

Pricing Game Options with Call Protection: Doubly Reflected Intermittent BSDEs and their Approximation

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Convertible bond with underlying stock S

- Coupons from time 0 onwards
- **Terminal payoff** at $\zeta = \tau \wedge \theta$

$$\mathbf{1}_{\zeta=\tau < T} \ell(\tau, S_\tau) + \mathbf{1}_{\vartheta < \tau} h(\vartheta, S_\vartheta) + \mathbf{1}_{\zeta=T} g(S_T)$$

- $[0, T]$ -valued **bond holder put time** τ and **bond issuer call time** θ
 - Cancelable American claim, or **game option**
- **Call protections** preventing the issuer from calling the bond on certain random time intervals
 - Typically monitored at **discrete monitoring times**
 - In a possibly very **path-dependent** way

Agenda

Mathematical issues

- Doubly reflected backward stochastic differential equations with an **intermittent** upper barrier, only active on random time intervals (RIBSDE)
- Related variational inequality approach (VI)
 - Highly-dimensional pricing problems (path dependence)
 - Deterministic pricing schemes ruled out by the curse of dimensionality

→ **Simulation methods**

Contributions

- A convergence rate for a discrete time approximation scheme by simulation to an RIBSDE
- VI approach
- Practical value of this approach on the benchmark problem of pricing by simulation highly path-dependent convertible bonds
- A demonstration of the real abilities of simulation/regression numerical schemes in high dimension (up to $d = 30$ in this work)

Outline

- 1 Markovian RIBSDE
 - Diffusion Set-Up with Marker Process
 - Markovian RIBSDE
 - Connection with Finance
 - Solution of the RIBSDE
- 2 Approximation Results
 - BSDE Approach
 - Variational Inequality Approach
- 3 Application:Pricing convertible Bonds with call protection
 - Simulation/Regression Approaches
 - Benchmark Model
 - No Call Protection
 - Call Protection
 - Reducible Case
 - General Case

Diffusion Set-Up with Marker Process

- Diffusion with Lipschitz coefficients in \mathbb{R}^q

$$dX_t = b(t, X_t) dt + \sigma(t, X_t) dW_t$$

- Call protection monitoring times $\mathfrak{T} = \{0 = T_0 < \dots < T_N = T\}$
- Marker process H keeping track of the path-dependence, in view of 'markovianizing' the model
- $\mathbb{R}^q \times \mathcal{K}$ -valued factor process $\mathcal{X} = (X, H)$ (finite set \mathcal{K})
 - $u = u(t, x, k) = u^k(t, x)$
- \mathcal{K} -valued pure jump marker process H supposed to be constant except for deterministic jumps at the T_I s

$$H_{T_I} = \kappa_I(X_{T_I}, H_{T_I-})$$

- Jump functions κ_I^k continuous in x outside $\partial\mathcal{O}$ (constant on \mathcal{O} and on ${}^c\mathcal{O}$) for an open, 'regular' domain $\mathcal{O} \subseteq \mathbb{R}^q$

Call Protection

- Subset K of \mathcal{K}
- Call forbidden/possible whenever $H_t \in K / \notin K$
- \mathfrak{T} -valued stopping times given as successive times of exit from and entrance to K , so $\vartheta_0 = 0$ and then

$$\vartheta_{2l+1} = \inf\{t > \vartheta_{2l}; H_t \notin K\} \wedge T, \quad \vartheta_{2l+2} = \inf\{t > \vartheta_{2l+1}; H_t \in K\} \wedge T$$

- Call forbidden/possible on the 'even'/'odd' intervals $[\vartheta_l, \vartheta_{l+1})$
 - $H_t \in K / \notin K$

Starting from $H_0 = k \notin K$ ('Call at the beginning')

$$0 = \vartheta_0 = \vartheta_1 < \vartheta_2 \leq \dots \leq \vartheta_{N+1} = T$$

Call possible on the first non-void time interval $[\vartheta_1 = 0 = \vartheta_0, \vartheta_2 > 0)$

Starting from $H_0 = k \in K$ ('No Call at the beginning')

$$0 = \vartheta_0 < \vartheta_1 \leq \dots \leq \vartheta_{N+1} = T$$

Call forbidden on the first non-void time interval $[\vartheta_0 = 0, \vartheta_1 > 0)$

