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Particle Approximations of Feynman-Kac representations

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- A. F-K representations for solutions of linear PDEs
 - classical approximations (Euler)
 - Kusuoka-Lyons-Victoir cubature methods
- B. F-K representations for solutions of linear SPDEs (Zakai)
 - particle filters (Sequential MCMCs)
 - applications of the K-L-V cubature method
- C. F-K representations for solutions of non-linear PDEs
 - Malliavin weights Monte-Carlo method
 - Kusuoka-Lyons-Victoir cubature based method

Joint work with S. Ghazali, K. Manolarakis

References :

- D.C., S. Ghazali, On the convergence rates of a general class of weak approximations of SDEs, in Stochastic differential equations: theory and applications, Eds P. Baxendale and S. Lototsky, 221-248, 2007.
- D.C., K. Manolarakis, N. Touzi, On the Monte-Carlo simulation of BSDEs: An improvement on the Malliavin weights, submitted 2008
- D.C., K. Manolarakis, Numerical solution for a BSDE using the Cubature methods, preprint 2008
- D.C., J. Xiong An approximate McKean-Vlasov model for the stochastic filtering problem, Conference Oxford sur les méthodes de Monte-Carlo séquentielles, 18-21 ESAIM Proc., 19 EDP Sci., Les Ulis, 2007
- D.C., J. Xiong Numerical solutions for a class of SPDEs over bounded domains. Conference Oxford sur les méthodes de Monte-Carlo séquentielles 121-125, ESAIM Proc. 19. EDP Sci. Les Ulis. 2007

Equivalently:
$$L = \sum_{i=1}^d f_i \frac{\partial}{\partial x_i} + \sum_{i,j=1}^d a_{ij} \frac{\partial^2}{\partial x_i \partial x_j}$$

where
$$a_{ij} = \frac{1}{2} \sum_{k=1}^d v_k^i v_k^j, \quad f_i = v_0^i + \sum_{k=1}^d \sum_{j=1}^d v_k^j \frac{\partial v_k^i}{\partial x_j}$$

(3)
$$X^{s,x}(t) = x + \int_s^t f(X^{s,x}(u)) du + \sum_{i=1}^d \int_s^t v_i(X^{s,x}(u)) dW_u^i \quad s \leq t \leq T$$

- The stochastic integral in (2) is a Stratonovich integral, whilst the one in (3) is Itô integral.
- In the following, wlog $s=0$

$$u(0, x) = E[\varphi(X^x(T))] = P_t \varphi(x),$$

where

$$\begin{cases} dX^x(t) = \sum_{i=0}^d v_i(X^x(t)) dW_t^i \\ X^x(0) = x \end{cases}$$

$$t \in [0, T]$$

$$\begin{aligned} W_t^0 &= t & dW_t^0 &= dt \\ (W^1, \dots, W^d) & \text{ - } d\text{-dimensional BM} \end{aligned}$$

Classical approximations

$0 = z_0 < z_1 < \dots < z_N = T$ equidistant partition of $[0, T]$ $z_{n+1} - z_n = \delta = \frac{T}{N}$.

Euler approximation $Y(z_n) = Y_n$

$$Y_{n+1} = Y_n + f(Y_n)\delta + \sum_{i=1}^d v_i(Y_n) \xi^{i,n} \quad \text{where } \xi^{i,n} \begin{cases} \text{mutually independent} \\ (*) \\ |E[\xi^{i,n}]|, |E[(\xi^{i,n})^3]|, |E[(\xi^{i,n})^2] - \delta| \leq K\delta^2 \end{cases}$$

$Y_0 = x$

Theorem If (a_{ij}) is strictly elliptic, then there exists K independent of δ s.t.

$$|E[\psi(X_T^*)] - E[\psi(Y_T)]| \leq K\delta$$

Remark $\xi^{i,n} \sim N(0, \delta)$ satisfies (*)

$$\xi^{i,n} = \begin{cases} \sqrt{\delta} & \text{prob } \frac{1}{2} \\ -\sqrt{\delta} & \text{prob } -\frac{1}{2} \end{cases} \quad \text{satisfies (*)}$$

