

A PDF method applied to local wind simulation

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Aim : propose a numerical method to improve the wind forecasting at small scales.

Joint work with

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- P. Drobinski, T. Salameh (Laboratoire de Météorologie Dynamique)
- E. Peirano (ADEME)



Development of the wind power resources in France :

in March 2006, the production was around 1 GW.
France plans to produce 10 GW in 2010.



Our objective :

To develop and evaluate a stochastic downscaling method to compute the wind field at small scale, near the surface, from a meteorological field of large scale ($> 10\text{-}50$ km), simulated by a meteorological model.

Geographical framework

The French part of the Mediterranean basin :
(Languedoc-Roussillon, Provence-Alpes-Côtes d'Azur, Rhône-Alpes) :

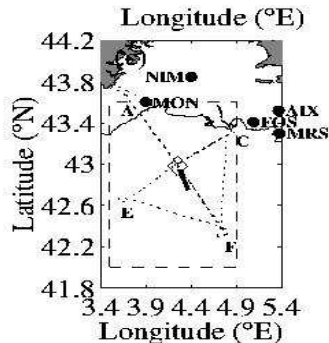
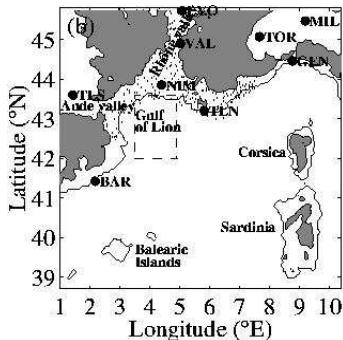


- First region in terms of production, with a high potential to develop.
- Mediterranean climate, mainly forced by the large scale climatic conditions, during the winter (November to March).
- Complex association between large scale and regional scale (10 – 100 km). Important role of orography, the ground and sea contrast.

Meteorological forcing of the Downscaling Stochastic Method (DSM)

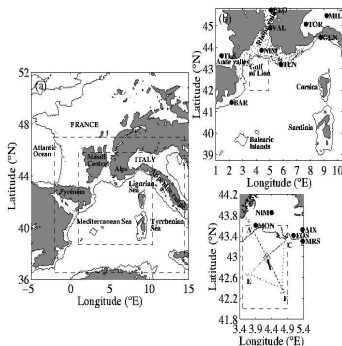
We used the numerical model *MM5* (mesoscale meteorological solver developed at the Pennsylvania State University/National Center for Atmospheric Research, USA)

Three computational model domains interactively imbricate:



Meteorological forcing of the (DSM)

MM5 solves the non-hydrostatic equations of motion in a terrain-following sigma coordinates.



Three computational model domains interactively imbricate are used. the horizontal mesh size is 27 km, 9 km and 3 km, respectively.

Coarse domain, medium domain and fine domain are centered at 43.7°N , 4.6°E and cover an area of $1350\text{ km} \times 1350\text{ km}$, $738\text{ km} \times 738\text{ km}$ and $120\text{ km} \times 174\text{ km}$, respectively.

The March 24, 1998 Mistral event :

MM5 Initial condition and boundary condition : taken from the ECMWF (European Centre for Medium Range Weather Forecast) reanalyses ERA-40. These reanalysis data are available every six hours on a $1^\circ \times 1^\circ$ latitude-longitude grid.

Validation of MM5 on this event (radiosoundings launched from Nîmes, on March 24, 1998) see Salameh, Drobinski et al *Atmos. Chem. Phys. Discuss* 2006

Forecasting solvers like *MM5* have a computational limit in both horizontal and vertical ranges. **Difficult (hopeless) to run *MM5* with an horizontal resolution finer than 1km.**

Our down-scaling approach

- Improve the resolution in a given sub-domain (local computations),
- by introducing a local model (compatible with the Navier Stokes equation),
- solved by a particle method which does not require any stability (CFL) condition.

Agenda

- The local model : Lagrangian modeling of turbulent flows.
- The numerical algorithm and related issues.
- Mathematical analysis of the (simplified) generalized (confined) Langevin model.
- Go back to the meteorological context.