Markovian RIBSDE

Reflected BSDE (\mathcal{S}) with data

$$f(t, X_t, y, z), \quad \xi = g(X_T), \quad \ell(t, X_t), \quad h(t, X_t), \quad \vartheta$$

- ‘Standard Lipschitz and L^2 -integrability assumptions’ (if not for ϑ)
- Mokobodski condition
 - Existence of a square-integrable quasimartingale Q between L and U
- Doubly reflected BSDE with lower barrier $L_t = \ell(t, X_t)$ and **intermittent** (the ‘I’ in RIBSDE) upper barrier given by, for $t \in [0, T]$

$$U_t = \sum_{l=0}^{[N/2]} \mathbf{1}_{[\vartheta_{2l}, \vartheta_{2l+1})} \infty + \sum_{l=1}^{[(N+1)/2]} \mathbf{1}_{[\vartheta_{2l-1}, \vartheta_{2l})} h(t, X_t)$$

- ‘Nominal’ upper obstacle $h(t, X_t)$ only active on the ‘odd’ random time intervals $[\vartheta_{2l-1}, \vartheta_{2l})$
- Call protection on the ‘even’ random time intervals $[\vartheta_{2l}, \vartheta_{2l+1})$

Risk-neutral pricing problems in finance

Driver coefficient function f typically given as

$$f = f(t, x, y) = c(t, x) - \mu(t, x)y$$

- Dividend and interest-rate related functions c and μ
 - Single-name credit risk (counterparty risk)
 - Recovery-adjusted dividend-yields c
 - Credit-spread adjusted interest-rates μ
 - Pre-default factor process X
- Affine in y , does not depend on z
 - Historical rather than RN modeling \rightarrow 'z-dependent' f
 - Market imperfections \rightarrow nonlinear f

Terminal cost functions typically given by

$$\ell(t, x) = \bar{P} \vee S, \quad h(t, x) = \bar{C} \vee S, \quad g(x) = \bar{N} \vee S$$

$\bar{P} \leq \bar{N} \leq \bar{C}$ Constants

$S = x_1$ first component of x

- Mokobodski condition satisfied with $Q = S$ provided S is a square-integrable Itô $\frac{1}{2}$ process

Highly path dependent call protection

Example (' l out of d ')

Given a constant **trigger level** \bar{S} and constants $l \leq d \leq N$, call possible iff S has been $\geq \bar{S}$ on at least l of the last d monitoring times

- $\mathcal{K} = \{0, 1\}^d$, $\kappa_l^k(x) = (\mathbf{1}_{S \geq \bar{S}}, k_1, \dots, k_{d-1})$
- H_t vector of the indicator functions of the events $S_{T_i} \geq \bar{S}$ at the last d monitoring dates preceding time t

Call possible iff $|H_t| \geq l \Leftrightarrow H_t \notin K$ with $|k| = \sum_{1 \leq p \leq d} k_p$ and $K = \{k \in \mathcal{K}; |k| < l\}$

Solution of the RIBSDE

Definition

A **solution** \mathcal{Y} to (S) is a triple $\mathcal{Y} = (Y, Z, A)$ such that:

(i) $Y \in \mathcal{S}^2, Z \in \mathcal{H}_q^2, A \in \mathcal{A}^2$

(ii) $Y_t = \xi + \int_t^T f(s, X_s, Y_s, Z_s) ds + A_T - A_t - \int_t^T Z_s dW_s \quad t \in [0, T]$

(iii) $L_t \leq Y_t$ on $[0, T]$, $Y_t \leq U_t$ on $[0, T]$

and $\int_0^T (Y_t - L_t) dA_t^+ = \int_0^T (U_t - Y_t) dA_t^- = 0$

(iv) A^+ is continuous, and

$\{(\omega, t); \Delta Y \neq 0\} = \{(\omega, t); \Delta A^- \neq 0\} \subseteq \bigcup_{l=0}^{[N/2]} \llbracket \vartheta_{2l} \rrbracket$

$\Delta Y = \Delta A^-$ on $\bigcup_{l=0}^{[N/2]} \llbracket \vartheta_{2l} \rrbracket$

- $\mathcal{S}^2, \mathcal{H}_q^2$ and \mathcal{A}^2 'usual L^2 spaces'

