

A Monte Carlo method for fully nonlinear parabolic PDEs with applications to financial mathematics

Arash Fahim,
Ecole polytechnique, Paris and Sharif University of technology,
tehran.

Joint work with Nizar Touzi and Xavier Warin

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- ▶ Price: $\frac{dS_t}{S_t} = \mu dt + \sigma dW_t$.
- ▶ Bank account with interest rate = r .
- ▶ Choose an admissible portfolio strategy. Self-financing portfolio: $dX^{x,\theta} = \theta_t dS_t + (X^{x,\theta} - S_t \theta_t) r dt$.
- ▶ Change of numéraire: $dX^{x,\theta} = \theta_t ((\mu - r) dt + \sigma dW_t)$.
- ▶ To maximize $\mathbb{E} [U(X_T^\theta) | X_0 = x]$ over all admissible portfolio.
- ▶ DPP approach:

$$v(t, x) := \sup_{\theta} \mathbb{E} [U(X_T^\theta) | X_t = x].$$

- ▶ H-J-B equation:

$$-\frac{\partial v}{\partial t} - \sup_{\theta} \left\{ \theta(\mu - r) \frac{\partial v}{\partial x} + \frac{1}{2} \theta^2 \sigma^2 \frac{\partial^2 v}{\partial x^2} \right\} = 0$$

$$v(T, \cdot) = U(\cdot).$$

- ▶ The equation is parabolic and fully nonlinear.

$$-\frac{\partial v}{\partial t} + \frac{(\mu - r)^2 \left(\frac{\partial v}{\partial s}\right)^2}{2\sigma^2 \frac{\partial^2 v}{\partial s^2}} = 0$$

$$v(T, x) = U(x).$$

- ▶ Jump in price: $\frac{dS_t}{S_{t-}} = \mu dt + \sigma dW_t + \int_{\mathbb{R}^*} \eta(z) \tilde{J}(dt, dz)$.
- ▶ Fully nonlinear nonlocal parabolic PDE:

$$-\frac{\partial v}{\partial t} - \sup_{\theta} \left\{ \theta(\mu - r) \frac{\partial v}{\partial x} + \frac{1}{2} \theta^2 \sigma^2 \frac{\partial^2 v}{\partial x^2} + \int_{\{|z| \geq 1\}} (v(t, x + \theta \eta(z)) - v(t, x)) d\nu(z) \right\} = 0$$

$$v(T, \cdot) = U(\cdot)$$

- ▶ Non financial motivation: coupled FBSDE (Markovian case).

$$X_t = x + \int_0^t b(s, X_s, Y_s) ds + \int_0^t \sigma(s, X_s, Y_s) dW_s$$

$$Y_t = g(X_T) + \int_t^T f(s, X_s, Y_s, Z_s) ds - \int_t^T Z_s dW_s$$

- ▶ Simulation:
 - 4 step scheme ([Ma, Protter, Yong, 1994]).
 - Picard type iteration ([Zhang, Bender] or [Menozzi, Delarue]).

4 step scheme

- ▶ Quasi-linear PDE:

$$-\frac{\partial v}{\partial t} - \text{tr}[\sigma\sigma^T(t, x, v)D^2v] + Dv \cdot b(t, x, v) + f(t, x, v, Dv\sigma(t, x, v)) = 0$$
$$v(T, \cdot) = g(\cdot).$$

- ▶ $Y_t = v(t, X_t)$ and $Z_t = Dv(t, X_t)\sigma(t, x, Y_t)$.
- ▶ A Monte Carlo method could be used in 4 step scheme.
- ▶ Despite iterative methods, 4 step scheme is **not** practically implementable in great dimensions.

Fully nonlinear parabolic PDE

$$\begin{aligned}
 -\mathcal{L}^X v(t, x) - F(t, x, v(t, x), Dv(t, x), D^2v(t, x)) &= 0, \text{ on } [0, T) \times \mathbb{R}^d, \\
 v(T, \cdot) &= g,
 \end{aligned}$$

where

$$\mathcal{L}^X \varphi(t, x) := \left(\frac{\partial \varphi}{\partial t} + \mu \cdot D\varphi + \frac{1}{2} a \cdot D^2 \varphi \right)(t, x)$$

Separation of linear and non linear part is **to some extent** arbitrary.