Modeling of turbulent flows

Statistical approach of turbulent flows :

The properties of the fluid are all random fields.

The Reynolds averages (or ensemble averages) are expectations:

$$\langle \mathcal{U} \rangle(t, x) := \int_{\Omega} \mathcal{U}(t, x, \omega) d\mathbb{P}(\omega).$$

The corresponding Reynolds decomposition of the velocity is

$$\mathcal{U}(t, x, \omega) = \langle \mathcal{U} \rangle(t, x) + \mathbf{u}(t, x, \omega).$$

The random field $\mathbf{u}(t, x, \omega)$ is the turbulent part of the velocity.

Incompressible Navier Stokes equation in \mathbb{R}^3 , for the velocity field $\mathcal{U} = (\mathcal{U}^{(1)}, \mathcal{U}^{(2)}, \mathcal{U}^{(3)})$ and the pressure \mathcal{P} :

$$\partial_t \mathcal{U} + (\mathcal{U} \cdot \nabla) \mathcal{U} = \nu \Delta \mathcal{U} - \frac{1}{\rho} \nabla \mathcal{P}, \quad t > 0, \quad x = (x_1, x_2, x_3) \in \mathbb{R}^3,$$

$$\nabla \cdot \mathcal{U} = \partial_{x_i} \mathcal{U}_i = 0, \quad t \geq 0, \quad x \in \mathbb{R}^3, \text{ constant mass density } \rho$$

$$\mathcal{U}(0, x) = \mathcal{U}_0(x), \quad x \in \mathbb{R}^3.$$

The Reynolds averaged equation for the mean velocity

Assuming Reynolds decomposition, we obtain the unclosed equation

$$\partial_t \langle \mathcal{U}^{(i)} \rangle + \sum_{j=1}^3 \langle \mathcal{U}^{(j)} \rangle \partial_{x_j} \langle \mathcal{U}^{(i)} \rangle + \sum_{j=1}^3 \partial_{x_j} \langle \mathbf{u}^{(i)} \mathbf{u}^{(j)} \rangle = \nu \Delta \langle \mathcal{U}^{(i)} \rangle - \frac{1}{\rho} \partial_{x_i} \langle \mathcal{P} \rangle,$$

$\nabla \cdot \langle \mathcal{U} \rangle = 0$, $t \geq 0$, $x \in \mathbb{R}^3$, constant mass density

$\langle \mathcal{U} \rangle(0, x) = \langle \mathcal{U}_0 \rangle(x)$, $x \in \mathbb{R}^3$. $\langle \mathbf{u}^{(i)} \mathbf{u}^{(j)} \rangle = \langle \mathcal{U}^{(i)} \mathcal{U}^{(j)} \rangle - \langle \mathcal{U}^{(i)} \rangle \langle \mathcal{U}^{(j)} \rangle$.

One needs to model the equation of the Reynolds stress ($\langle \mathbf{u}^{(i)} \mathbf{u}^{(j)} \rangle$, i, j).

Direct modeling of the Reynolds stress :

the so-called k - ε turbulence models,

where

$$\text{kinetic turbulent energy } k(t, x) := \sum_{i=1}^3 \frac{1}{2} \langle \mathbf{u}^{(i)} \mathbf{u}^{(i)} \rangle(t, x)$$

and

$$\text{pseudo-dissipation } \varepsilon(t, x) := \nu \sum_{i=1}^3 \sum_{j=1}^3 \langle \partial_{x_j} \mathbf{u}^{(i)} \partial_{x_j} \mathbf{u}^{(i)} \rangle(t, x).$$

An alternative approach to compute the Reynolds stress

Let $f_E(V; t, x)$ be the probability density function (PDF) of the random field $\mathcal{U}(t, x)$, then

$$\langle \mathcal{U}^{(i)} \rangle(t, x) = \int_{\mathbb{R}^3} V^{(i)} f_E(V; t, x) dV,$$
$$\langle \mathcal{U}^{(i)} \mathcal{U}^{(j)} \rangle(t, x) = \int_{\mathbb{R}^3} V^{(i)} V^{(j)} f_E(V; t, x) dV.$$

The closure problem is reported on the PDE satisfied by the probability density function f_E .