- A^\pm Jordan component of A

- Convention that $0 \times \pm\infty = 0$ in (iii)

- Obvious extension to a random terminal time θ

- For l decreasing from N to 0, let us define $\mathcal{Y}^l = (Y^l, Z^l, A^l)$ on $[\vartheta_l, \vartheta_{l+1}]$ as the solution, with A^l continuous, to the stopped RBSDE (for l even) or R2BSDE (for l odd) with data (with $Y_{\vartheta_{N+1}}^{N+1} \equiv g(X_T)$)

$$\begin{cases} f(t, X_t, y, z), Y_{\vartheta_{l+1}}^{l+1}, \ell(t, X_t) & (l \text{ even}) \\ f(t, X_t, y, z), \min(Y_{\vartheta_{l+1}}^{l+1}, h(\vartheta_{l+1}, X_{\vartheta_{l+1}})), \ell(t, X_{st}), h(t, X_t) & (l \text{ odd}) \end{cases}$$

- Let us define $\mathcal{Y} = (Y, Z, A)$ on $[0, T]$ by, for every $l = 0, \dots, N$:
 - $(Y, Z) = (Y^l, Z^l)$ on $[\vartheta_l, \vartheta_{l+1})$, and also at $\vartheta_{N+1} = T$ in case $l = N$. So in particular

$$Y_0 = \begin{cases} Y_0^0, & k \in K \\ Y_0^1, & k \notin K \end{cases}$$

where k is the initial condition of the marker process H .

- $dA = dA^l$ on $(\vartheta_l, \vartheta_{l+1})$,

$$\Delta A_{\vartheta_l} = \Delta A_{\vartheta_l}^- = (Y_{\vartheta_l}^l - h(\vartheta_l, X_{\vartheta_l}))^+ = \Delta Y_{\vartheta_l} (= 0 \text{ for } l \text{ odd})$$

and $\Delta A_T = \Delta Y_T = 0$.

Proposition

$\mathcal{Y} = (Y, Z, A)$ is the unique solution to (S)

Verification principle

Risk-neutral pricing problems in finance

Financial interpretation of a solution \mathcal{V} to (\mathcal{S})

Y_0 'NFLVR' Arbitrage price at time 0 for the game option with payoff functions c , l , h , g and call protection ϑ

Bilateral super-hedging price and infimal issuer super-hedging price

up to a local martingale cost process

Z Hedging strategy

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Approximation of the Forward Process

- Time-grid $t = \{0 = t_0 < t_1 < \dots < t_n = T\} \supseteq \mathfrak{T}$
- Euler scheme approximation of \hat{X}

$$\hat{X}_{t_{i+1}} = \hat{X}_{t_i} + b(t_i, \hat{X}_{t_i})(t_{i+1} - t_i) + \sigma(t_i, \hat{X}_{t_i})(W_{t_{i+1}} - W_{t_i})$$

- Approximation of the marker process H

$$\hat{H}_{T_i} = \kappa_l(\hat{X}_{T_i}, \hat{H}_{T_i-})$$

Approximation of the Call Protection Switching Times

Approximation $\hat{\vartheta}$ of ϑ obtained by using $\hat{\mathcal{X}} = (\hat{X}, \hat{H})$ instead of \mathcal{X} in the definition of ϑ

Proposition (Assuming σ non-degenerate and 'some regularity of σ and b around $\partial\mathcal{O}$)

For every $l \leq N + 1$

$$\mathbb{E}[|\vartheta_l - \hat{\vartheta}_l|] \leq C_\varepsilon |\mathfrak{t}|^{\frac{1}{2} - \varepsilon}$$

$$|\mathfrak{t}| = \max_{i \leq n-1} (t_{i+1} - t_i)$$

Approximation of the RIBSDE

- Projection operator $\widehat{\mathcal{P}}$ defined by

$$\widehat{\mathcal{P}}(t, x, y) = y + [\ell(t, x) - y]^+ - [y - h(t, x)]^+ \sum_{l=1}^{[(N+1)/2]} \mathbf{1}_{\{\widehat{v}_{2l-1} \leq t \leq \widehat{v}_{2l}\}}$$

