometimes exists a **static portfolio** of *simpler (liquid) derivatives*, such that the value of the portfolio **matches the value of the target (exotic) derivative at any time before the barrier is hit.**

# Barrier Options: Up-and-Out Put

- Consider the problem of
  - **static hedging** of
  - **barrier options**.
- For the sake of transparency, we focus on the **Up-and-Out Put (UOP)** option:
  - An **UOP** written on the underlying process  $S$  is issued with a maturity date  $T > 0$ , a strike price  $K > 0$ , and a flat upper barrier  $U > K$ .
  - At expiry, it **pays off**:

$$I_{\{\sup_{t \in [0, T]} S_t < U\}} \cdot (K - S_T)^+$$

# Definition of Static Hedge

- When it exists, a **static hedging** strategy of a barrier option is characterized by a function

$$G : [0, \infty) \rightarrow \mathbb{R},$$

such that a European option with payoff  $G(S_T)$  at  $T$  **has the same value as the target barrier option** up to and including the time when the barrier is hit.

- One can, then, **hedge the barrier option** by
  - 1 opening a **long position** in a **European option with payoff  $G$**  and
  - 2 trading it at **zero cost** for the corresponding "vanilla" option when/if the underlying hits the barrier.

# Example: Static Hedge in Black's Model

- Consider the **Black's model** where the risk-neutral process for the underlying  $S$  is given by a geometric Brownian motion:

$$dS_t = S_t \sigma dW_t,$$

with  $S_0 < U$  and  $\sigma \in \mathbb{R}$ .

- Carr-Bowie (1994)* show that static hedge of an UOP in such model is given by:

$$G(S) = (K - S)^+ - \frac{K}{U} \left( S - \frac{U^2}{K} \right)^+$$

- Hence an UOP can be replicated exactly by being **long one put** struck at  $K$  and **short  $\frac{K}{U}$  calls** struck at  $\frac{U^2}{K}$ .
- The exact same hedge works in a generalization of the Black model where  $\sigma$  is an **unknown** stochastic process independent of  $W$ .

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# Static Hedge of an UOP in Black's Model

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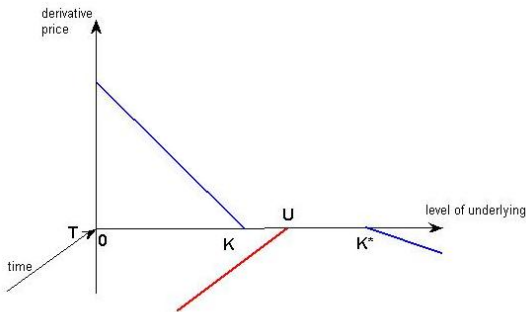


Figure : Static Hedge payoff (blue) and the boundary (red).  $K^* = \frac{U^2}{K}$

# Related results

- **Explicit exact model-dependent static hedge**
  - BS model: *Bowie-Carr* (1994).
  - Symmetric diffusion models: *Carr-Ellis-Gupta* (1998).
  - **Time-homogeneous diffusion models**: *Carr-N.* (2011), *Gesell* (2011).
- **Robust sub/superreplicating** static strategies: *Brown-Hobson-Rogers* (2001), *Cox-Hobson-Obłój* (2008), *Cox-Obłój* (2010), *Galichon-Henry-Labordère-Touzi-Obłój-Spoida* (to appear).
- **Robust static hedging with beliefs**: *N.-Obłój* (in progress). We use the exact model-dependent static hedges as building blocks to construct sub- and super replicating strategies that work in classes of models.
- **Optimization-based approach** to find approximate static hedge: *Sachs, Maruhn, Giese, Sircar, Avelaneda* .



# Exact static hedge in diffusion models

In *Carr-N. (2011)*, we provide **exact static hedging** strategies for barrier options in the following class of models.

