

# Numerical analysis of the DMC method in a simple case.

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### Structure of the talk



2 The Diffusion Monte Carlo method : a variance reduction technique

- 3 Analysis of the bias : the fixed-node approximation
- 4 Numerical implementation



### The Born-Oppenheimer approximation

For most applications, systems of limited size (e.g. molecules) are described by

- *M* nuclei with electric charges *z<sub>k</sub>* ∈ N\* and positions *x<sub>k</sub>* ∈ ℝ<sup>3</sup>, *k* ∈ {1,...,*M*}.
   slow variables → modelled by classical mechanics.
- *N* electrons with positions x<sub>i</sub> ∈ ℝ<sup>3</sup>, i ∈ {1,...,N} and charge −1. Very light particles. fast variables → modelled by (nonrelativistic) quantum mechanics.

In the Born-Oppenheimer approximation, for electronic computations, the nuclei positions are supposed to be constant.



### Electronic state

To simplify, we leave the spin variables aside. Electronic state  $\rightarrow$  modelled by the Hamiltonian

$$H = -\sum_{i=1}^{N} \frac{1}{2} \Delta_{x_i} - \sum_{i=1}^{N} \sum_{k=1}^{M} \frac{z_k}{|x_i - \bar{x}_k|} + \sum_{1 \le i < j \le N} \frac{1}{|x_i - x_j|}$$

where the positions  $\bar{x}_k$  of the nuclei are supposed to be fixed. Electronic ground state energy :

$$E_0 = \inf\left\{\langle\psi, H\psi\rangle, \quad \psi \in D(H), \ \int_{\mathbb{R}^{3N}} |\psi|^2 = 1\right\}$$
(1)

with

$$D(H) = \left\{ \psi \in \bigwedge_{i=1}^{N} L^{2}(\mathbb{R}^{3}) \quad , \quad H\psi \in \bigwedge_{i=1}^{N} L^{2}(\mathbb{R}^{3}) \right\}$$



### Electronic state

#### Antisymmetrized tensor product space

$$\bigwedge_{i=1}^{N} L^{2}(\mathbb{R}^{3}) = \{ \psi \in L^{2}((\mathbb{R}^{3})^{N}) : \forall \sigma \in \mathcal{S}_{n}, \ \forall x \in (\mathbb{R}^{3})^{N}, \ \psi(x_{\sigma}) = (-1)^{\varepsilon(\sigma)}\psi(x) \}$$

with 
$$x_{\sigma} = (x_{\sigma(1)}, \dots, x_{\sigma(N)})$$
 for  $x = (x_1, \dots, x_N)$  and  $\varepsilon(\sigma)$  signature of  $\sigma$ .

Justified by

- exchangeability of the electrons (in terms of the density  $|\psi|^2$ :  $|\psi(x_{\sigma})|^2 = |\psi(x)|^2$ )
- Pauli's exclusion principle for fermions ( $|\psi|^2$  vanishes when two positions  $x_i$  are equal)



### Ground state properties

**Ground state** = element  $\psi_0$  of D(H) which minimizes the energy (1)

$$\langle \psi_0, H\psi_0 \rangle = E_0$$
 with  $\|\psi_0\|_2 = 1$ .

#### **Theorem 1**

- When  $N \leq Z = \sum_{k=1}^{M} z_k$  (neutral molecule or positive ion), then H is self-adjoint in D(H) and there exists a ground state  $\psi_0$  (Zhislin 1960).
- Any ground state  $\psi_0$  belongs to  $C^{\theta}(\mathbb{R}^{3N})$  for  $\theta \in (0,1)$  and to  $C^{\infty}(\mathbb{R}^{3N} \setminus \Gamma)$  where

$$\Gamma = \{ (x_1, \ldots, x_N) \in (\mathbb{R}^3)^N : \exists i \neq j \text{ s.t. } x_i = x_j \text{ or } \exists i, k \text{ s.t. } x_i = \bar{x}_k \}.$$