In a series of papers (see e.g. Pope 85), Stephen B. Pope propose to model the p.d.f. f_E with a Lagrangian description of the flow.

The Lagrangian approach to compute the Reynolds stresses

Main ideas

- Describe the Lagrangian properties through a stochastic model with a state vector (X, U, ψ) which include
 - particle location,
 - particle velocity
 - scalar variables standing for any particles properties,and use a diffusion process to simulate its time rate of change.
- Those models are referred to as Langevin models.
- The associated SDE must be consistent with the macroscopic evolution of the fluid (in particular the averaged Navier-Stokes equation).

The Simplified Langevin model (Pope 94)

$$\left\{ \begin{array}{l} dX_t = U_t dt, \\ dU_t^{(i)} = \left[-\frac{1}{\rho} \frac{\partial \langle \mathcal{P} \rangle}{\partial x_i}(t, X_t) \right. \\ \quad \left. - \left(\frac{1}{2} + \frac{3}{4} C_0 \right) \frac{\varepsilon(t, X_t)}{k(t, X_t)} \left(U_t^{(i)} - \langle \mathcal{U}^{(i)} \rangle(t, X_t) \right) \right] dt \\ \quad + \sqrt{C_0 \varepsilon(t, X_t)} dW_t^{(i)}, \quad \forall i \in \{1, 2, 3\} \end{array} \right.$$

+ boundary conditions + wall boundary functions.

where $\varepsilon(t, x)$ and $k(t, x)$ are supposed to be known. $\langle \mathcal{P} \rangle(t, x)$ must be recovered by the Poisson equation

$$\nabla^2 \langle \mathcal{P} \rangle = - \frac{\partial^2 \langle \mathcal{U}^{(i)} \mathcal{U}^{(i)} \rangle}{\partial x_i \partial x_j}$$

which guarantees that the averaged Eulerian velocity is divergence free.

Compute the Reynolds averages $\langle \mathcal{U}^{(i)} \rangle(t, \mathbf{x})$ and $\langle \mathcal{U}^{(i)} \mathcal{U}^{(j)} \rangle(t, \mathbf{x})$

We call $f_L(V, \mathbf{x}; t)$ the probability density function of (U_t, X_t) .

Contrary to f_E , f_L satisfies a closed (nonlinear) PDE : the Fokker-Planck equation associated to the Langevin SDE.

In the case of incompressible flow with a constant mass density,

$$f_E(V; \mathbf{x}, t) = \frac{f_L(V, \mathbf{x}; t)}{\int_{\mathbb{R}^3} f_L(V, \mathbf{x}; t) dV},$$

and for any bounded measurable function $g(v)$,

$$\langle g(\mathcal{U}) \rangle(t, \mathbf{x}) = \mathbb{E} (g(U_t) / X_t = \mathbf{x}).$$

In particular,

$$\langle \mathcal{U}^{(i)} \rangle(t, \mathbf{x}) = \int_{\mathbb{R}} V^{(i)} \frac{f_L(V, \mathbf{x}; t)}{\int_{\mathbb{R}^3} f_L(U, \mathbf{x}; t) dU} dV = \mathbb{E} \left(U_t^{(i)} / X_t = \mathbf{x} \right).$$

The Basic model (Dreeben Pope 98)