2nd order approximation (d=1)

$$\begin{aligned} Y_{n+1} = & Y_n + f(Y_n) \delta + v(Y_n) \xi^n + \frac{1}{2} v(Y_n) v'(Y_n) ((\xi^n)^2 - \delta) \\ & + \frac{1}{2} (f(Y_n) f'(Y_n) + \frac{1}{2} f''(Y_n) (v(Y_n))^2) \delta^2 \\ & + (f'(Y_n) v(Y_n) + f(Y_n) v'(Y_n) + \frac{1}{2} v''(Y_n) v(Y_n)) \xi^n \cdot \delta \end{aligned}$$

where ξ^n mutually independent

$$(**) |E[\xi^n]| + |E[(\xi^n)^3]| + |E[(\xi^n)^2] - \delta| + |E[(\xi^n)^4] - 3\delta^2| \leq K \delta^3$$

then

$$|E[Y(X_T^*)] - E[Y(Y_T)]| \leq K \delta^2$$

Remark $\xi^n \sim N(0, \delta)$ satisfies (**)

$$\xi^n = \begin{cases} \sqrt{3\delta} & \text{prob } \frac{1}{6} \\ -\sqrt{3\delta} & \text{prob } \frac{1}{6} \\ 0 & \text{prob } \frac{2}{3} \end{cases}$$

Monte Carlo approximation of $E[\varphi(X_T)]$

Produce n independent realizations of X : X^1, \dots, X^n , then

$$E[|E[\varphi(X_T)] - \mu_n(\varphi)|] \leq \frac{C(\varphi)}{\sqrt{n}}$$

where

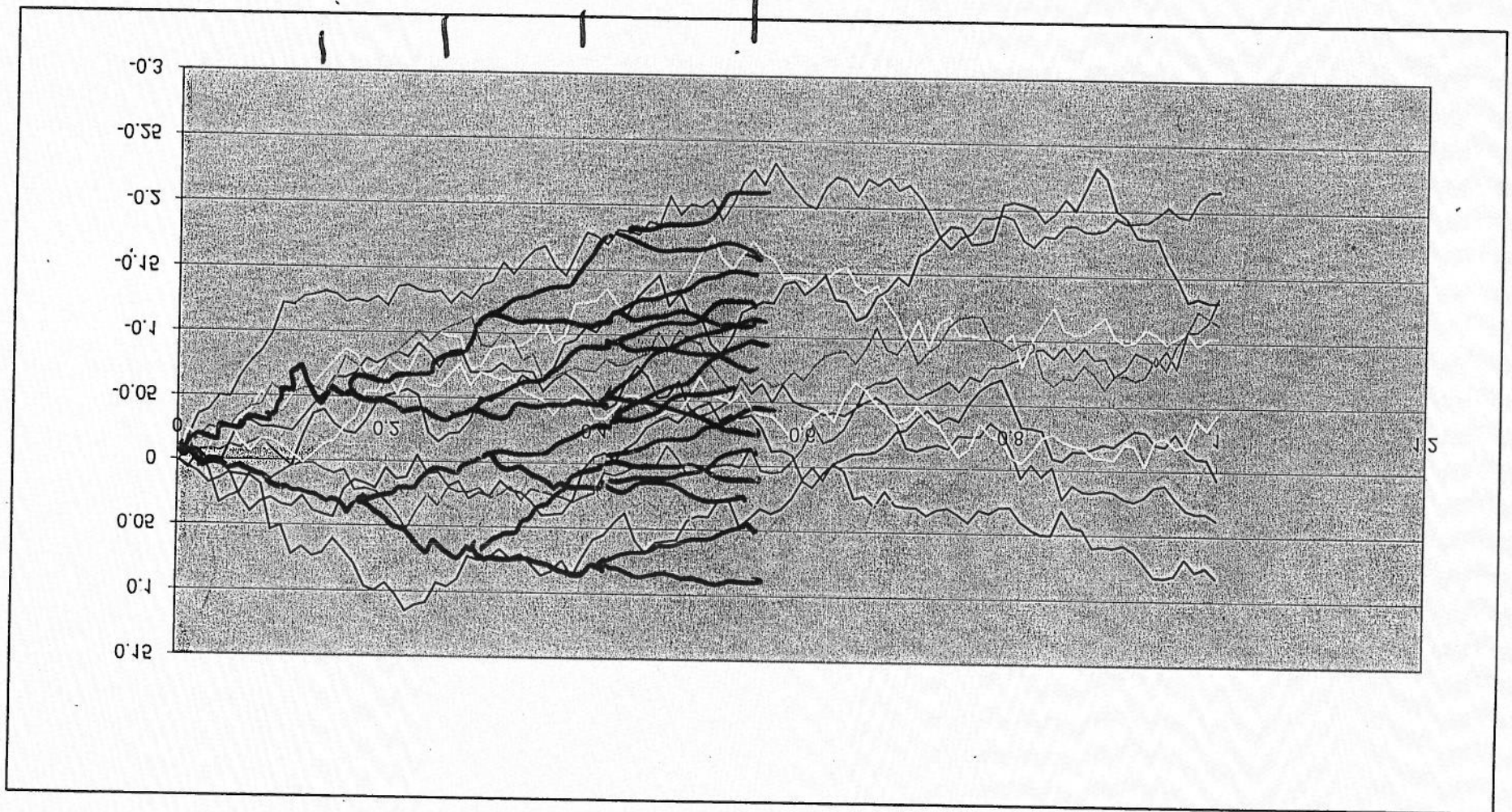
$$\mu_n(\varphi) = \frac{1}{n} \sum_{i=1}^n \varphi(X_T^i), \quad \mu_n = \frac{1}{n} \sum_{i=1}^n \int_{X_T^i}$$