Scheme idea

Let X be the diffusion corresponding to \mathcal{L}^X . If the solution v is smooth enough by Itô lemma:

$$\begin{aligned}\mathbb{E}_{t_i, x} [v(t_{i+1}, X_{t_{i+1}})] &= v(t_i, x) + \mathbb{E}_{t_i, x} \left[\int_{t_i}^{t_{i+1}} \mathcal{L}^X v(t, X_t) dt \right] \\ &= v(t_i, x) - \mathbb{E}_{t_i, x} \left[\int_{t_i}^{t_{i+1}} F(\cdot, v, Dv, D^2v)(t, X_t) dt \right]\end{aligned}$$

$$\begin{aligned}v(t_i, x) &\approx \mathbb{E}_{t_i, x} [v(t_{i+1}, X_{t_{i+1}})] \\ &\quad + hF(\cdot, \mathbb{E}_{t_i, x} v(t_{i+1}, X_{t_{i+1}}), \mathbb{E}_{t_i, x} Dv(t_{i+1}, X_{t_{i+1}}), \mathbb{E}_{t_i, x} D^2v(t_{i+1}, X_{t_{i+1}})).\end{aligned}$$

The choice of linear operator = the choice of diffusion process

Scheme

\hat{X} is Euler discretization of X .

$$v^h(T, \cdot) := g \quad \text{and} \quad v^h(t_i, x) := \mathbf{T}_h[v^h](t_i, x),$$

where

$$\begin{aligned} \mathbf{T}_h[v^h](t_i, x) &:= \mathbb{E}_{t_i, x} \left[v(t_{i+1}, \hat{X}_{t_{i+1}}) \right] \\ &+ hF \left(\cdot, \mathbb{E}_{t_i, x} v(t_{i+1}, \hat{X}_{t_{i+1}}), \mathbb{E}_{t_i, x} Dv(t_{i+1}, \hat{X}_{t_{i+1}}), \mathbb{E}_{t_i, x} D^2 v(t_{i+1}, \hat{X}_{t_{i+1}}) \right). \end{aligned}$$

Integration by part

Let

$$H_h^0 = 1, \quad H_h^1 = (\sigma^T)^{-1} \frac{W_h}{h}, \quad H_h^2 = (\sigma^T)^{-1} \frac{W_h W_h^T - h \mathbf{I}_d}{h^2} \sigma^{-1}.$$

For every function $\varphi : Q_T \rightarrow \mathbb{R}$ with exponential growth, we have:

$$\mathbb{E}[D^{(k)}\varphi(t_{i+1}, \hat{X}_h^{t_i, x})] = \mathbb{E}[\varphi(t_{i+1}, \hat{X}_h^{t_i, x}) H_h^k(t_i, x)],$$

- This result gives an Monte Carlo approximation for the derivatives.

2BSDE I

- ▶ If the solution v is smooth enough by applying Itô lemma to $Y_t := v(t, X_t)$ and $Z_t := Dv(t, X_t)$:

$$Y_t = g(X_T) + \int_t^T F(s, X_s, Y_s, Z_s, \Gamma_s) ds - \int_t^T Z_s dX_s$$

$$dZ_t = A_t dt + \Gamma_t dW_t$$

- ▶ $(Y_t, Z_t, \Gamma_t, A_t) = (v(t, X_t), Dv(t, X_t), D^2v(t, X_t), \mathcal{L}Dv(t, X_t))$.
- ▶ No existence result. Uniqueness due to [Cheridito-Soner-Touzi-Victoir].

2BSDE II

- ▶ $t_i = ih$. Discretization gives:

$$Z_i = \frac{1}{h} \mathbb{E}_i [Y_{i+1} \Delta W_{i+1}], \quad \Gamma_i = \frac{1}{h} \mathbb{E}_i [Z_{i+1} \Delta W_{i+1}]$$

$$Y_i = \mathbb{E}_i [Y_{i+1} h + F(t_i, X_i, Y_i, Z_i, \Gamma_s)], \quad A_i = \dots$$

- ▶ Suggested by [Cheridito-Soner-Touzi-Victoir]. Different from \mathbf{T}_h -scheme.

