

Higher-order Time Integration of Stochastic Differential Equations and Application to Coulomb Collisions

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Main results

- We have developed a higher (Milstein)-order Coulomb-Langevin scheme
 - ▶ improved convergence demonstrated
 - ▶ correct mean behavior demonstrated
- A new approach was needed
 - ▶ existing approach does not extend easily to higher order
- New method developed for sampling area integral terms
 - ▶ simple, accurate, efficient

Coulomb collisions are important in many plasma applications

- Any sufficiently dense plasma
 - ▶ Magnetic fusion (MFE), inertial fusion (ICF), plasma processing, near-earth (or planetary) space plasma
- Long history of study of Coulomb collisions in plasmas
 - ▶ Analytical results
 - ★ Landau '36-7; Rosenbluth et. al.; '57, Trubnikov'65
 - ▶ Monte-Carlo (SDE) methods
 - ★ Langevin (+ field-term) methods - of interest in the present work: Manheimer et. al., '97; Lemons et. al., '09; Cohen et. al., '10
 - ★ Binary-collision methods - used in our hybrid work: Takizuke and Abe '77; Nanbu '97; Dimits et. al., '09
 - ▶ Continuum (PDE) methods
 - ★ e.g., Xiong, et. al., '08; Abel et. al., '08

Coulomb collisions are long-range, unlike neutral-atomic/molecular collisions

- Dominated by many small-angle scattering “events”
 - ▶ large-angle scattering events are subdominant
- Appropriate description is a Fokker-Planck (forward Kolmogorov) equation (Landau, 1936/7 - not a Boltzmann equation):

$$\left. \frac{\partial f_\alpha}{\partial t} \right|_{\text{coll}} = \frac{\partial}{\partial \mathbf{v}} \cdot \left[\pi q_\alpha^2 L \sum_\beta q_\beta^2 \int d\tau' \left(f_\alpha \frac{\partial f'_\beta}{\partial \mathbf{v}'} - f'_\beta \frac{\partial f_\alpha}{\partial \mathbf{v}} \right) \frac{(u^2 \mathbf{I} - \mathbf{u}\mathbf{u})}{u^3} \right]$$

The Milstein method is the first in a hierarchy of higher-order methods for SDE's

$$\delta Y_{n,j}^i = a^i(t_{n,j}, \mathbf{Y}_{n,j})\delta t + b^i(t_{n,j}, \mathbf{Y}_{n,j})\delta W_{n,j}^i$$

- $t_{n,j} = t_n + j\delta t$, $t_n = t_0 + n\Delta t$, $\Delta t = N\delta t$
- $\delta W_{n,j}^i$ are independent normal random numbers with variance δt .
- $\Delta W \equiv W(t_{n+1}) - W(t_n) = \lim_{N \rightarrow \infty} \sum_{j=1}^N \delta W_{n,j}$ - Wiener increment
- First-order (in Δt) approximation to $\Delta \mathbf{Y}_N \equiv \lim_{N \rightarrow \infty} \sum_{k=1}^N \delta \mathbf{Y}_k$

$$\begin{aligned} \Delta Y_n^i &= a^i(t_n, \mathbf{Y}_n)\Delta t + b^i(t_n, \mathbf{Y}_n)\Delta W_n^i \\ &+ b_{,j}^i(t_n, \mathbf{Y}_n)b^j(t_n, \mathbf{Y}_n) \int_0^{\Delta t} dW^i(t_n + s) \int_0^s dW^j(t_n + \eta) \\ &+ \begin{cases} O(\Delta t^{3/2}) & - \text{strong} \\ O(\Delta t^2) & - \text{weak} \end{cases} \rightarrow \begin{cases} O(T\Delta t) & - \text{strong} \\ O(T\Delta t) & - \text{weak} \end{cases} \end{aligned}$$

The Milstein method is of interest because it is the first in a hierarchy of higher-order methods for SDE's

- This hierarchy includes schemes with improved (higher-order) weak convergence
- Significantly improves efficiency of multi-(time-)level schemes (Giles '07), which have lower computational complexity for a given overall error than single-level Monte-Carlo schemes.