Any ground state  $\psi_0$  solves the Euler-Lagrange equation

$$H\psi_0 = E_0\psi_0. \tag{2}$$



### The Tiling property

For any continuous function  $\psi$  on  $\mathbb{R}^{3N}$ , we define

 $U = \mathbb{R}^{3N} \setminus \psi^{-1}(0).$ 

For  $\sigma \in S_N$  and  $x = (x_1, \ldots, x_N) \in (\mathbb{R}^3)^N$ , we set  $x_{\sigma} = (x_{\sigma(1)}, \ldots, x_{\sigma(N)})$ . When  $\psi$  is antisymmetric, for any connected component C of U and  $\forall \sigma \in S_N$ ,

$$\sigma(\mathcal{C}) = \{x_{\sigma} : x \in \mathcal{C}\}$$

is also a connected component of *U*.

#### **Theorem 2**

• Any ground state  $\psi_0$  satisfies the tiling property : for any connected component C of  $U_0 = \mathbb{R}^{3N} \setminus \psi_0^{-1}(0)$ ,  $U_0 = \bigcup_{\sigma \in S_N} \sigma(C)$  (Ceperley 91).

• Moreover, for any connected component C of  $U_0$ ,

$$E_0 = E_{\mathcal{C}} \stackrel{\text{def}}{=} \inf \left\{ \frac{1}{2} \int_{\mathcal{C}} |\nabla \psi|^2 + \int_{\mathcal{C}} V \psi^2, \ \psi \in H^1_0(\mathcal{C}), \ \int_{\mathcal{C}} \psi^2 = 1 \right\}.$$



### Numerical computation of the ground state energy

#### Difficult problem because

- dimension 3N with N large,
- antisymmetry condition due to the fermionic nature of electrons

#### Some numerical methods:

- Hartree-Fock methods (*variational approximation* : *restriction of* D(H) *to Slater determinants* det( $\phi_i(x_j)$ )),
- Density Functional Theory (Thomas-Fermi, Kohn-Sham),
- Quantum Monte Carlo methods (Variational Monte Carlo, Diffusion Monte Carlo).

see E. Cancès, M. Defranceschi, W. Kutzelnigg, C. Le Bris and Y. Maday, Computational Quantum Chemistry: a Primer, Handbook of Numerical Analysis, volume X (2003).



### Trick : Schrödinger in complex time

Add a ficticious time variable :  $\phi(t, x) = e^{-tH}\psi_I(x)$  solves

$$\begin{cases} \partial_t \phi = -H\phi = \frac{1}{2}\Delta\phi - V\phi \\ \phi(0,.) = \psi_I(.) \in D(H) \end{cases}$$

with 
$$V(x) = -\sum_{i=1}^{N} \sum_{k=1}^{M} \frac{z_k}{|x_i - \overline{x}_k|} + \sum_{1 \le i < j \le N} \frac{1}{|x_i - x_j|}$$
.

If  $E_0$  isolated single eigenvalue for H associated with eigenstate  $\psi_0$ ( $\|\psi_0\|_2 = 1$ ) and  $\langle \psi_I, \psi_0 \rangle \neq 0$ , then

$$\phi(t, x) = e^{-tH}\psi_I(x) \sim e^{-E_0 t} < \psi_I, \psi_0 > \psi_0(x) \text{ (t large)}$$
$$\Rightarrow E_0 = -\lim_{t \to +\infty} \frac{1}{t} \log |\phi(t, x)|. \tag{3}$$

(



## Feynman-Kac interpretation If $(W_t)$ Brownian motion in $\mathbb{R}^{3N}$ , for $r \in [0, t]$ ,

$$d_r e^{-\int_0^r V(x+W_s)ds} \phi(t-r, x+W_r) = e^{-\int_0^r V(x+W_s)ds} \left(\nabla_x \phi(t-r, x+W_r).dW_r + \underbrace{\left[-\partial_t \phi + \frac{1}{2}\Delta\phi - V\phi\right]}_0(t-r, x+W_r)dr\right).$$

