

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

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Conclusions

- ▶ 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:

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

- ▶ 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:

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

- ▶ $Y_t = v(t, X_t)$ 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:

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

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

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

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

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 $Z_t := Dv(t, X_t)$:

$$\begin{aligned} Y_t &= g(X_T) + \int_t^T F(s, X_s, Y_s, Z_s, \Gamma_s) s d - \int_t^T Z_s dX_s \\ dZ_t &= A_t dt + \Gamma_t dW_t \end{aligned}$$

- ▶ $(Y_t, Z_t, \Gamma_t, A_t) =$
 $(v(t, X_t), Dv(t, X_t), D^2v(t, X_t), \mathcal{L}Dv(t, X_t)).$
- ▶ 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

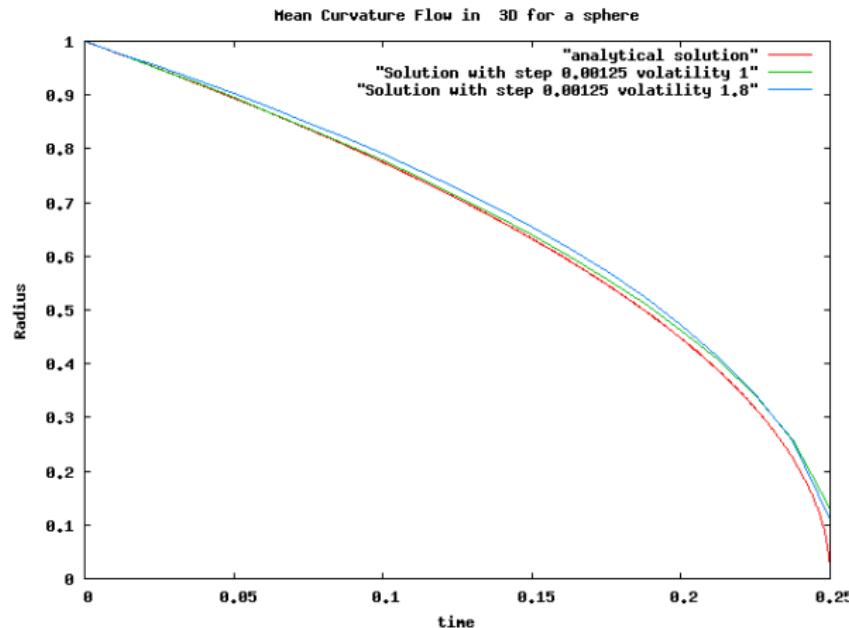


Figure: Solution of the mean curvature flow for the sphere problem

└ Numerical results

└ Non-financial test problem

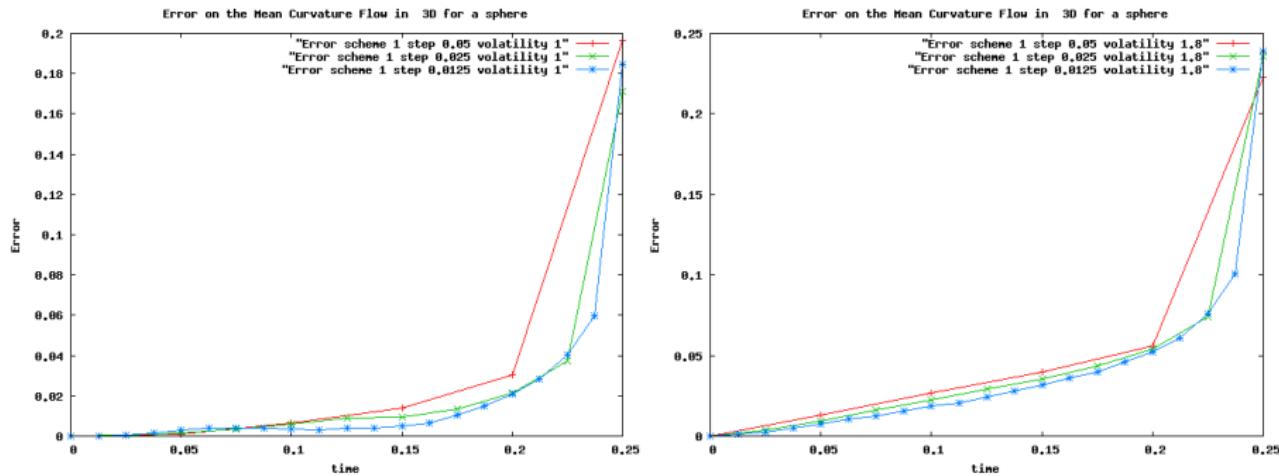


Figure: Mean curvature flow problem for different time step and diffusion: scheme 1