- **Pricing of contingent claims is linear**: it is done by taking expectations of discounted payoffs under some pricing measure.
- **Interest rate  $r$  is constant**.
- Under the pricing measure, the underlying  $S$  follows a **time-homogeneous diffusion**:

$$dS_t = \mu(S_t)dt + \sigma(S_t)dB_t$$

- We make some regularity assumptions on  $\mu$  and  $\sigma$ . In particular, our results hold for all models where  $\sigma(S)/S$  is bounded away from zero, and  $\mu(S)/S$  and  $\sigma(S)/S$  have limits at the boundary points, and are bounded along with their first three derivatives.

# Static Hedge of an UOP

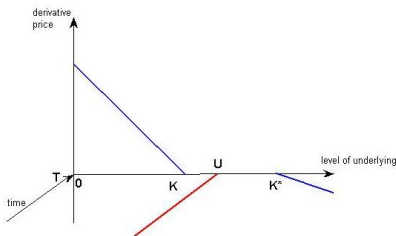


Figure : Static hedge payoff  $G$  (blue) and the barrier (red).

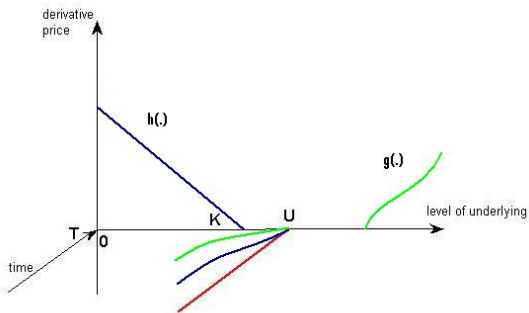
Function  $G$  has to be of the form  $G(S) = (K - S)^+ - g(S)$ , where:

- $g(S) = 0$  for  $S < U$ ,
- and the price of a **European option with payoff  $g$**  is equal to the price of a **put with strike  $K$**  and the same maturity, whenever the underlying hits the barrier.

# "Mirror" Image

Find  $g$ , s.t. it has support in  $(U, \infty)$  and

$$\mathbb{E}[h(S_\tau) | S_0 = U] = \mathbb{E}[g(S_\tau) | S_0 = U], \quad \text{for all } \tau > 0$$



**Figure :** Price functions of the options with payoffs  $h$  (blue) and  $g$  (green), along the barrier  $S = U$  (red)

# Problem formulation

- Consider a stochastic process  $X = (X_t)_{t \geq 0}$ , started from zero:  $X_0 = 0$ .
- Introduce
  - $\Omega_1$  – the set of regular functions (e.g. continuous, with at most exponential growth) with support in  $(-\infty, 0)$ ;
  - $\Omega_2$  – the set of regular functions with support in  $(0, \infty)$ .
- **Problem:** find a mapping  $\mathbf{R} : \Omega_1 \rightarrow \Omega_2$ , such that, for any  $f \in \Omega_1$ :

$$\mathbb{E} [f(X_t)] = \mathbb{E} [\mathbf{R}f(X_t)],$$

for all  $t \geq 0$ .

# Strong Reflection Principle

- If there exists a mapping  $S : \mathbb{R} \rightarrow \mathbb{R}$ , which maps  $(-\infty, 0)$  into  $(0, \infty)$ , and such that the process  $X$  is **symmetric** with respect to this mapping:

$$\text{Law}(S(X_t), t \geq 0) = \text{Law}(X_t, t \geq 0),$$

then, the reflection  $\mathbf{R}$  is easy to construct:

$$\mathbf{R}f = f \circ S$$

- For example, **Brownian motion**  $B$  is symmetric with respect to zero:

$$\text{Law}(-B_t) = \text{Law}(B_t), \quad \forall t \geq 0,$$

and, therefore:

$$\mathbf{R}f(x) = f(-x)$$

- What we call a **Classical (Strong) Reflection Principle** arises as a combination of the **continuity** of  $B$ , its **strong Markov property**, and the above **symmetry**.