Higher-order methods for SDE's have been applied in a variety of fields

- Finance
- Chemical Physics
- See, e.g., Kloeden and Platen. '92
- Almost all published Monte-Carlo treatments of Coulomb collisions have used the low-order Euler-Maruyama method.
 - ▶ One exception: Lemons et. al., '09
 - ★ Added higher order (Milstein) term for v , but
 - ★ Did not do tests that might have shown a difference
 - ★ Did not include higher order terms for angular scattering

Approach 1: Apply collisional drag and scattering in a frame aligned with particle velocity

- Manheimer et al, '97; Lemons et. al., '09; Cohen et. al., '10
- Basic underlying equation:

$$\begin{aligned}d\mathbf{v}(t) &= [F_d(v) dt + Q_{||}(v) dW^{||}(t)] \hat{\mathbf{v}}(t) + Q_{\perp}(v) d\mathbf{W}(t), \\d\mathbf{W}^{\perp}(t) &= dW^x(t) \hat{\mathbf{x}}(t) + dW^y(t) \hat{\mathbf{y}}(t).\end{aligned}$$

- Here

- ▶ interpret in Ito sense
- ▶ $\Delta\mathbf{v}(t) \equiv \int_t^{t+\Delta t} d\mathbf{v}(t)$
- ▶ $(\hat{\mathbf{x}}, \hat{\mathbf{y}}, \hat{\mathbf{v}})$ - frame aligned with \mathbf{v}

- ★ e.g., $\hat{\mathbf{x}} = \hat{\mathbf{y}}_0 \times \hat{\mathbf{v}} / |\hat{\mathbf{y}}_0 \times \hat{\mathbf{v}}|, \hat{\mathbf{y}} = \hat{\mathbf{v}} \times \hat{\mathbf{x}}$

Milstein-order velocity step for Approach 1

- First-order accurate (in Δt) approximation to $\Delta \mathbf{v}(t) \equiv \int_t^{t+\Delta t} d\mathbf{v}(t)$

$$\begin{aligned}\Delta \mathbf{v} &= Q_{\parallel 0} \Delta t^{1/2} \Delta W^{\parallel} \hat{\mathbf{v}}_0 + Q_{\perp 0} \Delta t^{1/2} (\Delta W^x \hat{\mathbf{x}}_0 + \Delta W^y \hat{\mathbf{y}}_0) \\ &+ \left\{ \Delta t F_d(v_0) + \frac{1}{2} Q_{\parallel 0} Q'_{\parallel 0} \Delta t \left([\Delta W^{\parallel}]^2 - 1 \right) \right. \\ &- \left. \frac{Q_{\perp 0}^2}{2v_0} \Delta t \left[\left([\Delta W^x]^2 - 1 \right) + \left([\Delta W^y]^2 - 1 \right) \right] \right\} \hat{\mathbf{v}}_0 \\ &+ Q_{\parallel 0} Q'_{\perp 0} \Delta t [A^{x\parallel} \hat{\mathbf{x}}_0 + A^{y\parallel} \hat{\mathbf{y}}_0] + O(\Delta t^{3/2})\end{aligned}$$

- After applying $\Delta \mathbf{v}$, to get new $\mathbf{v}_{0\text{new}} \equiv \mathbf{v}_{0\text{old}} + \Delta \mathbf{v}$, apply next $\Delta \mathbf{v}$ using a frame aligned with $\mathbf{v}_{0\text{new}}$

Several possible choices for other unit vectors

- 1 Second vector along line of constant longitude in lab frame

$$\begin{aligned}\hat{\mathbf{x}}_0 &= \hat{\boldsymbol{\theta}}_{\text{lab}} \\ \hat{\mathbf{y}}_0 &= \hat{\mathbf{v}}_0 \times \hat{\mathbf{x}}_0\end{aligned}$$

