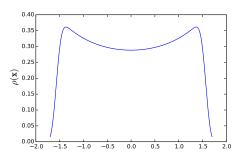
Pieter Van Nuffel

- ▶ 2006-2011: Master in Physics and Astronomy (UGent)
 - ► Master thesis: Simulation of Information Entropy in Financial Markets with Molecular Dynamics.
- ▶ 2012-2015: Teaching assistent (KU Leuven KULAK)
- 2015-current: PhD student
 - Promotor: Giovanni Samaey
 - Research topics and interests: Coarse Bifurcation Analysis of Stochastic Agent-based Network Models, Uncertainty Quantification, Multiscale Methods

Fokker-Planck equation

$$\frac{\partial \rho(x,t)}{\partial t} + \frac{\partial (a(x)\rho(x,t))}{\partial x} = D \frac{\partial^2 \rho(x,t)}{\partial x^2}$$



Simulate an ensemble of N particles evolving according to the corresponding SDE

$$\mathrm{d}x = a(x)\mathrm{d}t + \sqrt{2D}\cdot\mathrm{d}W_t.$$

$$\rho(t + n\Delta t) = \Phi_T(\rho) = (\mathcal{R} \circ \mathcal{E}(n\Delta t, \omega) \circ \mathcal{L}(\omega))(\rho(t))$$

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- ► Evolution E: Simulation of the SDE for N particles over n timesteps.

$$\mathbf{x}^{n+1} = \mathbf{x}^n + a(\mathbf{x}^n)\Delta t + \sqrt{2D\Delta t} \cdot \boldsymbol{\xi}^n$$

with

$$\xi^n(\omega) \sim \mathcal{N}(0,1)$$

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▶ Restriction $\mathcal{R}: \mathbf{x} \to \boldsymbol{\rho}$

$$\frac{1}{N}\sum_{i=1}^{N}w_i\cdot\chi_{\Delta_j}(x_i)=\rho_j$$



Evaluating Jacobian-vector products

$$oldsymbol{
ho}_* - oldsymbol{\Phi}_T(oldsymbol{
ho}_*) = 0$$

$$\mathbf{J}\mathbf{v} = D(oldsymbol{\Phi}_T) \cdot \mathbf{v} pprox rac{oldsymbol{\Phi}_T(oldsymbol{
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Problem: numerical noise for $\varepsilon \ll 1$

Repeating Φ_T with two sets of random numbers ω_1, ω_2 will give different results $\to \mathcal{O}(1/(\varepsilon^2 N))$ variance.

$$\begin{split} \mathsf{J} \mathsf{v} &= D(\Phi_{\mathcal{T}}) \cdot \mathsf{v} &\approx \frac{\Phi_{\mathcal{T}}(\rho + \varepsilon \mathsf{v}, \omega_1) - \Phi_{\mathcal{T}}(\rho, \omega_2)}{\varepsilon} \\ &\approx \frac{\Phi_{\mathcal{T}}(\rho, \omega_1) + \varepsilon D(\Phi_{\mathcal{T}})(\rho, \omega_1) \cdot \mathsf{v} - \Phi_{\mathcal{T}}(\rho, \omega_2)}{\varepsilon} \end{split}$$

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Solution:

Perturbations on the density \rightarrow perturbations in the weights

$$\frac{1}{N} \sum_{i=1}^{N} w_i \cdot \chi_{\Delta_j}(x_i) = \rho_j$$

$$\frac{1}{N} \sum_{i=1}^{N} w_{\varepsilon}^i \cdot \chi_{\Delta_j}(x^i) = \rho_j + \varepsilon v_j.$$

For the perturbated density: compute the weight per bin as $w^j_arepsilon=1+rac{arepsilon v_j}{
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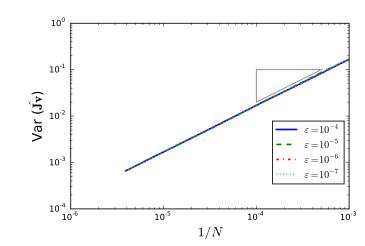
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$$\mathtt{Var}(\mathbf{\hat{Jv}}) = \hat{\mathbb{E}}\left[\left(\mathbf{\hat{Jv}} - \hat{\mathbb{E}}[\mathbf{\hat{Jv}}]\right)^2\right] \sim \mathcal{O}(1/\textit{N})$$



$$\mathsf{Bias}(\hat{\mathsf{Jv}},\mathsf{Jv}_{\mathit{FP}}) = \hat{\mathbb{E}}[\hat{\mathsf{Jv}}] - \mathsf{Jv}_{\mathit{FP}}$$

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and ho calculated by explicitly solving

$$\frac{\partial \rho(x,t)}{\partial t} + \frac{\partial (a(x)\rho(x,t))}{\partial x} = D \frac{\partial^2 \rho(x,t)}{\partial x^2}$$

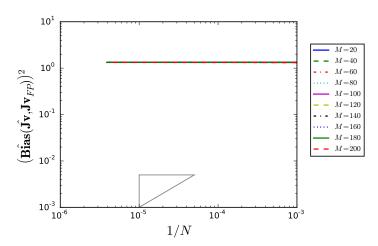
using

$$\rho_{i}^{n+1} = \rho_{i}^{n} + \Delta t \left(\frac{D}{\Delta x^{2}} \left(\rho_{i+1}^{n} - 2\rho_{i}^{n} + \rho_{i-1}^{n} \right) - \frac{a(x)}{\Delta x} (\rho_{i}^{n} - \rho_{i-1}^{n}) \right)$$

for the value of ρ at position $x = i\Delta x$ and time $t = (n+1)\Delta t$



$$\mathsf{Bias}(\mathsf{J}\mathsf{v},\mathsf{J}\mathsf{v}_\mathit{FP}) = \hat{\mathbb{E}}[\mathsf{J}\mathsf{v}] - \mathsf{J}\mathsf{v}_\mathit{FP}$$



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