$$\mathbf{T}_h\text{-scheme} : D^{(2)}\varphi \approx \mathbb{E}[\varphi(t_{i+1}, \hat{X}_h^{t_i, x}) H_h^2]$$

$$\text{CSTV-scheme} : D^{(k)}\varphi \approx \mathbb{E}[\varphi(t_{i+1}, \hat{X}_h^{t_i, x}) H_{\frac{h}{2}}^1 H_{\frac{h}{2}}^1]$$

Convergence I

Assumption for convergence

The nonlinearity F is Lipschitz-continuous with respect to (r, p, γ) uniformly in (t, x) , Lipschitz continuous on x uniformly on other variables, $1/2$ -Hölder on t uniformly on other variables and $|F(t, x, 0, 0, 0)|_\infty < \infty$. Moreover, F is uniformly elliptic and dominated by the diffusion of the linear operator \mathcal{L}^X , i.e.

$$\varepsilon I_d \leq \nabla_\gamma F \leq a \text{ on } \mathbb{R}^d \times \mathbb{R} \times \mathbb{R}^d \times \mathbb{S}_d \quad \text{for some } \varepsilon > 0.$$

Convergence II

Convergence Theorem(F., Touzi, Warin)

Assume that the fully nonlinear PDE has comparison for bounded functions. Then for every bounded Lipschitz function g , there exists a bounded function v so that

$$v^h \longrightarrow v \quad \text{locally uniformly.}$$

In addition, v is the unique bounded viscosity solution of the problem.

Regularity

v is Lipschitz on x and $1/2$ -Hölder on t .

Rate of convergence I

Assumption for rate of convergence: HJB

The Lipschitz nonlinearity F is of the Hamilton-Jacobi-Bellman type:

$$F(t, x, r, p, \gamma) = \inf_{\alpha \in \mathcal{A}} \{ \mathcal{L}^\alpha(t, x, r, p, \gamma) \}$$

$$\mathcal{L}^\alpha(t, x, r, p, \gamma) := \frac{1}{2} \text{Tr}[\sigma^\alpha \sigma^{\alpha T}(t, x) \gamma] + b^\alpha(t, x) p + c^\alpha(t, x) r + f^\alpha(t, x)$$

where the functions μ , σ , σ^α , b^α , c^α and f^α satisfy:

$$|\mu|_\infty + |\sigma|_\infty + \sup_{\alpha \in \mathcal{A}} (|\sigma^\alpha|_1 + |b^\alpha|_1 + |c^\alpha|_1 + |f^\alpha|_1) < \infty.$$

Rate of convergence II

HJB+

The nonlinearity F satisfies **HJB**, and for any $\delta > 0$, there exists a finite set $\{\alpha_i\}_{i=1}^{M_\delta}$ such that for any $\alpha \in \mathcal{A}$:

$$\inf_{1 \leq i \leq M_\delta} |\sigma^\alpha - \sigma^{\alpha_i}|_\infty + |b^\alpha - b^{\alpha_i}|_\infty + |c^\alpha - c^{\alpha_i}|_\infty + |f^\alpha - f^{\alpha_i}|_\infty \leq \delta.$$

Rate of convergence Theorem(F., Touzi, Warin)

Assume that the final condition g is bounded Lipschitz-continuous.

Then, there is a constant $C > 0$ such that:

- (i) under Assumption HJB, we have $v - v^h \leq Ch^{1/4}$,
- (ii) under the stronger condition HJB+, we have $-Ch^{1/10} \leq v - v^h \leq Ch^{1/4}$.

Conditional expectation

- ▶ Method is backward. We can calculate $v^h(t, x)$ if we know $v^h(t + h, \cdot)$ at any point.
- ▶ Should we do simulate many paths at each point to estimate $v^h(t, x)$? [Touzi, Bouchard] and [Longstaff, Schwartz]
- ▶ $\mathbb{E}[\varphi(X_h^x)] = \mathbb{E}[\varphi(X_{t_i+1}) | X_{t_i} = x]$.
- ▶ For the second conditional expectation, we do not need to use the paths which satisfies $X_{t_i} = x$.
- ▶ Weighting the sample paths with their t_i time distance from x .

$$\mathbb{E}[\varphi(X_{t_i+1}) | X_{t_i} = x] \approx \frac{\sum_{l=1}^N \varphi(X_{t_i+1}^l) \kappa(X_{t_i}^l - x)}{\sum_{l=1}^N \kappa(X_{t_i}^l - x)}.$$