#### Therefore

$$e^{-\int_{0}^{t} V(x+W_{s})ds}\psi_{I}(x+W_{t}) = \phi(t,x) + \int_{0}^{t} e^{-\int_{0}^{r} V(x+W_{s})ds} \nabla_{x}\phi(t-r,x+W_{r}).dW_{r}$$

and 
$$\phi(t, x) = \mathbb{E}\left(\psi_I(x + W_t)e^{-\int_0^t V(x + W_s)ds}\right).$$
  
By (3),

$$\Rightarrow E_0 = -\lim_{t \to +\infty} \frac{1}{t} \log \left| \mathbb{E} \left( \psi_I(x + W_t) e^{-\int_0^t V(x + W_s) ds} \right) \right|$$



### Variance reduction

$$E_0 = -\lim_{t \to +\infty} \frac{1}{t} \log \left| \mathbb{E} \left( \psi_I(x + W_t) e^{-\int_0^t V(x + W_s) ds} \right) \right|$$

**Problem** : large fluctuations of *V* in the exponential factor  $\Rightarrow$  variance too large. Need of variance reduction.

Principle of the Diffusion Monte Carlo method (importance sampling)

- choose ψ<sub>I</sub> ∈ D(H) as close as possible to the ground state ψ<sub>0</sub>,
- modify the dynamics of the Brownian motion by adding the drift term  $\frac{\nabla \psi_l}{\psi_l}$ ,
- replace *V* by  $E_L = \frac{H\psi_I}{\psi_I}$  in the exponential factor (when  $\psi_I = \psi_0$ ,  $E_L = E_0$  constant).



#### The Diffusion Monte Carlo method Yields very good results and is widely used in the chemistry community.

Benchmark guantum Monte Carlo 1435

TABLE L Atomization energies (kcal/mol) for the 55 molecules in the G1 set (Refs. 1, 2). Diffusion Monte Carlo (DMC) calculations and experimental (Expt.) results are listed. For DMC, statistical error bars are given in parentheses. Experimental errors are listed in parentheses (a dash indicates no error was available).

Molecule	DMC	Expt.
LiH	55.3(2)	56.00(1)
BeH	43.0(2)	46.90(1)
CH	79.5(2)	79.90(2)
$CH_{2}({}^{3}B_{1})$	181.9(4)	179.6(4)
$CH_2({}^{1}A_1)$	169.7(4)	170.6(4)
CH <sub>3</sub>	290.9(2)	289.3(2)
CH	395.0(2)	392.5(1)
NH	78.2(4)	79.0(4)
NH <sub>2</sub>	169.2(4)	170.0(3)
NH <sub>3</sub>	276.5(2)	276.7(1)
OH	101.2(3)	101.4(3)
H <sub>2</sub> O	219.4(2)	219.35(1)
HF	135.9(2)	135.2(2)
$SiH_2(^1A_1)$	145.5(2)	144.4(2)

J.C. Grossman, J. Chem. Phys., 117 (2002).

The Diffusion Monte Carlo method : a variance reduction technique





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L The Diffusion Monte Carlo method : a variance reduction technique



### The DMC method

#### Let

$$E(t) \stackrel{\text{def}}{=} \frac{\langle e^{-tH}\psi_I, H\psi_I \rangle}{\langle e^{-tH}\psi_I, \psi_I \rangle}$$

By spectral decomposition,

$$E(t) \sim rac{E_0 < \psi_I, \psi_0 >^2 e^{-E_0 t}}{< \psi_I, \psi_0 >^2 e^{-E_0 t}} \to E_0 \ \ ext{as } t \to +\infty.$$

For  $f(t, x) = \psi_I(x)e^{-tH}\psi_I(x) = \psi_I(x)\phi(t, x)$  and  $E_L = \frac{H\psi_I}{\psi_I}$ ,

$$E(t) = \frac{\langle f(t,.), E_L(.) \rangle}{\langle f(t,.), 1 \rangle}$$

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### Probabilistic interpretation of E(t)

The function  $f(t, x) = \psi_I(x)e^{-tH}\psi_I(x) = \psi_I(x)\phi(t, x)$  solves

$$\partial_t f = \frac{1}{2} \Delta f - \nabla .(bf) - \mathbf{E}_L f, \ f(0,.) = \psi_I^2(.)$$
(4)

where  $b = \frac{\nabla \psi_l}{\psi_l}$  and  $E_L = \frac{H\psi_l}{\psi_l} = -\frac{1}{2} \frac{\Delta \psi_l}{\psi_l} + V$ . Assume  $\|\psi_l\|_2 = 1$ .