$$C(\varphi)^2 = \text{Var}(\varphi(X_T)) = E[(\varphi(X_T))^2] - (E[\varphi(X_T)])^2$$

Remarks:

• μ_n is an approximation of R_{X_T} Error = $F(\delta, n)$
 $|E[\varphi(X_T)] - \mu_n(\varphi)| = \text{Systematic Error} + \text{Statistical Error}$

• $\bar{\mu}_n = \frac{1}{n} \sum_{i=1}^n \int_{(X_{t_0}^i, X_{t_1}^i, \dots, X_{t_n}^i)}$ is an approximation of $R_{X_{[0,T]}}$



Disadvantages:

- The methods require the ellipticity condition imposed on the matrix (a_{ij}) or the Hörmander condition imposed on $\{V_0, V_1, \dots, V_d\}$ to obtain the optimal convergence rates.
- The approximating processes can leave the SDE's reachability surface, thus leading to instability problems and poor results.
- There is equal computational effort allocated to each interval of the partition: wasteful at the beginning, not enough towards the end.
- A "tree-like" approximation would be desirable.

A new class of schemes have been introduced to answer these problems :

Kusuoka [2001, 2004]

Kusuoka and Ninomiya [2004]

Lyons and Victoir [2004]

Ninomiya [2003]

Ninomiya and Victoir [2004]

Lyons and Litterer [2008]

Ninomiya and Ninomiya [2009]

Major improvements in computational effort :

- Ninomiya [2003] reports a speed-up ratio of the Kusuoka-Ninomiya algorithm over Euler of "about 6580"
- Ninomiya and Victoir [2004] report that their algorithm combined with the Romberg extrapolation is "about 800" times faster than Euler.

The UFG Condition [Kusuoka & Stroock]

i. $\mathcal{A} = \{\emptyset\} \cup \bigcup_{m=1}^{\infty} \{0, 1, \dots, d\}^m$, $\mathcal{A}_1 = \mathcal{A} \setminus \{\emptyset, (0)\}$, $\mathcal{A}_1(m) = \{\alpha \in \mathcal{A}, 1 \leq |\alpha| \leq m\}$

ii. Two norms $|\emptyset| = 0$ $|\alpha| = r$ if $\alpha = (i_1, \dots, i_r)$
 $\|\alpha\| = |\alpha| + \#\{i \leq j \leq |\alpha| \mid i_j = 0\}$

iii. Concatenation on \mathcal{A}

$$(i_1, \dots, i_r) * (j_1, \dots, j_s) = (i_1, \dots, i_r, j_1, \dots, j_s)$$

iv. $V_{[\alpha]}$ $\alpha \in \mathcal{A}$ defined as: $V_{[\emptyset]} = 0$, $V_{[(j)]} = V_j$

$$V_{[\alpha * (j)]} = [V_{[\alpha]} V_j] = V_{[\alpha]} V_j - V_j V_{[\alpha]}$$

The UFG condition: there exists $\rho > 0$ such that for all $\alpha \in \mathcal{A}_1$

$$V_{[\alpha]} = \sum_{\beta \in \mathcal{A}_1(\rho)} \gamma_{\alpha, \beta} V_{[\beta]}$$

$$\gamma_{\alpha, \beta} \in C_b^\infty(\mathbb{R}^d)$$

Remarks:

- The UFG condition is weaker than the Uniform Hörmander Condition
- Algebra generated by $\{V_{[\alpha]}\}_{\alpha \in A}$ is finite dimensional as a $C_b^\infty(\mathbb{R}^d)$ -module.