Mean curvature I

Mean curvature flow

$$v_t - \Delta v + \frac{Dv \cdot D^2 v Dv}{|Dv|^2} = 0 \quad \text{and} \quad v(0, x) = g(x)$$

Mean curvature II

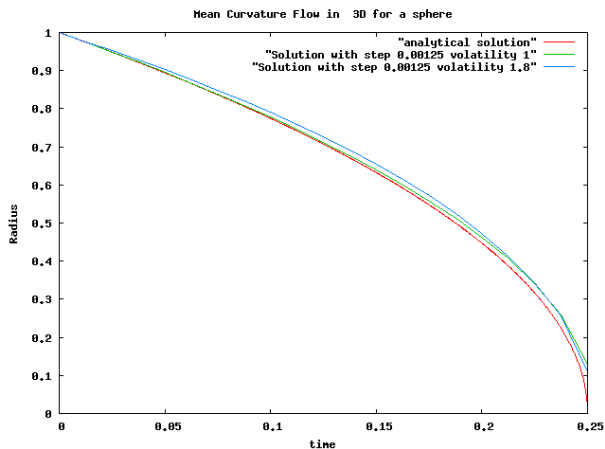


Figure: Solution of the mean curvature flow for the sphere problem

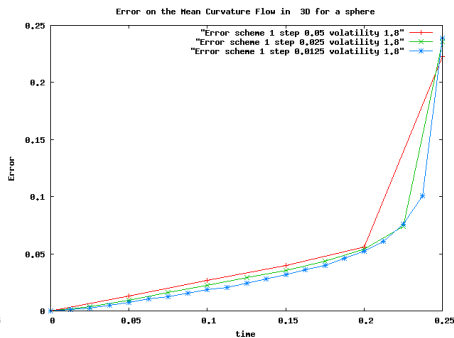
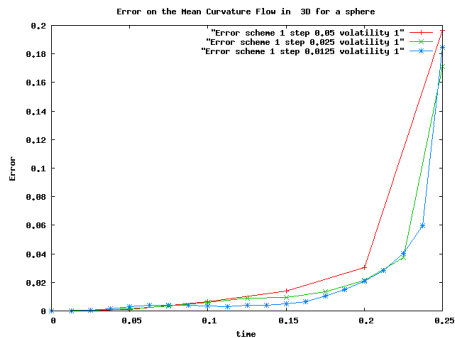


Figure: Mean curvature flow problem for different time step and diffusion: scheme 1

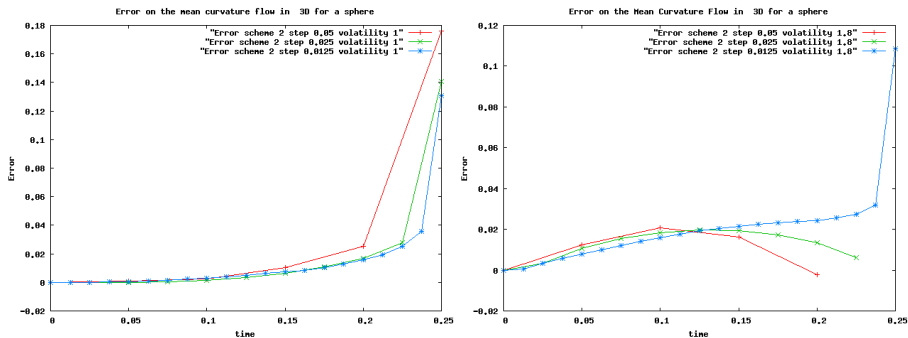


Figure: Mean curvature flow problem for different time step and diffusions: scheme 2