Without the term -E<sub>L</sub>f, (4) Fokker-Planck equation for the density of X<sub>t</sub> solving

$$dX_t = dW_t + b(X_t)dt, \quad X_0 \sim \psi_I^2(x)dx.$$
(5)

• With this term, h(t, x) defined by

$$\forall g: \mathbb{R}^{3N} \to \mathbb{R}, \ \int_{\mathbb{R}^{3N}} g(x)h(t,x)dx = \mathbb{E}\left(g(X_t)e^{-\int_0^t E_L(X_s)ds}\right)$$

solves (4).



### Probabilistic interpretation of E(t)

One expects E(t) to be equal to

$$\frac{\langle h(t,.), E_L(.) \rangle}{\langle h(t,.), 1 \rangle} = \frac{\mathbb{E}\left(E_L(X_t)e^{-\int_0^t E_L(X_s)ds}\right)}{\mathbb{E}\left(e^{-\int_0^t E_L(X_s)ds}\right)} \stackrel{\text{def}}{=} E^{DMC}(t).$$

#### Variance reduction :

- if  $\psi_I$  close to the ground state  $\psi_0$ ,  $E_L = \frac{H\psi_I}{\psi_I} \sim \frac{H\psi_0}{\psi_0} = E_0$  fluctuates less than *V*,
- the drift  $b = \nabla \log(|\psi_I|)$  drives the process  $X_t$  solving (5) where  $|\psi_I|$  (and hopefully  $|\psi_0|$ ) is large.

 $b = \frac{\nabla \psi_I}{\psi_I} \text{ and } E_I = \frac{H \psi_I}{\psi_I} \text{ both explode near the nodal surface } \psi_I^{-1}(0)$ which is not empty because of the antisymmetry of  $\psi_I$  (ex:  $\{x : x_1 = x_2\} \subset \psi_I^{-1}(0)$ )  $\Rightarrow$  previous interpretation formal. Questions :  $E(.) = E^{DMC}(.)$ ?  $E^{DMC}(t) \rightarrow_{t \to +\infty} E_0$ ?





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### The fixed node approximation

#### **Theorem 3**

Assume that  $U_I = \mathbb{R}^{3N} \setminus \psi_I^{-1}(0)$  has a finite number of connected components. Under some technical assumptions on V and  $\psi_I$ , as  $t \to +\infty$ ,  $E^{DMC}(t)$  converges exponentially to

 $E_0^{DMC} = \inf\{ <\psi, H\psi >: \psi \in D(H), \ \|\psi\|_2 = 1, \ \psi_I^{-1}(0) \subset \psi^{-1}(0) \}.$ 

In addition,  $E_0^{DMC} \ge E_0$  with equality iff  $\psi_I^{-1}(0) \subset \psi_0^{-1}(0)$  where  $\psi_0$  ground state of H. (Cancès, Jourdain, Lelièvre, M3AS 2006)

The zeros of  $\psi_I$  ({ $x : \psi_I(x) = 0$ }) are called the nodes of  $\psi_I \rightarrow$  Fixed Node Approximation.