└ Numerical results

└ Non-financial test problem

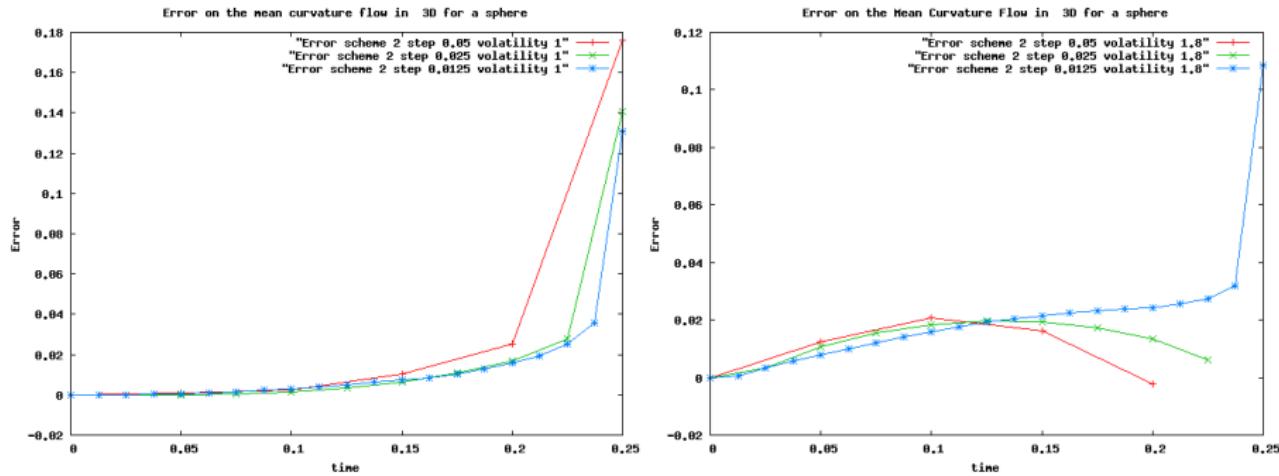


Figure: Mean curvature flow problem for different time step and diffusions: scheme 2