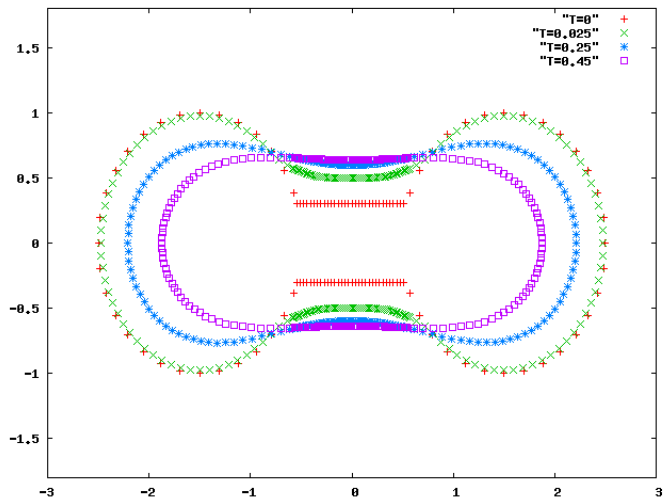


Figure: Mean curvature flow problem in 2D

Portfolio optimization

Two dimensional

Heston model:

$$dS_t = \mu S_t dt + \sqrt{Y_t} S_t dW_t^{(1)}$$

$$dY_t = k(m - Y_t) dt + c\sqrt{Y_t} \left(\rho dW_t^{(1)} + \sqrt{1 - \rho^2} dW_t^{(2)} \right).$$

HJB PDE

$$v(T, x, y) = -e^{-\eta x}$$

$$0 = -v_t - k(m - y)v_y - \frac{1}{2}c^2 y v_{yy} - \sup_{\theta \in \mathbb{R}} \left(\frac{1}{2}\theta^2 y v_{xx} + \theta(\mu v_x + \rho c y v_{xy}) \right)$$

$$= -v_t - k(m - y)v_y - \frac{1}{2}c^2 y v_{yy} + \frac{(\mu v_x + \rho c y v_{xy})^2}{2y v_{xx}}.$$

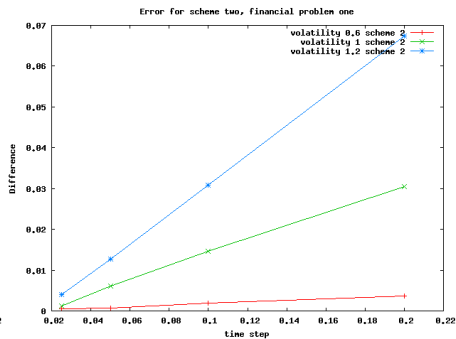
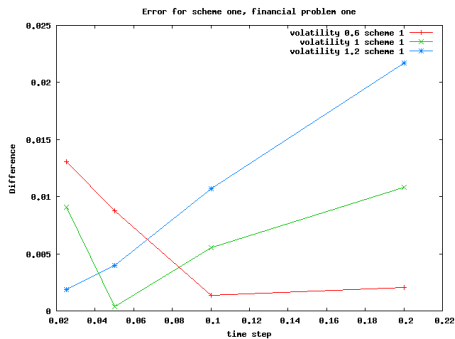


Figure: Difference between calculation and reference for scheme one and two

Portfolio optimization

Five dimensional

Vasicek model:

$$dr_t = \kappa(b - r_t)dt + \zeta dW_t^{(0)}.$$

CEV-SV model and Heston model.

$$\begin{aligned} dS_t^{(i)} &= \mu_i S_t^{(i)} dt + \sigma_i \sqrt{Y_t^{(i)}} S_t^{(i)\beta_i} dW_t^{(i,1)}, \quad \beta_2 = 1, \\ dY_t^{(i)} &= k_i (m_i - Y_t^{(i)}) dt + c_i \sqrt{Y_t^{(i)}} dW_t^{(i,2)} \end{aligned}$$

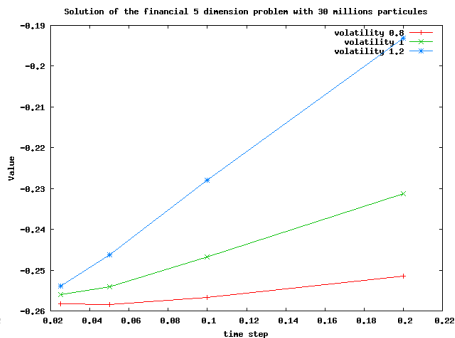
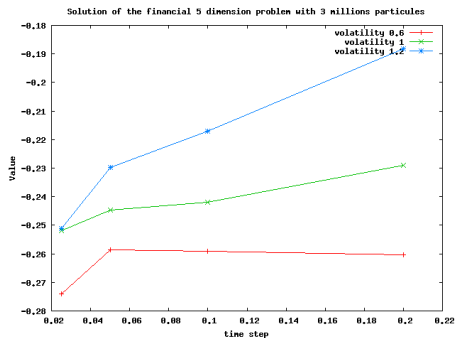