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### Elements of proof Step 1: $\forall x \in \mathbb{R}^{3N} \setminus \psi_I^{-1}(0)$ , the SDE

$$X_t^x = x + W_t + \int_0^t b(X_s^x) ds$$

admits a unique solution. In addition,  $\forall t \ge 0, \ \psi_I(x)\psi_I(X_t^x) > 0.$ 

<u>Step 2</u>: As  $b = \frac{\nabla \psi_I}{\psi_I}$ ,  $\frac{1}{2}\nabla \psi_I^2 - b\psi_I^2 = 0$  and  $\psi_I^2(x)dx$  reversible measure for the SDE. Hence

$$\begin{aligned} \forall g: \mathbb{R}^{3N} \to \mathbb{R}, \ \int_{\mathbb{R}^{3N}} g(x)h(t,x)dx \stackrel{\text{def}}{=} \mathbb{E}\left(g(X_t)e^{-\int_0^t E_L(X_s)ds}\right) \\ &= \int_{\mathbb{R}^{3N}} g(x)\psi_I^2(x)\mathbb{E}\left(e^{-\int_0^t E_L(X_s^x)ds}\right)dx. \end{aligned}$$

Hence  $h(t, x) = \psi_I^2(x) \mathbb{E}\left(e^{-\int_0^t E_L(X_s^x) ds}\right).$ 



### Elements of proof

$$h(t, x) = \psi_I^2(x) \mathbb{E}\left(e^{-\int_0^t E_L(X_s^x)ds}\right). \text{ Therefore}$$

$$E^{DMC}(t) \stackrel{\text{def}}{=} \frac{\langle h(t, .), \frac{H\psi_I}{\psi_I} \rangle}{\langle h(t, .), 1 \rangle} = \frac{\langle \chi(t, .), H\psi_I \rangle}{\langle \chi(t, .), \psi_I \rangle}$$
where  $\chi(t, x) = \psi_I(x) \mathbb{E}\left(e^{-\int_0^t E_L(X_s^x)ds}\right)$  vanishes on  $\psi_I^{-1}(0).$ 
As  $E(t) = \frac{\langle e^{-iH}\psi_I, H\psi_I \rangle}{\langle e^{-iH}\psi_I, \psi_I \rangle}$  and  $e^{-tH}\psi_I$  does not vanish on  $\psi_I^{-1}(0)$ , in general
$$E^{DMC}(t) \neq E(t).$$



### Elements of proof

Step 3 : For each connected component C of  $\mathbb{R}^{3N} \setminus \psi_I^{-1}(0)$ ,

$$x \in \mathcal{C} \implies \forall t \ge 0, \ X_t^x \in \mathcal{C}$$

Consequence : the restriction of  $\chi(t, x)$  to  $\mathbb{R}_+ \times \mathcal{C}$  unique solution of

$$\partial_t u = -Hu = \frac{1}{2}\Delta u - Vu, \ u(0,x) = 1_{\mathcal{C}}(x)\psi_I(x), \ u = 0 \text{ on } \partial \mathcal{C}.$$

Let

$$E_{\mathcal{C}} = \inf\{\langle \psi, H\psi \rangle_{\mathcal{C}} \colon \psi \in H^1_0(\mathcal{C}), \|\psi\|_{L^2(\mathcal{C})} = 1\}.$$

 $E_{\mathcal{C}}$  is attained for some  $\psi_{\mathcal{C}}$  positive on  $\mathcal{C}$ .

$$E^{DMC}(t) = \frac{\sum_{\mathcal{C}} < H\psi_{I}, \chi(t, .) >_{L^{2}(\mathcal{C})}}{\sum_{\mathcal{C}} < \psi_{I}, \chi(t, .) >_{L^{2}(\mathcal{C})}} \overset{t \to +\infty}{\sim} \frac{\sum_{\mathcal{C}} \mathbf{E}_{\mathcal{C}} < \psi_{I}, \psi_{\mathcal{C}} >_{L^{2}(\mathcal{C})}^{2} e^{-\mathbf{E}_{\mathcal{C}}t}}{\sum_{\mathcal{C}} < \psi_{I}, \psi_{\mathcal{C}} >_{L^{2}(\mathcal{C})}^{2} e^{-\mathbf{E}_{\mathcal{C}}t}}$$

As  $\forall \mathcal{C}, \langle \psi_I, \psi_{\mathcal{C}} \rangle^2_{L^2(\mathcal{C})} \neq 0$ ,  $\lim_{t \to +\infty} E^{DMC}(t) = \min_{\mathcal{C}} E_{\mathcal{C}}$ .