# Applications

- The **Strong Reflection Principle** for Brownian motion is used
  - to compute the **joint distribution of Brownian motion** and its running maximum:

$$\mathbb{P}(B_T \leq K, \max_{t \in [0, T]} B_t \leq U)$$

- or, more generally, solve the **Static Hedging problem** when the underlying is a Brownian motion (or any process symmetric with respect to the barrier).
- It turns out that the **Weak Reflection Principle** is enough to solve the above problems.
- We show how to extend this principle to a large class of Markov processes, **which do not possess any strong symmetries!**

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# Weak Reflection Principle for time-homogeneous diffusions

- To simplify the resulting expression, we assume that  $\mu \equiv 0$ .

$$dX_t = \sigma(X_t)dB_t$$

- Then, the reflection mapping  $\mathbf{R}$  is given by

$$\mathbf{R}h(x) = \frac{2}{\pi i} \int_{\varepsilon - \infty i}^{\varepsilon + \infty i} \frac{z\psi_1(x, z)}{\partial_x \psi_1(0, z) - \partial_x \psi_2(0, z)} \int_{-\infty}^0 \frac{\psi_1(s, z)}{\sigma^2(s)} h(s) ds dz,$$

- where the functions  $\psi^1(x, z)$  and  $\psi^2(x, z)$  are the **fundamental solutions** of the associated **Sturm-Liouville** equation:

$$\frac{1}{2}\sigma^2(x)\psi_{xx}(x, z) - z^2\psi(x, z) = 0$$



# Solution to the Static Hedging problem

(N.-Carr, 2011)

- Recall that, in order to solve the static hedging problem, we only need to compute the **mirror image of the put payoff**.

$$h(x) = (K - x)^+$$

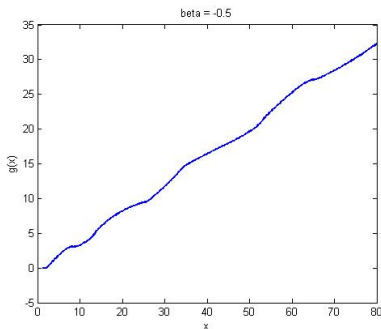
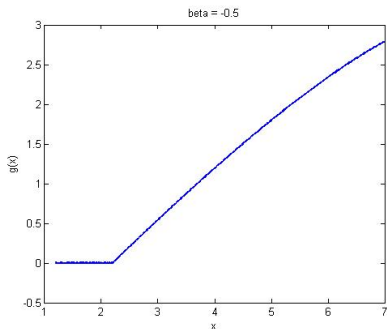
- Thus, the Static Hedge of an UOP option (with barrier  $U$  and strike  $K < U$ ) is given by

$$G(x) = (K - x)^+ - g(x),$$

where

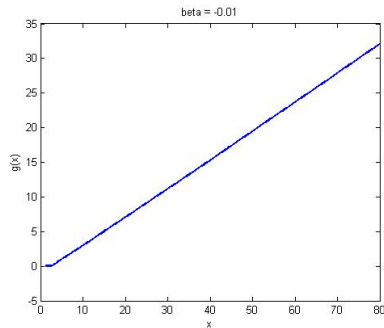
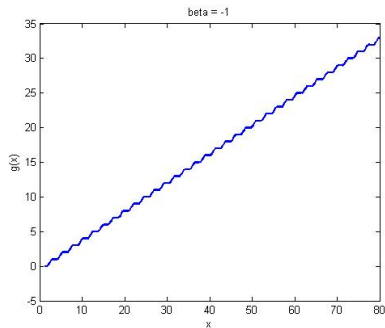
$$g(x) = \frac{1}{\pi i} \int_{\varepsilon - \infty i}^{\varepsilon + \infty i} \frac{\psi^1(x, z) \psi^1(K, z)}{\psi_x^1(U, z) - \psi_x^2(U, z)} \frac{dz}{z},$$

# Constant Elasticity of Variance: $\mu = 0$ , $\sigma(S) = S^{1+\beta}$



**Figure :** The "mirror image"  $g$  in the zero-drift CEV model with barrier  $U = 1.2$  and strike  $K = 0.5$ : the case of  $\beta = -0.5$ , for small (left) and large (right) values of the argument.