- Reflection operating only on a subset \mathfrak{r} of \mathfrak{t} in the approximation scheme for \mathcal{Y}

$$\mathfrak{r} = \{0 = r_0 < r_1 < \dots < r_\nu = T\} \text{ with } \mathfrak{T} \subseteq \mathfrak{r} \subseteq \mathfrak{t}$$

Components Y and Z of a solution $\mathcal{Y} = (Y, Z, A)$ to (\mathcal{S}) approximated by a triplet of processes $(\hat{Y}, \tilde{Y}, \bar{Z})$ defined on t

Terminal condition

$$\hat{Y}_T = \tilde{Y}_T = g(\hat{X}_T)$$

and then for i decreasing from $n - 1$ to 0

$$\begin{cases} \bar{Z}_{t_i} &= \mathbb{E} \left[\hat{Y}_{t_{i+1}} \left(\frac{W_{t_{i+1}} - W_{t_i}}{t_{i+1} - t_i} \right) \mid \mathcal{F}_{t_i} \right] \\ \tilde{Y}_{t_i} &= \mathbb{E} \left[\hat{Y}_{t_{i+1}} \mid \mathcal{F}_{t_i} \right] + (t_{i+1} - t_i) f(t_i, \hat{X}_{t_i}, \tilde{Y}_{t_i}, \bar{Z}_{t_i}) \\ \hat{Y}_{t_i} &= \tilde{Y}_{t_i} \mathbf{1}_{\{t_i \notin \tau\}} + \hat{\mathcal{P}}(t_i, \hat{X}_{t_i}, \tilde{Y}_{t_i}) \mathbf{1}_{\{t_i \in \tau\}} \end{cases}$$

Continuous-time extension of the scheme still denoted by $(\hat{Y}, \tilde{Y}, \bar{Z})$
 \hat{Z} Integrand in a stochastic integral representation of \hat{Y}

Theorem (No call or no call protection, Chassagneux 08)

In case of Lipschitz barriers and for $|\tau| \sim |t|^{\frac{2}{3}}$ (resp. semi-convex barriers and for $|\tau| \sim |t|^{\frac{1}{2}}$), one has

$$\max_{i \leq n-1} \sup_{t \in [t_i, t_{i+1})} \mathbb{E}[|Y_t - \tilde{Y}_{t_i}|^2] + \max_{i \leq n-1} \sup_{t \in [t_i, t_{i+1})} \mathbb{E}[|Y_{t-} - \hat{Y}_{t_i}|^2] \leq C|t|^\alpha$$

with $\alpha = \frac{1}{3}$ (resp. $\frac{1}{2}$).

Theorem (Call protection, this work, assuming f does not depend on z)

In case of Lipschitz barriers and for $|\tau| \sim |t|^{\frac{1}{2}}$ (resp. semi-convex barriers and for $|\tau| \sim |t|$), one has

$$\max_{i \leq n-1} \sup_{t \in [t_i, t_{i+1})} \mathbb{E}[|Y_t - \tilde{Y}_{t_i}|^2] + \max_{i \leq n-1} \sup_{t \in [t_i, t_{i+1})} \mathbb{E}[|Y_{t-} - \hat{Y}_{t_i}|^2] \leq C_\varepsilon |t|^{\alpha-\varepsilon}$$

with $\alpha = \frac{1}{4}$ (resp. $\frac{1}{2}$).

- Proof of the theorem based on a suitable concept of time-continuous discretely reflected BSDEs
 - Bermudan options
- Possible extension to the case where f depends on z
- Representations of \tilde{Y} and \hat{Z} using approximated optimal policies
 - Cf. 'MC Backward versus Forward' in the numerical part

Variational Inequality Approach

- Comparing the simulation results them with those of an alternative, deterministic numerical scheme
- Deterministic scheme for (S) based on an analytic characterization of (S)
- Let $\mathcal{E} = [0, T] \times \mathbb{R}^q \times \mathcal{K}$ and for $l = 1, \dots, N$

$$\mathcal{E}_l = [T_{l-1}, T_l] \times \mathbb{R}^q \times \mathcal{K}, \mathcal{E}_l^* = [T_{l-1}, T_l) \times \mathbb{R}^q \times \mathcal{K}$$

- The \mathcal{E}_l^* s and $\{T\} \times \mathbb{R}^q \times \mathcal{K}$ partition \mathcal{E}