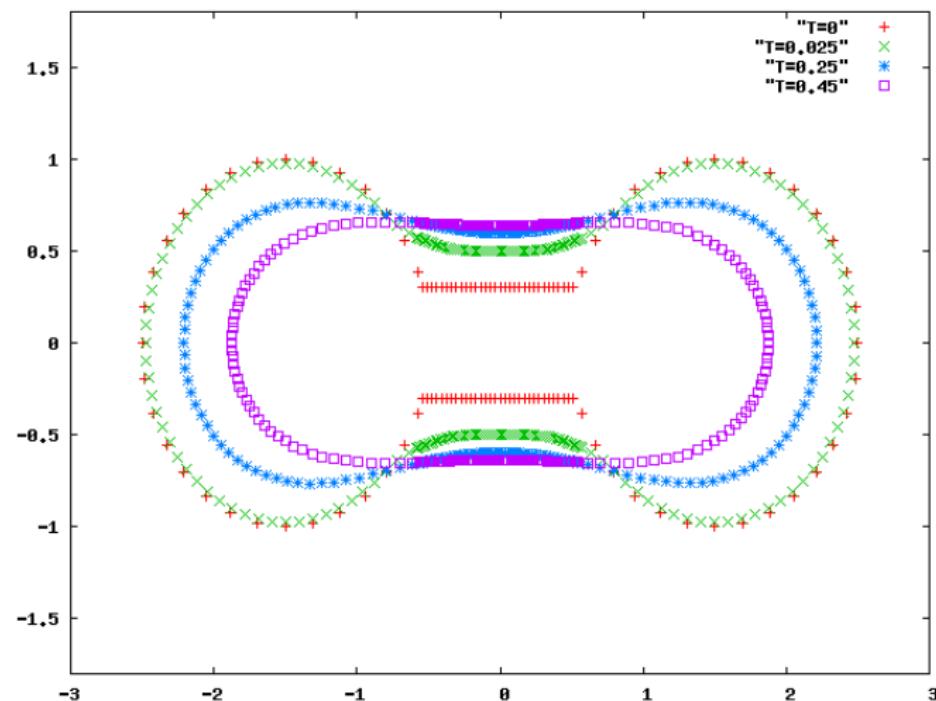


Figure: Mean curvature flow problem in 2D

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

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

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

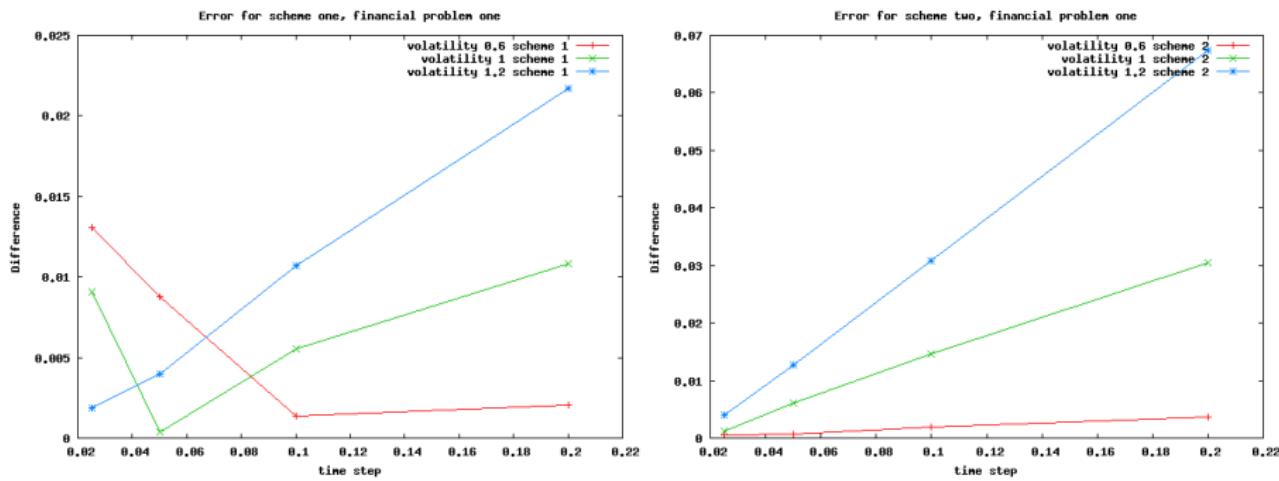


Figure: Difference between calculation and reference for scheme one and two

Portfolio optimization

Five dimensional

Vasicek model:

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

CEV-SV model and Heston model.

$$dS_t^{(i)} = \mu_i S_t^{(i)} dt + \sigma_i \sqrt{Y_t^{(i)}} S_t^{(i)\beta_i} dW_t^{(i,1)}, \quad \beta_2 = 1,$$

$$dY_t^{(i)} = k_i \left(m_i - Y_t^{(i)} \right) dt + c_i \sqrt{Y_t^{(i)}} dW_t^{(i,2)}$$

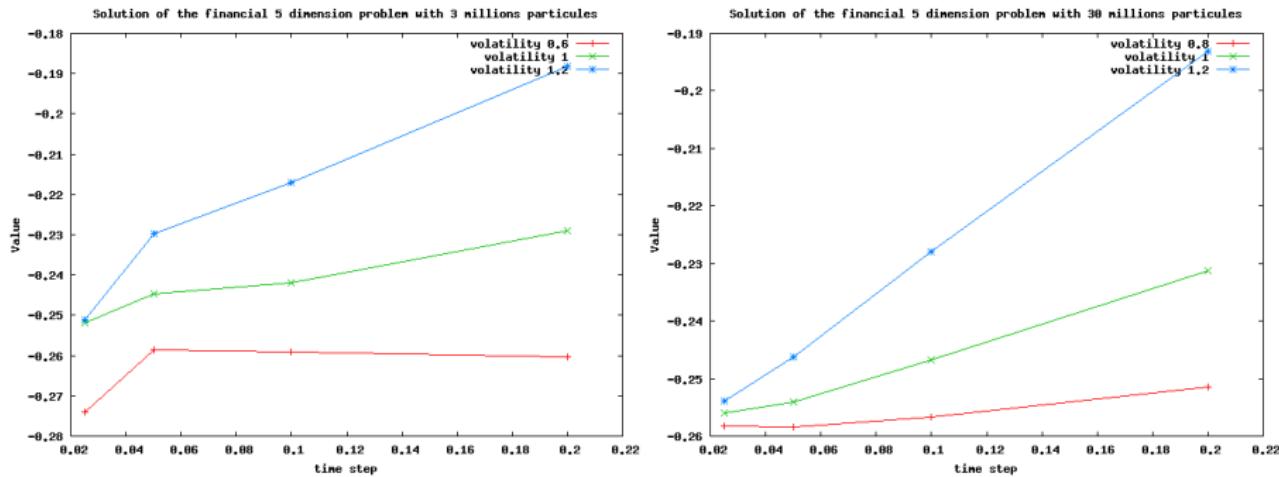


Figure: Five dimensional financial problem and its results for different volatilities with 3 millions and 30 millions particles

Nonlocal parabolic fully nonlinear PDE

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

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

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

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

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

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.