# Other CEV: Bachelier and Black-Scholes



**Figure :** The "mirror image"  $g$  in the zero-drift CEV model with barrier  $U = 1.2$  and strike  $K = 0.5$ : the cases of  $\beta = -1$  (left) and  $\beta \approx 0$  (right).

# Computation and extensions

$$g(S) = \frac{1}{\pi i} \int_{\varepsilon - \infty i}^{\varepsilon + \infty i} \frac{\psi^1(S, z) \psi^1(K, z)}{\psi_S^1(U, z) - \psi_S^2(U, z)} \frac{dz}{z},$$

- If  $\sigma(S)$  is **piece-wise constant**, the fundamental solutions  $\psi_1(S, z)$  and  $\psi_2(S, z)$  can be easily computed as **linear combinations of exponentials**, on each sub-interval in  $S$ .
- This family of models is **sufficient for all practical purposes**.
- The proposed static hedge also succeeds in all models that arise by running the time-homogeneous diffusion on an **independent continuous stochastic clock**.
- One can obtain a **semi-robust extension of this static hedging strategy**. More precisely, a strategy that succeeds in all models, as long as the market implied volatility stays within given bounds (beliefs about implied volatility are fulfilled).

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# Robust static hedge with beliefs on implied volatility

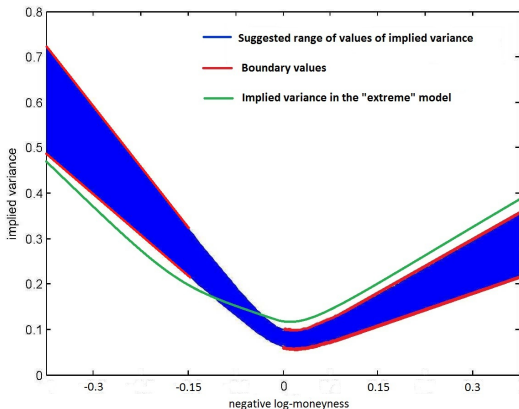


Figure : Range of possible values of (beliefs on) implied volatility (blue), and the extremal implied volatility produced by a diffusion model (green)

# Weak Reflection Principle for Lévy processes with one-sided jumps

- Note that, in principle, this method can be used any **strong Markov process** which **does not jump across the barrier**.
- The only problem is to establish the **Weak Reflection** of this process with respect to the barrier.
- In our ongoing work, we have developed the **Weak Reflection principle for Lévy processes with one-sided jumps**. The solution takes a similar form, in the sense that the reflection operator **R** is given as an integral transform, with a kernel that can be computed through the characteristic exponent of the Lévy process  $\psi$ . For example, the image of  $h(y) = \mathbf{1}_{\{y \leq K\}}$ , for  $K < 0$ , is given by

$$\mathbf{R}h(y) = \frac{1}{2\pi i} \int_{\mathbb{R}^2} \frac{e^{\varepsilon y + i w y + i K z}}{z} \left( \frac{\psi'(w)}{\psi(w) - \psi(-iz)} - \frac{1}{w + iz} \right) dz dw$$

# Beyond Finance

- The Weak Reflection Principle allows to compute the **joint distribution** of a time-homogeneous diffusion and its **running maximum**, through the marginal distribution of the process itself:

$$\mathbb{P}(X_T \leq K, \max_{t \in [0, T]} X_t \leq U) = \mathbb{E}(g^{K, U}(X_T)),$$

where  $g^{K, U}$  is the mirror image of function  $(K - \cdot)^+$ , with respect to the barrier  $U$ .

- The **connection to PDE's** yields various applications in **Physics** and **Biology**.



# Summary

- I have presented a solution to the **Static Hedging** problem for barrier options.
- This solution provides **exact**, but **model-dependent**, hedge in all regular enough time-homogeneous diffusion models.
- In our ongoing work with J. Obloj, we develop the **semi-robust** hedges based on the above results.
- Static Hedging problem motivated the development of a new technique, the **Weak Reflection Principle**.
- We have developed the Weak Reflection principle to diffusion processes and one-sided Lévy processes.