Continuity of ϑ with respect to (t, x, i)

- Continuous outside $\mathfrak{T} \times \mathbb{R}^q \times \mathcal{K}$
- Cadlag on $(\mathfrak{T} \times \mathbb{R}^q \times \mathcal{K}) \setminus (\mathfrak{T} \times \partial\mathcal{O} \times \mathcal{K})$
- Cad but not 'lag' on $\mathfrak{T} \times \partial\mathcal{O} \times \mathcal{K}$

Cauchy cascade

Definition

(i) Cauchy cascade (g, ν) on \mathcal{E}

- Terminal condition g at T
- Sequence $\nu = (u_l)_{1 \leq l \leq N}$ of functions u_l s on the \mathcal{E}_l s
- Jump condition for $x \notin \partial \mathcal{O}$ (with $u_{N+1} \equiv g$):

$$u_l^k(T_l, x) = \begin{cases} \min(u_{l+1}(T_l, x, \kappa_l^k(x)), h(T_l, x)) & \text{if } k \notin K \text{ and } \kappa_l^k(x) \in K \\ u_{l+1}(T_l, x, \kappa_l^k(x)) & \text{else} \end{cases}$$

(ii) Continuous Cauchy cascade

- Cauchy cascade with continuous ingredients g at T and u_l s on the \mathcal{E}_l s, except maybe for **discontinuities** of the u_l^k s on $\mathcal{T} \times \partial \mathcal{O}$

(iii) Function on \mathcal{E} defined by a Cauchy cascade

- Concatenation on the \mathcal{E}_l^* s of the u_l s + terminal condition g at T

Cascade Characterization of \mathcal{Y}

Proposition

$Y_t = u(t, \mathcal{X}_t)$, $t \in [0, T]$, for a deterministic *pricing function* u , defined by a continuous Cauchy cascade $(g, \nu = (u_I)_{1 \leq I \leq N})$ on \mathcal{E}

Analytic characterization of u ?

Generator of X

$$\mathcal{G}\phi(t, x) = \partial_t \phi(t, x) + \partial \phi(t, x) b(t, x) + \frac{1}{2} \text{Tr}[a(t, x) \mathcal{H}\phi(t, x)]$$

$$a(t, x) \sigma(t, x) \sigma(t, x)^T$$

$\partial \phi$, $\mathcal{H}\phi$ Row-gradient and Hessian of ϕ with respect to x

Cauchy cascade (\mathcal{VI})

For l decreasing from N to 1,

- At $t = T_l$ for every $k \in \mathcal{K}$ and $x \in \mathbb{R}^q$

$$u_l^k(T_l, x) = \begin{cases} \min(u_{l+1}(T_l, x, \kappa_l^k(x)), h(T_l, x)), & k \notin K \text{ and } \kappa_l^k(x) \in K \\ u_{l+1}(T_l, x, \kappa_l^k(x)), & \text{else} \end{cases}$$

with $u_{N+1} \equiv g$

- On the time interval $[T_{l-1}, T_l)$ for every $k \in \mathcal{K}$,

$$\begin{cases} \min(-\mathcal{G}u_l^k - f^{u_l^k}, u_l^k - \ell) = 0, & k \in K \\ \max(\min(-\mathcal{G}u_l^k - f^{u_l^k}, u_l^k - \ell), u_l^k - h) = 0, & k \notin K \end{cases}$$

with for any function $\phi = \phi(t, x)$

$$f^\phi = f^\phi(t, x) = f(t, x, \phi(t, x))$$

- Technical difficulty due to the potential **discontinuity** in x of the functions u_t^k s on $\partial\mathcal{O}$
 - Characterizing ν in terms of a suitable notion of **discontinuous viscosity solution** of (\mathcal{VI}) ?
 - **Convergence results?** for deterministic approximation schemes to u
- **Curse of dimensionality**
 - $(\mathcal{VI}) = \text{Card}(\mathcal{K})$ equations in the u^k s
 - $\sim (q + d)$ – dimensional pricing problem with $d = \log(\text{Card}(\mathcal{K}))$
 - **Simulation schemes** the only viable alternative for d greater than few units