Include the instantaneous turbulence frequency ω , satisfying

$$\left\{ \begin{array}{l} dX_t = U_t dt, \\ dU_t^{(i)} = \left[-\frac{1}{\rho} \frac{\partial \langle \mathcal{P} \rangle}{\partial x_i}(t, X_t) \right. \\ \quad \left. - \left(\frac{1}{2} + \frac{3}{4} C_0 \right) \langle \omega \rangle(t, X_t) \left(U_t^{(i)} - \langle \mathcal{U}^{(i)} \rangle(t, X_t) \right) \right] dt \\ \quad + \sqrt{C_0 k(t, X_t) \langle \omega \rangle(t, X_t)} dW_t^{(i)}, \quad \forall i \in \{1, 2, 3\} \\ \\ d\omega_t = -C_3 \langle \omega \rangle(t, X_t) (\omega_t - \langle \omega \rangle(t, X_t)) dt - S_\omega \langle \omega \rangle(t, X_t) \omega_t dt \\ \quad + \sqrt{2C_3 C_4 \langle \omega \rangle^2(t, X_t) \omega_t} dW_t^{(4)}. \end{array} \right.$$

where

$$S_\omega = C_{\omega 2} + C_{\omega 1} \frac{\langle \mathbf{u}^{(i)} \mathbf{u}^{(j)} \rangle(t, \mathbf{x})}{\varepsilon(t, \mathbf{x})} \frac{\partial \langle \mathcal{U}^{(i)} \rangle}{\partial x_j}(t, \mathbf{x}).$$

$\varepsilon(t, \mathbf{x})$ is recovered by the closure relation $\langle \omega \rangle(t, \mathbf{x}) = \frac{\varepsilon(t, \mathbf{x})}{k(t, \mathbf{x})}$.

Application to meteorology

Use the same parametrisation than MM5 :

$$\left\{ \begin{array}{l} dX_t = U_t dt, \\ dU_t^{(i)} = \left[-\frac{1}{\rho} \frac{\partial \langle \mathcal{P} \rangle}{\partial x_i}(t, X_t) \right. \\ \quad \left. - \left(\frac{1}{2} + \frac{3}{4} C_0 \right) \frac{\varepsilon(t, X_t)}{k(t, X_t)} \left(U_t^{(i)} - \langle \mathcal{U}^{(i)} \rangle(t, X_t) \right) \right] dt \\ \quad + \sqrt{C_0 \varepsilon(t, X_t)} dW_t^{(i)}, \quad \forall i \in \{1, 2, 3\} \end{array} \right.$$

- The kinetic turbulent energy $k(t, x) = \frac{1}{2} \langle \mathbf{u}^{(i)} \mathbf{u}^{(i)} \rangle(t, x)$ is computed inside the model.
- $\varepsilon(t, x)$ is recovered by the closure relation $\varepsilon(t, x) = \frac{C}{L} k^{3/2}(t, x)$.

Application to meteorology, with the MM5 forcing

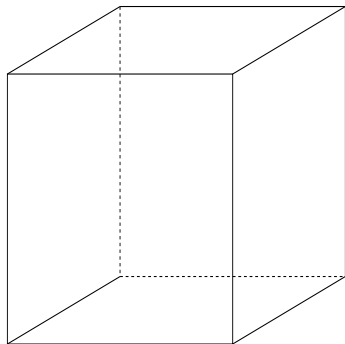
Let \mathbb{D} our local (space) computational domain

$$\left\{ \begin{array}{l} dX_t = U_t dt, \\ dU_t = \left[-\frac{1}{\rho} \nabla \langle \mathcal{P} \rangle (t, X_t) \right. \\ \quad \left. - \left(\frac{1}{2} + \frac{3}{4} C_0 \right) \frac{\varepsilon(t, X_t)}{k(t, X_t)} (U_t - \langle \mathcal{U} \rangle (t, X_t)) \right] dt \\ \quad + \sqrt{C_0 \varepsilon(t, X_t)} dW_t, \\ \quad - \sum_{0 \leq s \leq t} 2 (U_{s^-} \cdot n(X_s)) n(X_s) \mathbb{1}_{\{X_s \in \partial \mathbb{D}\}} \\ \quad - \sum_{0 \leq s \leq t} 2 (U_{s^-} \cdot n^\perp(X_s)) n^\perp(X_s) \mathbb{1}_{\{X_s \in \partial \mathbb{D}\}} \\ \quad + \sum_{0 \leq s \leq t} V_{MM5}(s, X_s) \mathbb{1}_{\{X_s \in \partial \mathbb{D}\}}. \end{array} \right.$$