### Elements of proof Step 4: $\min_{\mathcal{C}} E_{\mathcal{C}} = E_0^{DMC}$

- $\leq$  : minimization component by component  $\leq$  global minimization
- $\geq : \text{let } \psi_{\mathcal{C}_0} \in H_0^1(\mathcal{C}_0) \text{ with } \|\psi_{\mathcal{C}_0}\|_{L^2(\mathcal{C}_0)} = 1 \text{ be such that} \\ < \psi_{\mathcal{C}_0}, H\psi_{\mathcal{C}_0} >_{\mathcal{C}_0} = \min_{\mathcal{C}} E_{\mathcal{C}}. \\ \text{Since } \mathcal{C}_0 \text{ connected component of } \mathbb{R}^{3N} \setminus \psi_I^{-1}(0) \text{ with } \psi_I \\ \text{ antisymmetric, one can extend } \psi_{\mathcal{C}_0} \text{ into a function } \psi \\ \text{ antisymmetric on } \mathbb{R}^{3N} \text{ such that} \end{cases}$

$$\begin{cases} \psi_{I}^{-1}(0) \subset \psi^{-1}(0) \\ < \psi, H\psi > = \|\psi\|_{2}^{2} < \psi_{\mathcal{C}_{0}}, H\psi_{\mathcal{C}_{0}} >_{\mathcal{C}_{0}} \end{cases}$$

#### **Remark 4**

If  $\psi_I$  satisfies the tiling property, one can check that for any connected component C of  $\mathbb{R}^{3N} \setminus \psi_I^{-1}(0)$ ,  $E_C = E_0^{DMC}$ . Working with a single component is enough to compute  $E_0^{DMC}$ .





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Numerical implementation



### Computation of $E^{DMC}(t)$ in practice

Use of particles (called walkers by physicists). Evolution of the position  $\tilde{X}_t^i$  of the *i*-th particle at time *t* by

• a time-discretization of  $dX_t^i = dW_t^i + b(X_t^i)dt$ . In general, Euler scheme :

$$\bar{X}_{t+\delta t}^{i} = \bar{X}_{t}^{i} + W_{t+\delta t}^{i} - W_{t}^{i} + b(\bar{X}_{t}^{i})\delta t,$$

- rejection of the new position until  $\psi_I(\bar{X}_{t+\delta t}^i)\psi_I(\bar{X}_t^i) > 0$  (otherwise particle *i* crosses a nodal surface of  $\psi_I$  between *t* and  $t + \delta t$ ). This does not prevent multiple crossings,
- use of a Metropolis-Hastings acceptation/rejection step to ensure that the distribution of  $\tilde{X}_{t+\delta t}^i$  is close to  $\psi_I^2(x)dx \rightarrow$  improves the integrability of  $E_L(\tilde{X}_{t+\delta t}^i) = \frac{H\psi_I}{\psi_I}(\tilde{X}_{t+\delta t}^i)$ . Introduction of an effective time to handle the possibility of staying at the same place in the approximation of the exponential weight  $e^{-\int_0^t E_L(X_s)ds}$ (*Umrigar, Nightingale, Runge J. Chem. Phys.* 1993).



### Computation of $E^{DMC}(t)$ in practice

$$E^{DMC}(t) = \frac{\mathbb{E}\left(E_L(X_t)e^{-\int_0^t E_L(X_s)ds}\right)}{\mathbb{E}\left(e^{-\int_0^t E_L(X_s)ds}\right)}$$

In order to control the variance, replication of the particles with high exponential weight and killing of the particles with low weight.

- In general, the number of particles is not preserved during the replication/killing steps,
- Also implementations with constant number of walkers (*Assaraf, Caffarel, Khelif Phys. Rev. E* 2000).