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Regressions

regression approaches for computing by **regression functions** (conditional expectations)

$$x \mapsto \rho(x) = \mathbb{E}(\xi | X = x)$$

ξ, X Real- and \mathbb{R}^q -valued square integrable random variables

Pairs $(X^j, \xi^j)^{1 \leq j \leq m}$ simulated independently according to the law of $(X, \xi) \rightarrow$ Estimate the conditional expectation $\mathbb{E}(\xi | X)$

Parametric regression of the ξ^j s against the $(\varphi^l(X^j))_{1 \leq l \leq p}^{1 \leq j \leq m}$, where (φ^l) is a well chosen 'basis' of functions from \mathbb{R}^q to \mathbb{R}

Regression basis

- **parametric** vs **non-parametric**

Parametric versus Non parametric estimation

parametric regression

$m(x^1, \dots, x^d) = a_0 + \sum_{i=1}^d a_i x^{(i)}$ and the coefficients a_0, \dots, a_d are estimated by LSM

Nonparametric regression

$$Y_i = m(X_i) + \varepsilon$$

Estimation of $m(x)$ by the average of those Y_i where X_i is close to x

$$m_n(x) = \sum_{i=1}^n W_{n,i}(x) \cdot Y_i$$

$$m_n(x) = \frac{\sum_{i=1}^n I_{x_i \in A_{n,j}} Y_i}{\sum_{i=1}^n I_{x_i \in A_{n,j}}}$$

Conclusion

For multivariate X , it is not clear how to choose a proper form of a parametric estimate, and a wrong form will lead to a bad estimate. This inflexibility concerning the structure of the regression function is avoided by so called non parametric regression estimate

Convertible bond with underlying stock S

- Coupons from time 0 onwards
- **Terminal payoff** at $\zeta = \tau \wedge \vartheta$

$$\mathbf{1}_{\zeta=\tau < T} \ell(\tau, S_\tau) + \mathbf{1}_{\vartheta < \tau} h(\vartheta, S_\vartheta) + \mathbf{1}_{\zeta=T} g(S_T)$$

- $\ell(t, S_t) = \bar{P} \vee S_t$; $h(t, S_t) = \bar{C} \vee S_t$; $g(S_T) = \bar{N} \vee S_T$
- $[0, T]$ -valued **bond holder put time** τ and **bond issuer call time** ϑ
- Cancelable American claim, or **game option**
- **Call protections** preventing the issuer from calling the bond on certain random time intervals
 - Typically monitored at **discrete monitoring times**
 - In a possibly very **path-dependent** way

Benchmark Model

Local drift and volatility pre-default model for a stock $X = S$

$$\frac{dS_t}{S_t} = b(t, S_t)dt + \sigma(t, S_t)dW_t$$

$$b(t, S) = r(t) - q(t) + \eta\gamma(t, S), \quad \gamma(t, S) = \gamma_0(S_0/S)^\alpha, \quad \sigma(t, S) = \sigma$$

$r(t)$ Riskless short interest rate

$q(t)$ Dividend yield

$\gamma(t, S)$ Local default intensity ($\gamma_0, \alpha \geq 0$)

$0 \leq \eta \leq 100\%$ Loss Given Default of the firm issuing the bond

Coupon rate function

$$c(t, S) = \bar{c}(t) + \gamma(t, S) ((1 - \eta)S \vee \bar{R})$$

\bar{c} Nominal coupon rate function

\bar{R} Nominal recovery on the bond upon default

Discounting

$\mu(t, S) = r(t) + \gamma(t, S)$ Credit-risk adjusted interest rate

$\beta_t = e^{-\int_0^t \mu(s, S_s) ds}$ Risk-neutral credit-risk adjusted discount factor

General Conditions for the Numerical Experiments

General Data

\bar{P}	\bar{N}	\bar{C}	η	σ	r	q	γ_0	α	m
0	100	103	1	0.2	0.05	0	0.02	1.2	10^4

m number of Monte Carlo trajectories

Time-step $t_{i+1} - t_i = h$

six hours (four time steps per day) in the case of simulation methods
one day in the case of deterministic schemes

Space-steps in the S variable

$S^{j+1} - S^j = 0.5$ in the case of the (fully implicit) deterministic schemes
Cells of diameter one (segments of length one) in the case of
simulation/regression methods involving a method of cells
in S