- 2 Second vector orthogonal to fixed plane

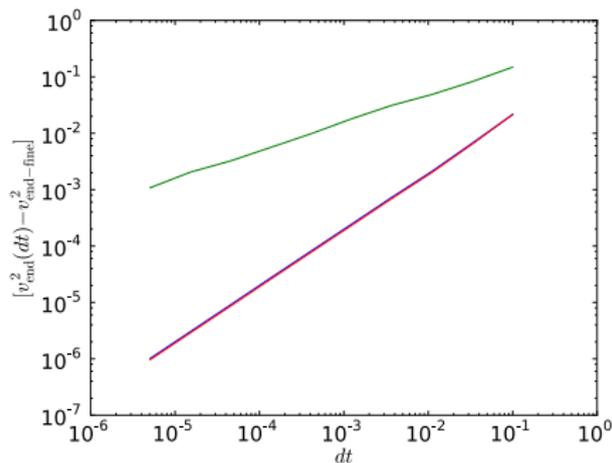
$$\begin{aligned}\hat{\mathbf{x}}_0 &= \hat{\mathbf{y}}_{\text{lab}} \times \hat{\mathbf{v}}_0 / |\hat{\mathbf{y}}_{\text{lab}} \times \hat{\mathbf{v}}_0| \\ \hat{\mathbf{y}}_0 &= \hat{\mathbf{v}}_0 \times \hat{\mathbf{x}}_0\end{aligned}$$

- 3 Rotate unit vector system as a rigid body about the single axis that gives the change in $\hat{\mathbf{v}}_0$

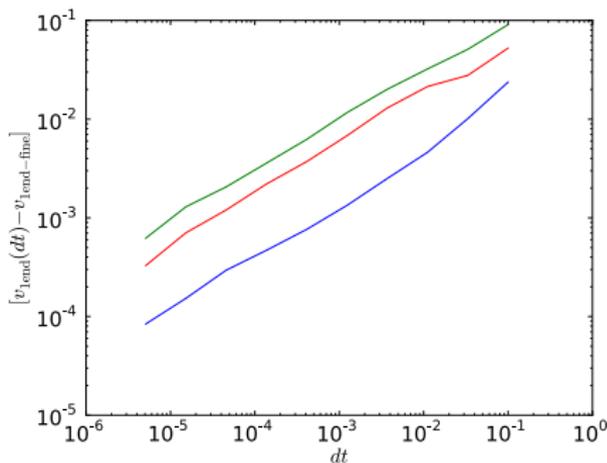
Approach 1 achieves $O(\Delta t)$ strong convergence for v , but not for angular component of the evolution

- 400 realizations; time step range = 3^{10} ; end time $\nu(v_{\text{th}}) t_{\text{end}} = 0.1$
- Green-Euler, red-Milstein fixed-plane, blue-Milstein rigid rot.

$$\left| v_{\text{end}}^2(dt) - v_{\text{end-fine}}^2 \right|$$



$$\left| v^x(dt) - v_{\text{fine}}^x \right|$$



(better) Approach 2: formulate whole problem as SDE's for spherical coordinates wrt a fixed (lab.) frame

- Coordinates: v , $\mu = \cos \theta$, ϕ ; θ = polar angle, ϕ = azimuthal angle
- From Rosenbluth et. al., '57,

$$\frac{1}{\Gamma_{tf}} \left(\frac{\partial f_t}{\partial t} \right)_c = -\frac{1}{v^2} \frac{\partial}{\partial v} \left[\left(v^2 \frac{\partial h}{\partial v} + \frac{\partial g}{\partial v} \right) f_t \right] + \frac{1}{2v^2} \frac{\partial^2}{\partial v^2} \left(v^2 \frac{\partial^2 g}{\partial v^2} f_t \right) + \frac{1}{2v^3} \frac{\partial g}{\partial v} \left\{ \frac{\partial}{\partial \mu} \left[(1 - \mu^2) \frac{\partial f_t}{\partial \mu} \right] + \frac{1}{(1 - \mu^2)} \frac{\partial^2 f_t}{\partial \phi^2} \right\}.$$

$$\Gamma_{tf} = \frac{4\pi q_t^2 q_f^2 \lambda}{m_t^2}.$$

- For a Maxwellian field-particle plasma, have analytical expressions for $g(v)$ and $h(v)$ (Trubnikov, '65).