### Numerical analysis in a simplified case

- dimension :  $3N \rightarrow 1$ ,
- antisymmetry  $\rightarrow$  oddness,
- $H = -\frac{1}{2}\frac{d^2}{dx^2} + \frac{\alpha^2 x^2}{2} + \theta x^4$  with  $\alpha, \theta > 0$
- $\psi_I = \sqrt{2\alpha} \left(\frac{\alpha}{\pi}\right)^{\frac{1}{4}} x e^{-\frac{\alpha^2 x^2}{2}}$  odd ground state of  $H_0 = -\frac{1}{2} \frac{d^2}{dx^2} + \frac{\alpha^2}{2} x^2$ (energy :  $\frac{3\alpha}{2}$ )
- $b = \frac{\psi_I'}{\psi_I} = \frac{1}{x} \alpha x$
- $E_L = \frac{H\psi_I}{\psi_I} = \frac{H_0\psi_I}{\psi_I} + \theta x^4 = \frac{3\alpha}{2} + \theta x^4$
- For the SDE  $dX_t = dW_t + \left(\frac{1}{X_t} \alpha X_t\right) dt$ , possibility to simulate according to the conditional law of  $X_{s+r}$  given  $X_s$  and to the reversible measure  $21_{\{x>0\}} \psi_l^2(x) dx$  (initialization).

Simplification : even if *b* is explosive at 0, no simultaneous explosion of  $E_L$ . One-dimensional model  $\rightarrow$  does not take into account explosions at points where two particles coincide.

Numerical implementation



### Numerical analysis

For 
$$E_L = \frac{3\alpha}{2} + \theta x^4$$
, computation of  
 $E^{DMC}(T) = \frac{\mathbb{E}\left(E_L(X_T)e^{-\int_0^T E_L(X_s)ds}\right)}{\mathbb{E}\left(e^{-\int_0^T E_L(X_s)ds}\right)} = \frac{3\alpha}{2} + \theta \underbrace{\frac{\mathbb{E}\left(X_T^4 e^{-\theta \int_0^T X_s^4 ds}\right)}{\mathbb{E}\left(e^{-\theta \int_0^T X_s^4 ds}\right)}}_{E^D(T)}$ 

 $E_{\kappa}^{D}(T)$ 

Time discretization of the integral:  $K \in \mathbb{N}^*$  steps

## Lemma 5 For $K \in \mathbb{N}^*$ , $\left| E^D(T) - \underbrace{\frac{\mathbb{E}\left( X_T^4 e^{-\frac{\theta T}{K} \sum_{k=1}^K X_{kT/K}^4 \right)}}{\mathbb{E}\left( e^{-\frac{\theta T}{K} \sum_{k=1}^K X_{kT/K}^4 \right)}} \right| \le \frac{C_T}{K}.$

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### Particle approximation

- $(X_0^i)_{1 \le i \le N}$  i.i.d. according to  $2\psi_I^2(x) \mathbb{1}_{\{x>0\}} dx$
- To control variance,  $\nu$  resampling of the particles over [0, T]:  $l = \frac{K}{\nu}$  number of discretization steps and  $\Delta t = \frac{T}{\nu}$  time between two resampling.

For  $0 \le n \le \nu - 1$ , to obtain  $(X^i_{(n+1)\Delta t})_{1 \le i \le N}$  from  $(X^i_{n\Delta t})_{1 \le i \le N}$ 

- Mutation step :simulation of  $\xi_n = ((X_{n\Delta t + \frac{T}{K}}^i, \dots, X_{n\Delta t + \frac{T}{K}}^i)_{1 \le i \le N})$ according to SDEs driven by independent Brownian motions
- Selection step : conditionally to the result, generation of  $\overline{(X_{(n+1)\Delta t}^i)_{1 \le i \le N}}$  indep. and s.t. the following consistency equality holds

$$\mathbb{E}\left(\frac{1}{N}\sum_{i=1}^{N}\delta_{X_{(n+1)\Delta t}^{i}}\bigg|\xi_{n}\right) = \sum_{j=1}^{N}\rho^{j}\delta_{X_{n\Delta t+\frac{T}{K}}^{j}} \text{ with }\rho^{j} = \frac{e^{-\frac{\theta T}{K}\sum_{k=1}^{l}(X_{n\Delta t+\frac{kT}{K}}^{j})^{4}}}{\sum_{i=1}^{N}e^{-\frac{\theta T}{K}\sum_{k=1}^{l}(X_{n\Delta t+\frac{kT}{K}}^{i})^{4}}}$$