Theorem [Kusuoka & Stroock]

Under the UFG condition, for $p = 1, \dots, \overbrace{\|\alpha_1 * \dots * \alpha_r\|}^r$

$$\|V_{[\alpha_1]} \dots V_{[\alpha_r]} P_t \varphi\|_\infty \leq C_p^T \frac{1}{t^{\frac{1}{2}(r-p)}} \|\varphi\|_p \quad t \in (0, t]]$$

where

$$\|\varphi\|_p = \sum_{i=1}^p \|\nabla^i \varphi\|_\infty \quad \varphi \in C_b^p(\mathbb{R}^d)$$

$$\|\nabla^i \varphi\|_\infty = \max_{\{1, \dots, i\} \in \{1, \dots, d\}} \left\| \frac{\partial^i \varphi}{\partial x_{d_1} \dots \partial x_{d_i}} \right\|_\infty.$$

M-Perfect Families

Stochastic Taylor expansion of $\varphi(X_t)$

$$\varphi(X_t) = \varphi(x) + \underbrace{\sum_{\substack{\alpha \in \mathcal{A}_0(m) \\ \alpha = (i_1, \dots, i_r)}} V_{\alpha} \varphi(x) \int_0^t \int_0^{s_0} \dots \int_0^{s_{r-2}} \circ dW_{s_{r-1}}^{i_1} \dots \circ dW_{s_1}^{i_{r-1}} \circ dW_{s_0}^{i_r}} + R_{m,t,\varphi}(x)}_{\varphi_t^m(x)}$$

Let $\bar{X}(x) = \{\bar{X}_t(x)\}_{t \in [0, \tau], x \in \mathbb{R}^d}$ be a family of (explicitly solvable) processes such that

$$\lim_{y \rightarrow x} \bar{X}_t(y) = \bar{X}_t(x) \text{ P.a.s. for any } (t, x) \in [0, t] \times \mathbb{R}^d$$

Let $Q_t \varphi(x) = E[\varphi(\bar{X}_t(x))]$ $t \in [0, \tau]$, $Q_t \varphi \in C_b(\mathbb{R}^d)$ for $\varphi \in C_b(\mathbb{R}^d)$

Definition The family $\bar{X}(x) = \{\bar{X}_t(x)\}_{(t,x) \in [0, \tau] \times \mathbb{R}^d}$ is m-perfect for X if

$$\sup_{x \in \mathbb{R}^d} |Q_t \varphi(x) - E[\varphi_t^m(x)]| \leq C \left[\sum_{l=m+1}^M t^{1/2} \|\varphi\|_{V_i} + t^{m+1} \|\nabla \varphi\|_{\infty} \right]$$

where
$$\|\varphi\|_{V,i} = \sum_{v=1}^i \sum_{\substack{\alpha_1, \dots, \alpha_v \in \mathcal{A}_T \\ \|\alpha_1 + \dots + \alpha_v\| = i}} \|\mathcal{V}_{\{\alpha_1\}} \dots \mathcal{V}_{\{\alpha_v\}} \varphi\|_\infty$$

Remarks:

- $\mathcal{Q}_t \varphi$ will have the same truncation as $P_t \varphi$
- $P_t \varphi$ will be approximated by $\mathcal{Q}_{h_n}^m \mathcal{Q}_{h_{n-1}}^m \dots \mathcal{Q}_{h_1}^m$ $h_i = \tau_i - \tau_{i-1}$.

Lemma: Let $\{\bar{\mathcal{X}}_t(x)\}$ be an m -perfect family. then

$$\|P_t(P_S \varphi) - \mathcal{Q}_t(P_S \varphi)\|_\infty \leq C \|\varphi\|_p \sum_{j=m+1}^M \frac{t^{j/2}}{S^{j-\frac{p}{2}}} \quad \varphi \in C_b^p(\mathbb{R}^d)$$

The Global Error corresponding to an m -Perfect Family

Let $T, \gamma > 0$, $\Pi_n = \{z_j = (\frac{j}{n})^\gamma T\}_{j=0}^n$ $h_j = z_j - z_{j-1} \in [0, 1]$