Coulomb test-particle problem as SDE's for spherical coordinates wrt a fixed frame

- Write as Ito form drag-diffusion (forward Kolmogorov) equation:

$$\begin{aligned} \left(\frac{\partial \hat{f}_t}{\partial t} \right)_c &= -\frac{\partial}{\partial v} [F_d(v) \hat{f}_t] + \frac{\partial^2}{\partial v^2} [D_v(v) \hat{f}_t] + \frac{\partial}{\partial \mu} [2D_a(v)\mu \hat{f}_t] \\ &\quad + \frac{\partial^2}{\partial \mu^2} [D_a(v) (1 - \mu^2) \hat{f}_t] + \frac{\partial^2}{\partial \phi^2} \left[\frac{D_a(v)}{(1 - \mu^2)} \hat{f}_t \right], \end{aligned}$$

where $\hat{f}_t = 2\pi v^2 f_t$

- Corresponding Ito-Langevin equations:

$$\begin{aligned} dv(t) &= F_d(v) dt + \sqrt{2D_v(v)} dW_v(t), \\ d\mu(t) &= -2D_a(v)\mu dt + \sqrt{2D_a(v)(1 - \mu^2)} dW_\mu(t), \\ d\phi(t) &= \sqrt{\frac{2D_a(v)}{(1 - \mu^2)}} dW_\phi(t). \end{aligned}$$

Milstein scheme for Coulomb test-particle problem

$$\Delta v = F_{d0} \Delta t + \sqrt{2D_{v0}} \Delta W_v + \kappa_M D'_{v0} \frac{1}{2} (\Delta W_v^2 - \Delta t),$$

$$\begin{aligned} \Delta \mu &= -2D_{a0} \mu_0 \Delta t + \sqrt{2D_{a0} (1 - \mu_0^2)} \Delta W_\mu, \\ &+ \kappa_M \left[-2D_{a0} \mu_0 \frac{1}{2} (\Delta W_\mu^2 - \Delta t) + \sqrt{\frac{D_{v0}}{D_{a0}}} \sqrt{(1 - \mu_0^2)} D'_{a0} A_{v\mu} \right], \end{aligned}$$

$$\Delta \phi = \sqrt{\frac{2D_a(v)}{1 - \mu_0^2}} \Delta W_\phi + \kappa_M \left[\sqrt{\frac{D_{v0}}{D_{a0}}} \frac{D'_{a0}}{\sqrt{1 - \mu_0^2}} A_{v\phi} + \frac{2D_{a0} \mu_0}{1 - \mu_0^2} A_{\mu\phi} \right],$$

$$\Delta \psi = \psi(t_{i+1}) - \psi(t_i),$$

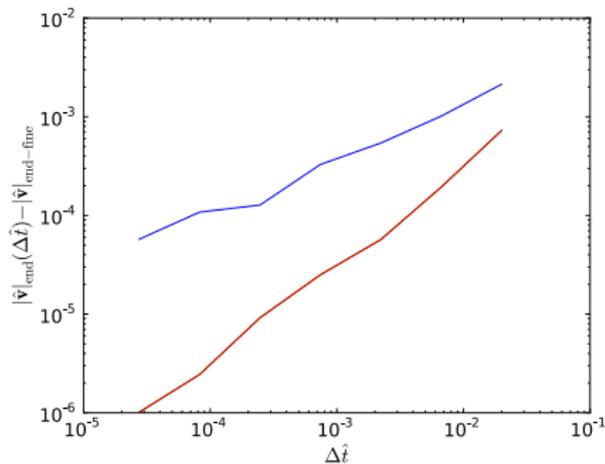
$$\psi_0 = \psi(t_i),$$

$$A_{kl} = \int_{t_i}^{t_{i+1}} dW_l(s) \int_{t_i}^s dW_k(\xi),$$

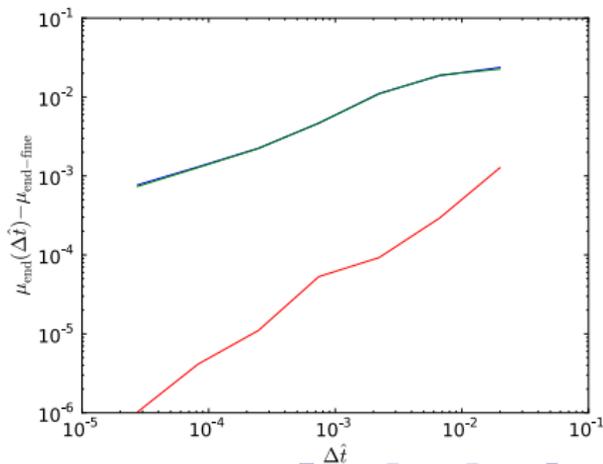
New approach achieves $O(\Delta t)$ strong convergence for v and for angular component