### Selection step

Resampling methods which satisfy the assumption Multinomial sampling :  $(X^i_{(n+1)\Delta t})_{1 \le i \le N}$  conditionally i.i.d. according to  $\sum_{j=1}^{N} \rho^j \delta_{X^j_{n\Delta t + \frac{lT}{K}}}$ 

Residual sampling : Let  $a^j = \lfloor N\rho^j \rfloor$ . Choose  $X^i_{(n+1)\Delta t} = X^j_{n\Delta t + \frac{T}{r}}$  for

 $1 + \sum_{m=1}^{j-1} a^m \leq i \leq \sum_{m=1}^{j} a^m \text{ and choose the remaining} \\ N - \sum_{m=1}^{N} a^m \text{ positions i.i.d. according to} \\ \sum_{j=1}^{N} \frac{N\rho^{j} - a^{j}}{N - \sum_{m=1}^{N} a^m} \delta_{X_{n\Delta t + \frac{lT}{K}}^{j}}.$ 

Stratified sampling : for  $1 \le i \le N$  let  $U^i$  be i.i.d.  $\sim \mathcal{U}[0, 1]$  and

$$X_{(n+1)\Delta t}^{i} = \sum_{j=1}^{N} \mathbb{1}_{\{\sum_{m=1}^{j-1} \rho^{m} \le \frac{i-U^{i}}{N} \le \sum_{m=1}^{j} \rho^{m}\}} X_{n\Delta t + \frac{lT}{K}}^{j}$$

Stratified reminder sampling : First step in residual sampling + stratified instead of multinomial sampling in the 2<sup>nd</sup>



### Particle approximation

#### **Theorem 6**

$$\mathbb{E}\left|E^{D}(T) - \frac{1}{N}\sum_{i=1}^{N}(X^{i}_{\nu\Delta t})^{4}\right| \leq \frac{C_{\nu}}{\sqrt{N}} + \frac{C_{T}}{K}.$$

(El Makrini, Jourdain, Lelièvre M2AN 2007).

- main contribution w.r.t. *Del Moral et al*:  $C_{\nu}$  does not depend on *K*,  $E_L$  unbounded
- for fixed  $\nu$  and T, convergence as  $N, K \rightarrow +\infty$
- optimal number  $\nu$  of resampling?

- Numerical implementation





Expectation of the absolute value of the error (reference energy computed by a spectral method) w.r.t.

- the number K of discretization steps (N = 5000,  $\nu = 30$  multinomial resampling,  $\theta = 2, T = 5$ , 300 independent realizations)
- the number N of particles ( $\nu = 50$  multinomial resampling,  $\theta = 0.5$ , T = 5, K = 1000, 2000 independent realizations).

Exact simulation of the SDE (dotted curves)/ use of a discretization scheme proposed by Alfonsi, MCMA 2006 (solid curves).



### Comparison of the resampling methods



Evolution of the variance (computed over 200 indep. simulations) with time : N = 1000, T = 5, K = 1000,  $\nu = 20$ ,  $\theta = 2$ .

Numerical analysis of the DMC method in a simple case .

- Numerical implementation







Expectation of the absolute value of the error (reference energy computed by a spectral method) w.r.t. the number  $\nu$  of selection steps (multinomial resampling, N = 5000, T = 5, K = 1000,  $\theta = 2$ , 300 independent realizations)

 $\nu_* = 25$  optimal !



### Choice of the number $\nu$ of selection steps



Evolution of the variance without resampling with time *T* (same parameters) Minimal for  $t_* \sim 0.25$ .  $T/t_* = 20$  close to  $\nu_* = 25$ . Suggests to compute  $t_*$  (without resampling, variance easily

estimated over a few independent particles) and choose  $\nu = T/t_*$ .