# Summary (cont'd)

- The **Weak Reflection Principle** allows us to
  - control the expected value of a function of the process,
  - at any time when the process is at the barrier of a given domain,
  - by changing the function outside of this domain.
- Applications include **Finance, Physics, Biology, Computational Methods.**
- Further extensions:
  - **Specific applications in Physics and Biology?**
  - **More general domains?**
  - **More general stochastic processes?**

# Non-existence result

- *Bardos-Douady-Fursikov (2004)* treat this problem for a general parabolic PDE, and prove the existence of **approximate solutions**  $g_\varepsilon$ , such that

$$\sup_{t \in [0, T]} \left| u^h(U, t) - u^{g_\varepsilon}(U, t) \right| < \varepsilon$$

- They show that an exact solution doesn't exist in general...
- Their proof is **not constructive** - finding even an approximate solution is left as a separate problem.
- The example of **non-existence** relies heavily on the **time-dependence of the coefficients** in the corresponding PDE!

# Naive numerical approximation

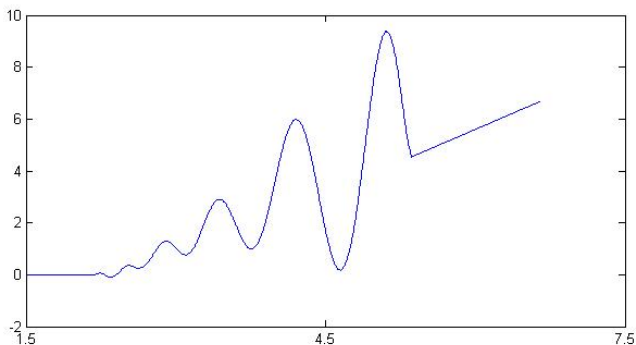


Figure : Payoff function  $g_\varepsilon$  as a result of the naive least-square optimization approach

# Square root process revisited

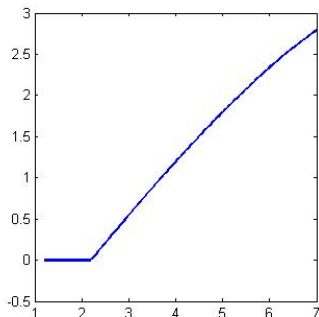


Figure : 2. Function  $g$ , for  $\beta = -0.5$ ,  $U = 1.2$ ,  $K = 0.5$

**Notice that there is a constant  $K^* \geq U$ , such that the support of  $g$  is exactly  $[K^*, \infty]$ .**

# Short-Maturity Behavior and Single-Strike Hedge

- **Key observation:** when time-to-maturity is small, only the values of  $g$  around  $K^*$  matter!
- Thus, for small maturities, the approximation of the payoff function  $g$  with a **scaled call payoff** should perform well.
- We have:

$$\int_K^U \frac{dy}{\sigma(y)} = \int_U^{K^*} \frac{dy}{\sigma(y)}, \quad \eta = \sqrt{\frac{\sigma(K)}{\sigma(K^*)}}.$$

- Using the above, we can construct the **single-strike sub- and superreplicating strategies**: there exists  $\delta > 0$ , such that, whenever  $S_t = U$ ,

$$[1 - \delta(T - t)] P_t(K) - \eta C_t(K^*) \leq 0 \leq [1 + \delta(T - t)] P_t(K) - \eta C_t(K^*)$$

# Function $g$ : properties and numerical computation

- There exists a constant  $K^* \geq U$ , such that the support of  $g$  is exactly  $[K^*, \infty)$ .
- Introduce the "signed geodesic distance":

$$Z(x) := \sqrt{2} \int_U^x \frac{dy}{\sigma(y)}$$

- Then  $K^*$  is a solution of the equation

$$Z(K^*) + Z(K) = 0$$

- The function  $g$  is "analytic with respect to the geodesic distance  $Z$ " in  $(K^*, \infty)$ :

$$g(x) = \sum_{k=1}^{\infty} c_k (Z(x) - Z(K^*))^k,$$

and there exists an algorithm for computing  $c_k$ 's.