- v evolution unaffected by angular evolution, and \therefore by area terms
- Angular evolution has poor convergence without area terms
- 16 realizations; time step range = 3^8 ; end time $\nu(v_{th}) t_{end} = 0.1$
- Blue-Euler, Green-Milstein diagonal, Red-full Milstein

$$||v_{end}(\Delta t) - v_{end-fine}||$$



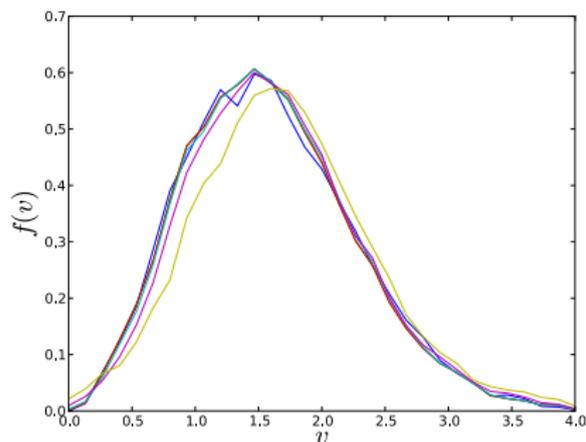
$$|\mu_{end}(\Delta t) - \mu_{end-fine}|$$



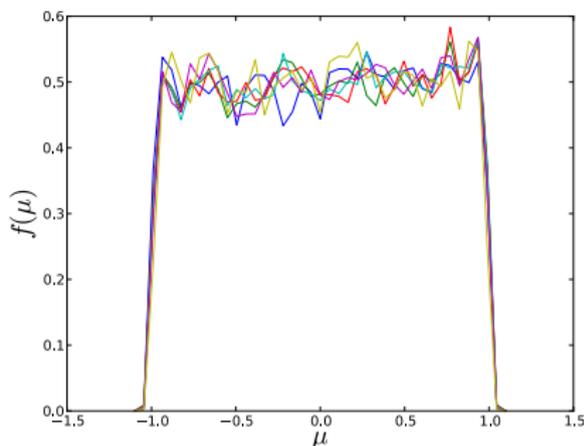
New approach gives correct dependences for velocity-space density functions (“distributions”)

- Blue - initial; other curves at t_{end} ; yellow - coarsest Δt
- 10000 particles; end time $\nu(v_{\text{th}}) t_{\text{end}} = 10$;
 $\Delta/t_{\text{end}} = 3^{-4}, 3^{-5}, 3^{-6}, 3^{-7}$

$f(v)$



$f(\mu)$



Theory and numerical implementations exist for the sampling of the stochastic integral terms

$$\int_0^{\Delta t} dW^i(t_n + s) \int_0^s dW^j(t_n + \eta) = \begin{cases} \frac{1}{2} \left[(\Delta W_n^i)^2 - \Delta t \right], & i = j \\ \frac{1}{2} \left[\Delta W_n^i \Delta W_n^j + L_n^{i,j} \right], & i \neq j \end{cases}$$

- Levy, '51

$$P_{cL} (L_n^{i,j} | \Delta W_n^i, \Delta W_n^j) = \hat{P}_{cL} (L_n^{i,j} | R_n^{i,j})$$

$$r_n^{i,j} = \sqrt{(\Delta W_n^i)^2 + (\Delta W_n^j)^2}$$

$$\begin{aligned} \phi_{cL} (k | R) &\equiv \langle \exp(-ikL) \rangle |_R \\ &= \frac{k/2}{\sinh(k/2)} \exp \left\{ \frac{R^2}{2} \left[1 - \frac{(k/2) \cosh(k/2)}{\sinh(k/2)} \right] \right\}. \end{aligned}$$

We have developed a simple accurate method for sampling area integrals

- Existing methods

- ▶ Interpolation from 2D table based on Levy's results (Gaines and Lyons '94)
 - ★ accurate and efficient
 - ★ somewhat involved
 - ★ challenging for conditional sampling - adaptive integration
- ▶ Discrete approximations (Clark and Cameron '80; Kloeden and Platen '92; Gaines and Lyons '97)
 - ★ simple to implement
 - ★ straightforward for adaptive integration
 - ★ expensive for good accuracy (many random numbers per L sample)