Figure: Five dimensional financial problem and its results for different volatilities with 3 millions and 30 millions particles

Nonlocal parabolic fully nonlinear PDE

$$\begin{aligned}
 -\mathcal{L}^X v(t, x) - F(t, x, v(t, x), Dv(t, x), D^2 v(t, x), v(t, \cdot)) &= 0, & \text{on } [0, T) \times \mathbb{R}^d, \\
 v(T, \cdot) &= g, & \text{on } \mathbb{R}^d.
 \end{aligned}$$

$$\begin{aligned}
 \mathcal{L}^X \varphi(t, x) &:= \left(\frac{\partial \varphi}{\partial t} + \mu \cdot D\varphi + \frac{1}{2} a \cdot D^2 \varphi \right)(t, x) \\
 &+ \int_{\mathbb{R}_*^d} (\varphi(t, x + \eta(t, x, z)) - \varphi(t, x) - \mathbf{1}_{\{|z| \leq 1\}} D\varphi(t, x) \eta(t, x, z)) d\nu(z).
 \end{aligned}$$

Discretization

$$\hat{X}_h^{t,x,\kappa} = x + \mu(t, x)h + \sigma(t, x)W_h + \int_{\{|z| \geq \kappa\}} \eta(z) \tilde{J}([0, h], dz),$$

$$\hat{X}_{t_{i+1}}^{x,\kappa} = \hat{X}_h^{t_i, \hat{X}_{t_i}^{x,\kappa}, \kappa} \quad \text{and} \quad \hat{X}_0^{x,\kappa} = x.$$

$$N_t^\kappa = \int_{\{|z| \geq \kappa\}} \tilde{J}([0, T], dz).$$

Notice that N_t^κ has intensity equal to $\lambda_\kappa := \int_{\{|z| \geq \kappa\}} \nu(dz)$.

$$\hat{X}_h^{t,x,\kappa} = x + (\mu - \lambda_\kappa)h + \sigma W_h + \sum_{i=1}^{N_h} \eta(Y_i),$$

where Y_i s are i.i.d. \mathbb{R}_*^d -valued random variables, independent of W distributed as $\frac{1}{\lambda_\kappa} \nu(dz)$.

Integration by part

We can choose κ dependent on h such that $h\lambda_\kappa = \lambda$ for some constant λ .

$$\hat{\nu}_h^\kappa(\psi)(t, x) := \mathbb{E} \left[\int_{\{|z| \geq \kappa\}} \psi(t + h, \hat{X}_h^x + \eta(z)) d\nu(z) \right].$$

Lemma

For every function $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}$ with exponential growth, we have:

$$\hat{\nu}_h^\kappa(\psi)(t, x) = \frac{1}{h} \mathbb{E}[\psi(t + h, \hat{X}_h^{t, x, \kappa}) N_h^\kappa].$$

N_h^κ is a Poisson process with parameter λ .

Integral operator

$$\mathcal{I}[t, x, r, p, \psi] := \int_{\mathbb{R}_*^d} (\psi(t, x + \eta(t, x, z)) - \psi(t, x) - \mathbf{1}_{\{|z| \leq 1\}} D\psi(t, x) \eta(t, x, z)) d\nu(z)$$

$$\mathcal{I}_h^\kappa[t, x, r, p, \psi] := \hat{v}_h^\kappa(\psi)(t, x) - \lambda r - \lambda p \cdot \mathbb{E} [\mathbf{1}_{\{|Y| \leq 1\}} \eta(Y)].$$

Final remarks

- ▶ The convergence is in L_∞ -norm.
- ▶ despite the convergence is established on uniform ellipticity assumption, the numerical tests shows the convergence in not uniformly elliptic cases.
- ▶ The rate of convergence may not be optimal. It is shown in [F., Touzi, Warin] that the rate for linear equation is $1/2$ from both sides.
- ▶ Rate of convergence is derived for convex(concave) nonlinearities.
- ▶ The proofs in nonlocal case do not change dramatically.
- ▶ Possible extensions: Elliptic problems, Obstacle problem.