Monte Carlo different methods(American Option)

Monte Carlo Backward Method

$$\begin{cases} Q_N := \varphi(S_{t_N}) \\ Q_{j-1} := \max(\varphi(S_{t_{j-1}}), \mathbb{E}(B(t_{j-1}, t_j) Q_j | \mathcal{F}_{t_{j-1}})), \quad 1 \leq j \leq N. \end{cases}$$

Monte Carlo Forward Method

$$\begin{cases} Q_0 = \mathbb{E}(B(0, \tau^*) \varphi(S_{\tau^*})) \text{ with} \\ \tau^* := \min \{t_j; \varphi(S_{t_j}) = Q_j\} \end{cases}$$

No Call Protection

Standard Game Option

$$\vartheta_1 = 0, \vartheta_2 = T$$

call possible on $[\vartheta_1, T]$, forbidden before

Simulated mesh $(S_i^j)_{0 \leq i \leq n}^{1 \leq j \leq m} \rightarrow$ Estimate $(u_i^j) = u(t_i, S_i^j)_{0 \leq i \leq n}^{1 \leq j \leq m}$

$u_n = g$, then for $i = n-1 \dots 0$, for $j = 1 \dots m$,

$$u_i^j = \min \left(h_i(S_i^j), \max \left(\ell_i(S_i^j), e^{-\mu_i^j h} \mathbb{E}_i^j(u_{i+1} + h c_{i+1}) \right) \right)$$

$\mathbb{E}_i^j(u_{i+1} + h c_{i+1})$ Conditional expectation given $t = t_i, S_i = S_i^j$

- Computed by **parametric regression** of $(u_{i+1} + h c_{i+1})_{1 \leq j \leq m}$ against $(S_i)_{1 \leq j \leq m}$, using a global parametric regression basis $1, S, S^2$ in S

Regression estimate of the **delta**

$$\delta_i^j = \frac{\mathbb{E}_i^j \{u_{i+1}(S_{i+1} - S_i)\}}{(\sigma_i S_i^j)^2 h}$$

Alternative **MC forward** estimates of price and delta at time 0

Backward vs Forward MC

Maturity $T = 125$ days, Nominal coupon rate $\bar{c} = 0$

MC Fd less volatile than MC Bd (Deviations over 50 trials, $S_0 = 100.55$)

	Value VI	Dev MC Bd	Dev MC Fd
Price	102.049	0.821	0.010
Delta	0.416	0.071	0.019

MC Fd more accurate than MC Bd (%Err=1 \leftrightarrow relative difference of 1% between MC and VI)

S_0	VI Price	%Err Bd	%Err Fd	VI delta	%Err Bd	%Err Fd
98.55	101.246	1.90	0.04	0.376	1.07	0.07
99.55	101.637	1.92	0.01	0.396	0.95	0.50
100.55	102.049	1.99	0.01	0.416	2.77	0.67
101.55	102.479	1.65	0.07	0.435	3.97	3.47

Call Protection

Non-decreasing sequence of $[0, T]$ -valued stopping times
Effective call payoff process

$$U_t = \Omega_t^c \infty + \Omega_t h(t, X_t) = U(t, S_t, H_t)$$

$$\Omega_t = \mathbf{1}_{\{H_t \notin K\}}$$

Simulated mesh $(S_i^j, H_i^j)_{0 \leq i \leq n}^{1 \leq j \leq m} \rightarrow$ Estimate $(u_i^j) = u(t_i, S_i^j, H_i^j)_{0 \leq i \leq n}^{1 \leq j \leq m}$

$u_n = g$, then for $i = n - 1 \dots 0$, for $j = 1 \dots m$

$$u_i^j = \min \left(U_i \left(S_i^j, H_i^j \right), \max \left(\ell_i \left(S_i^j \right), e^{-rh} \mathbb{E}_i^j \left(u_{i+1} + hc_{i+1} \right) \right) \right)$$

\min plays no role outside the support of U_i , where $U_i(S, H)$ is equal to $+\infty$

Highly path dependent call protection

Example (' l out of d ')

Given a constant **trigger level** \bar{S} and constants $l \leq d \leq N$, call possible iff S has been $\geq \bar{S}$ on at least l of the last d monitoring times