- Our method is a simplification of that of Gaines and Lyons '94

- ▶ based on an accurate approximation to Levy's PDF
- ▶ can implement with 1D tables or analytical functions
- ▶ can be used to significantly reduce memory and computation requirements for conditional sampling

Our approximation for the Levy-area PDF is based on approximate shape invariance of $P_{cL}(L|R)$

- Approximation to conditional PDF of L given R

$$P_{cL}(L|R) \approx P_{c-anL}(L|R) = s(R) P_{0L}(s(R)L)$$

$$P_{0L}(L) \equiv P_{cL}(L|R=0) = \frac{\pi}{2} \frac{1}{\cosh^2(L/2)} \quad - \quad \text{exact}$$

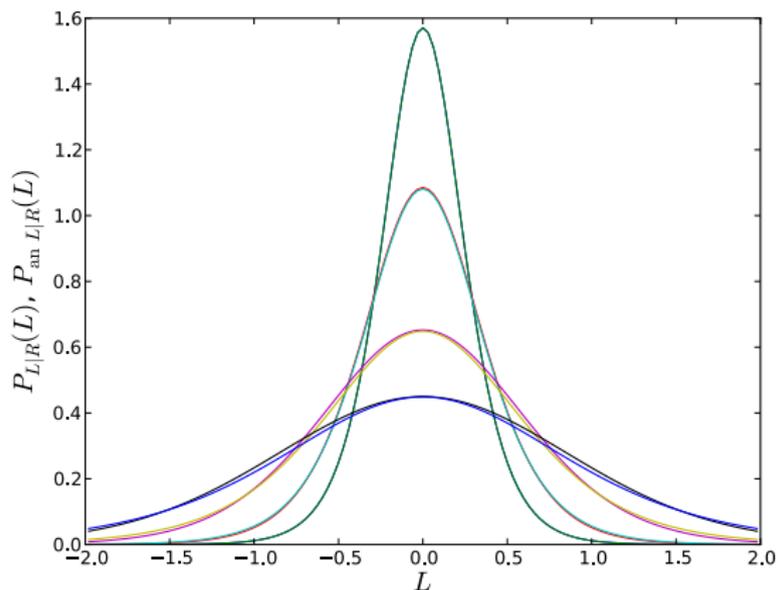
$$s(R) = P_{cL}(L=0|R)$$

- Can calculate $s(R)$ from 1D table or analytical fit
- Resulting algorithm for sampling L

$$L_R(R) = \frac{s(R)}{2\pi} \log\left(\frac{u}{1-u}\right).$$

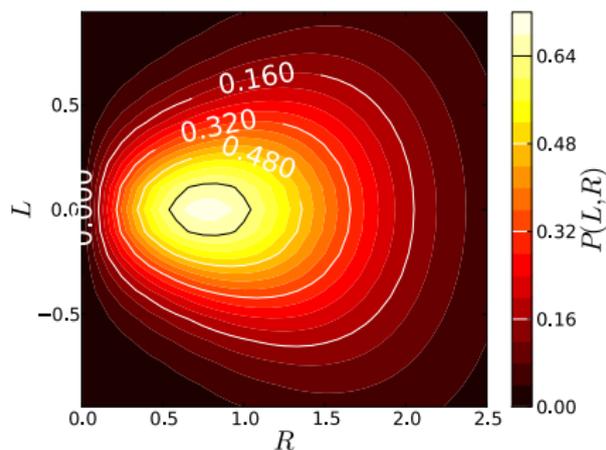
Our approximation for the Levy-area PDF is accurate to $\sim 1\%$

Exact and approximate conditional PDF's of L given R vs. L for $R = 0, 1, 2, 3$

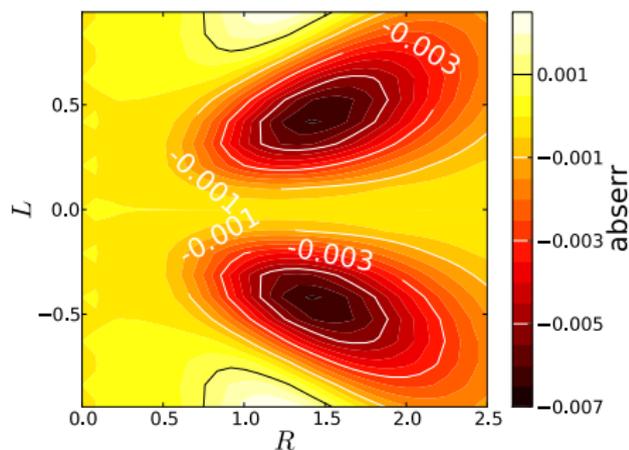


Our approximation for the Levy-area PDF is accurate to $\sim 1\%$

joint PDF of L and R



absolute error in $P(L, R)$



For strong convergence studies, Wiener increments and area integrals must be compounded