Theorem [Crisan & Ghazali 2007] Let $\text{Error}^{\gamma, n}(\varphi) = \|P_t \varphi - Q_{h_n} Q_{h_{n-1}} \dots Q_{h_1} \varphi\|_\infty$
Then

i. For any $\varphi \in C_b^m(\mathbb{R}^d)$ and $\gamma > \frac{m-1}{M}$

$$\text{Error}^{\gamma, n}(\varphi) \leq c(\gamma, M, t) \frac{1}{h^{\frac{m-1}{2}}} \|\varphi\|_M$$

ii. For any $\varphi \in C_b^1(\mathbb{R}^d)$ and $\gamma > m-1$

$$\text{Error}^{\gamma, n}(\varphi) \leq \bar{c}(\gamma, M, t) \frac{1}{h^{\frac{m-1}{2}}} \|\varphi\|_1$$

provided $\sup_{x \in \mathbb{R}^d} |\bar{X}_t(x) - x| \leq c\sqrt{t}$ for t small.

Example: The Lyons-Victoir cubature method:

A cubature of order m for a finite measure μ is a set of points $\{x_i\}_{i=1}^N \in \mathbb{R}^d$ and a set of non-negative numbers $\lambda_i, i=1, \dots, N$ such that

$$\int_{\mathbb{R}^d} x^k \mu(dx) = \sum_{i=1}^N \lambda_i x_i^k \quad \forall k \leq m \quad \left[\mu \approx \sum_{i=1}^N \lambda_i \delta_{\{x_i\}} \right]$$

Hence for any smooth function $f: \mathbb{R}^d \rightarrow \mathbb{R}^d$ we have

$$\int_{\mathbb{R}^d} f(x) \mu(dx) \approx \sum_{i=1}^N \lambda_i f(x_i)$$

The cubature method of Lyons and Victoir realises the same idea on the Wiener space.

Theorem (Lyons & Victoir (2004)): There exist paths $\omega_1, \dots, \omega_N$ with bounded variation and $\lambda_1, \lambda_2, \dots, \lambda_N$ such that

$$E \left[\int_0^t \int_0^{s_0} \dots \int_0^{s_{i-2}} \circ dW_{s_{r-1}}^{i_1} \dots \circ dW_{s_1}^{(r-1)} \circ dW_{s_0}^{(r)} \right]$$

Remarks

- A 3-perfect family is required to obtain an approximation of the same order as the Euler method.
- A 5-perfect family is required to obtain an approximation of the same order as a classical 2nd order approximation
- In general, the rougher φ is, the slower the rate of convergence becomes.
- By varying γ , we can obtain the optimal rate regardless of the smoothness of φ (it needs to be at least Lipschitz)
- to limit the exponential-increase in the computational effort one can apply one of 4 methods
 - naive Monte-Carlo method
 - the Tree Based Branching Algorithm of Crisan and Lyons
 - the recombination scheme of Schmeiser, Soreff and Teichmann
 - the recombining algorithm of Lyons and Litterer

$$= \sum_{j=1}^N \lambda_j \int_0^{t+s_0} \int_0^{s_{i-2}} \dots \int_0^{s_{i-1}} d\omega_{s_{r-1}}^{i_1} \dots d\omega_{s_1}^{i_{r-1}} d\omega_{s_0}^{i_r}$$

for any multi-index $j = (i_1, \dots, i_r)$ such that $\|j\| \leq m$.

The measure $\mathbb{Q}_t^m = \sum_{i=1}^N \lambda_i \int \omega_i$ on $C([0, T], \mathbb{R}^d)$ is called a cubature measure of order m . In particular

$$\mathbb{E}[\varphi_t^m(x)] = \mathbb{Q}_t^m[\varphi_t^m(x)]$$

Choose $X^{j,x}$ such that

$$dX^{j,x}(t) = v_0(X^{j,x}(t))dt + \sum_{j=1}^d v_j(X^{j,x}(t))d\omega_t^j.$$

Then $\bar{X}_t(x) = \sum_{j=1}^N \lambda_j \bar{X}_t^{j,x}$ is an m -perfect family.

Cubature of order 3 $N = 2^d$, $\lambda_j = \frac{1}{2^d}$ $\omega_t^j = t(z_{j,1}^+, \dots, z_{j,d}^+)$
 $z_{j,1}^+, \dots, z_{j,d}^+ \in \{-1, 1\}$