- $\mathcal{K} = \{0, 1\}^d$, $\kappa_l^k(x) = (\mathbf{1}_{S \geq \bar{S}}, k_1, \dots, k_{d-1})$
- H_t vector of the indicator functions of the events $S_{T_i} \geq \bar{S}$ at the last d monitoring dates preceding time t

Call possible iff $|H_t| \geq l \Leftrightarrow H_t \notin K$ with $|k| = \sum_{1 \leq p \leq d} k_p$ and $K = \{k \in \mathcal{K}; |k| < l\}$

- $\mathbb{E}_i^j(u_{i+1} + hc_{i+1})$ Conditional expectation given $t = t_i, S_i = S_i^j, H_i = H_i^j$
- computed by **non-linear regression** of $(u_{i+1} + hc_{i+1})_{1 \leq j \leq m}$ against $(S_i, H_i)_{1 \leq j \leq m}$, using for example a **method of cells** in (S, H)

Numerical Data

'I out of d' with $\bar{S} = 103$

Maturity $T = 180$ days, Nominal coupon rate $\bar{c} = 1.2/\text{month}$

Other data unchanged

Reducible Case

In case $l = d$ one can reduce the problem to **two space dimensions** instead of $d + 1$

- S and the **number N of consecutive monitoring dates** T_i s with $S_{T_i} \geq \bar{S}$ from time t backwards (capped at l)

Two simulation schemes

MC_d a method of cells in (S, H)

MC¹ a method of cells in (S, N)

MC_d more accurate than MC¹ ($S_0 = 100$), Intermittent Path

l	1	5	10	20	30
VI ¹ price	103.91	105.10	106.03	107.22	108.01
MC ¹ %Err	0.04	0.16	0.47	0.88	1.34
MC _d %Err	0.04	0.15	0.03	0.04	0.24

General Case

Computation Times

d	1	5	10	20	30
VI_d	332s	5332s	44h	—	—
MC_d	154s	212s	313s	474s	628s
Rel Err	range 1 bp—1%			—	—

Will use two methods for the computation of the conditional expectations in MC_d :

MC_d a method of cells in (S, H) ,

$MC_d^\#$ a method of cells in $(S, |H|^\#)$

Approximate $MC_d^\#$ Algorithm

$|H|^\#$ number of ones in H starting from the $(l - |H|)^{th}$ zero

- $|H|^\# = N$ in case $l = d$

Example ($d = 10, l = 8$)

- $H = (1, 1, 1, 1, \mathbf{0}, 1, 1, 1, 0, 0)$
 $l - |H| = 8 - 7 = 1$, $|H|^\# = 3$ (number of ones on the right of the first zero, in bold in H),
- $H = (1, 1, 1, 0, 1, 1, 1, \mathbf{0}, 0, 0)$ $l - |H| = 8 - 6 = 2$, $|H|^\# = 0$ (number of ones on the right of the second zero, in bold in H)

Rationale Entries of H preceding its $(l - |H|)^{th}$ zero irrelevant to the price

- Necessarily superseded by new ones before the bond may become callable
- Approximate algorithm \sim reducible case based on the 'good regressor' $|H|^\#$ for estimating highly path-dependent conditional expectations

MC_d good , $MC_d^\#$ 'rather good' ($d = 5, S_0 = 100$), HP Intermittent

I	2	3	5
VI_d price	104.07	104.43	105.10
MC_d %Err	0.21	0.15	0.15
$MC_d^\#$ %Err	0.19	0.23	0.18

MC_d good , $MC_d^\#$ 'OK' ($d = 10, S_0 = 100$), HP Intermittent

I	2	5	10
VI_d price	104.27	104.87	106.03
MC_d %Err	0.01	0.15	0.03
$MC_d^\#$ %Err	0.04	0.26	0.38

Deviations over 50 trials and relative difference
($d = 30, S_0 = 102.55$), HP Intermittent

	1	5	10	20	30
Dev MC_d		0.056	0.061	0.086	0.152
Dev MC_d^\sharp		0.060	0.069	0.092	0.175
% Err		0.09	0.24	0.72	1.06

'Good regressor' algorithm MC_d^\sharp rather accurate in practice
Ability to work with a 'good' (as opposed to exact), low-dimensional regressor

- An interesting feature of simulation as opposed to deterministic numerical schemes