- Need to calculate trajectories representing a given underlying realization with different Δt
- Compounding is also needed for multilevel (Giles) schemes
- Compounding for Wiener increments: given $\delta_j W \equiv \int_{t_{j-1}}^{t_j} dW(s)$, where $t_j = t_{j-1} + \delta t$, and $\Delta t = n\delta t$

$$\Delta W \equiv \int_0^{\Delta t} dW(s) = \sum_{j=1}^n \delta_j W.$$

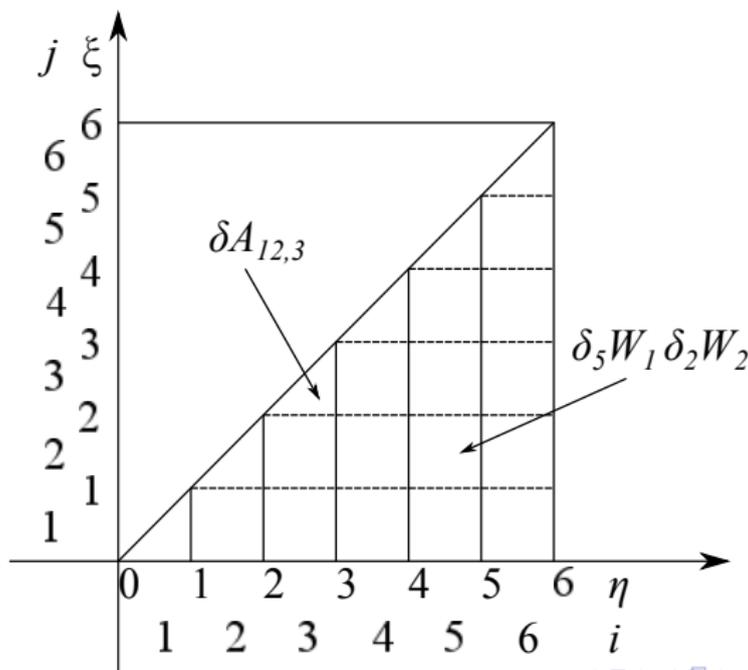
- Compounding area integrals:

$$\delta A_{12j} \equiv \int_{t_{j-1}}^{t_j} dW_1(\eta) \int_{t_{j-1}}^{\eta} dW_2(\xi),$$

$$\Delta A_{12} \equiv \int_0^{\Delta t} dW_1(\eta) \int_0^{\eta} dW_2(\xi),$$

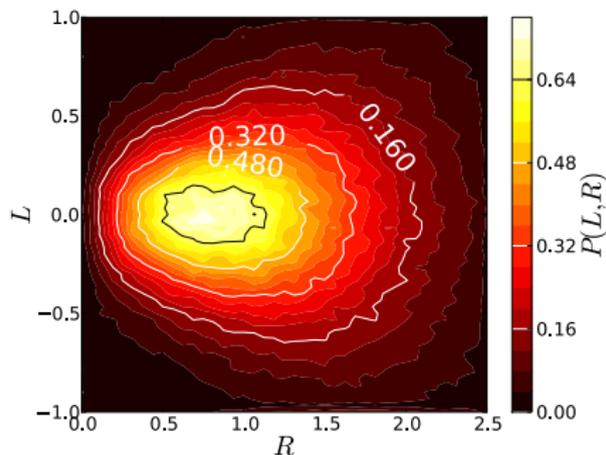
Compounding of area integrals

$$\Delta A_{12} = \sum_{i=2}^n \delta_i W_1 \left(\sum_{j=1}^{i-1} \delta_j W_2 + \delta A_{12,j} \right).$$

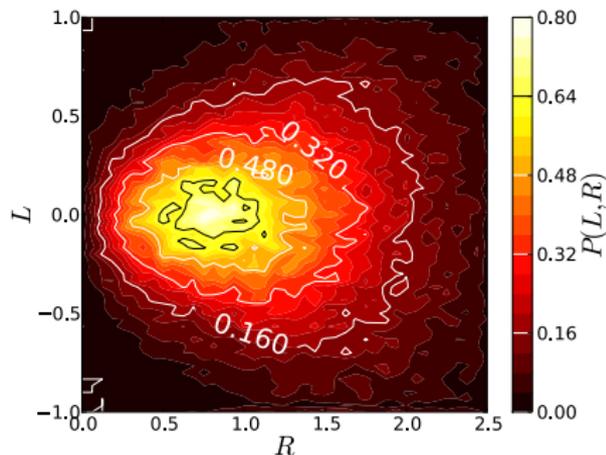


Our sampling and compounding algorithms and implementations work

PDF for 9×10^4 samples



compounded by factor of 5



- Strong scaling results for 2D Milstein (e.g., above collision results) provide a demonstration

Conditional sampling is needed for (time-) adaptive SDE integration

- Sample finer triplets $(\delta_j W_1, \delta_j W_2, \delta_j A_{12})$ given the coarser ones $(\Delta W_1, \Delta W_2, \Delta A_{12})$
- Reverse of compounding
- Existing methods are based on discrete representations
 - ▶ expensive because many (pseudo)random numbers needed per sample
- Direct conditional sampling can be done
 - ▶ construct $P_c(\delta L|\delta R, \Delta L, \Delta R)$ using Levy's result for $P_{cL}(L|R)$
 - ▶ store as 4D table
 - ▶ interpolate
- Our approximation $P_{c-anL}(L|R) = s(R) P_{0L}(s(R)L)$ reduces dimensionality of conditional sampling PDF to 3
 - ▶ much more manageable memory requirement

Summary

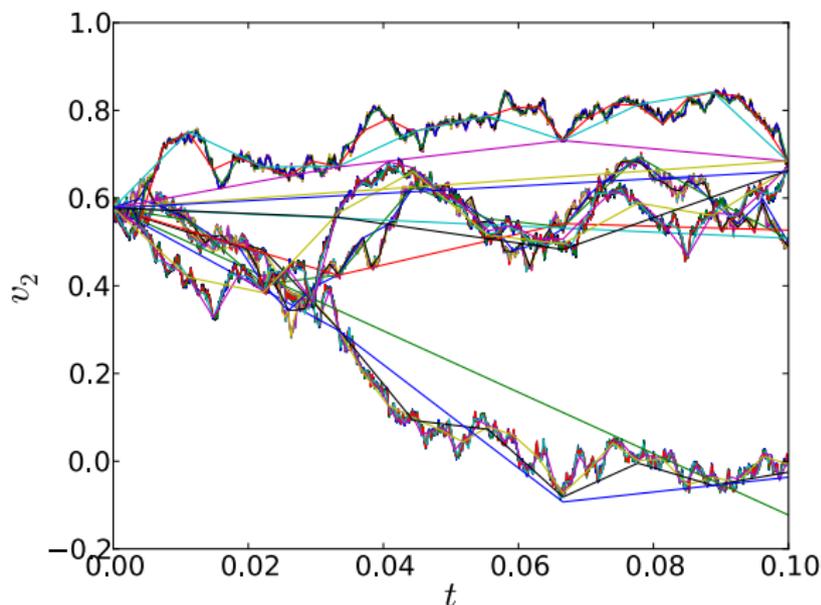
- We have developed a higher (Milstein)-order Coulomb-Langevin scheme
 - ▶ improved convergence demonstrated
 - ▶ correct mean behavior demonstrated
- A new approach was needed
 - ▶ existing approach does not extend easily to higher order
- New method developed for sampling area integral terms
 - ▶ simple, accurate, efficient
 - ▶ implemented (along with compounding)
- Future work in this direction
 - ▶ higher-order weak schemes
 - ▶ implement Giles' multilevel scheme
 - ▶ higher order adaptive SDE integrator

backup slides

Strong convergence results

- Convergence of trajectories (e.g., v at a given time) as $\Delta t \rightarrow 0$.

4 trajectories computed with different time steps



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