The three last terms should guarantee

$$\langle \mathcal{U} \rangle (t, x) := \mathbb{E} [U_t / X_t = x] = V_{MM5}(t, x), \forall x \in \partial \mathbb{D}.$$

The numerical framework :



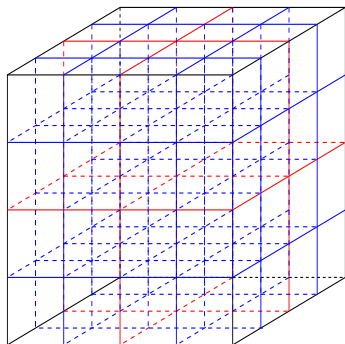
Our computational domain \mathbb{D} (for example, a given cell of the *MM5* solver).

Boundary condition :

$$\forall x \in \partial\mathbb{D}, \langle \mathcal{U} \rangle(t, x) = V_{MM5}(t, x)$$

(*MM5* guideline.)

The numerical framework : particle method

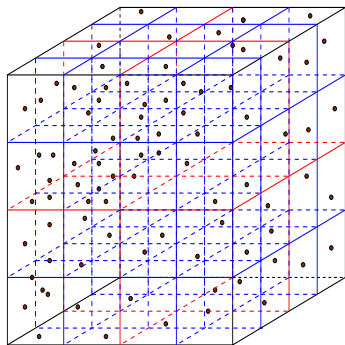


The computational space is divided in cells of given size.

Particle in cell (P.I.C.) technique to approximate the Eulerian fields like $\langle \mathcal{U}^{(i)} \rangle (t, x)$.

We compute the Eulerian fields (mean fields) at the center of each sub-cell only.

The numerical framework : particle method



- Introduce N .
- Constant mass density constraint \Rightarrow constant number of particles in each cell.
- Approximation of the conditional expectation :
if \mathcal{V}_x denotes the cell centered in x then the approximation of $\langle \mathcal{U}^{(i)} \rangle (t, x)$ is given by

$$\langle \mathcal{U}^{(i)} \rangle (t, x) \simeq \frac{C}{\text{Volume}(\mathcal{V}_x)} \left(\frac{1}{N} \sum_{l=1}^N U_t^{(i),l,N} \mathbb{1}_{\{X_t^{l,N} \in \mathcal{V}_x\}} \right).$$

$$\begin{aligned}
 \langle \mathcal{U}^{(l)} \rangle (t, \mathbf{x}) &= \mathbb{E} \left(U_t^{(l)} / X_t = \mathbf{s} \right) \\
 &\simeq \frac{\mathbb{E} \left[U_t^{(l)} \phi_\varepsilon(\mathbf{x}, X_t) \right]}{\mathbb{E} \left[\phi_\varepsilon(\mathbf{x}, X_t) \right]} \\
 &\simeq \frac{\frac{1}{N} \sum_{i=1}^N U_t^{(l), i, N} \phi_\varepsilon(\mathbf{x}, X_t^{i, N})}{\frac{1}{N} \sum_{i=1}^N \left[\phi_\varepsilon(\mathbf{x}, X_t^{i, N}) \right]}.
 \end{aligned}$$

When $\phi_\varepsilon(\mathbf{x}, y) = \mathbb{1}_{\{\mathbf{x} \text{ and } y \text{ are in the same sub-cell of size } \varepsilon\}}$,

$$\langle \mathcal{U}^{(i)} \rangle (t, \mathbf{x}) \simeq \frac{C}{\text{Volume}(\mathcal{V}_x)} \left(\frac{1}{N} \sum_{l=1}^N U_t^{(i), l, N} \mathbb{1}_{\{X_t^{l, N} \in \mathcal{V}_x\}} \right).$$

The numerical algorithm

The N -Particles dynamic : for $j = 1, \dots, N$

$$\left\{ \begin{array}{l} dX_t^{j,N} = U_t^{j,N} dt, \\ dU_t^{(i)j,N} = -\frac{1}{\rho} \frac{\partial \langle \mathcal{P} \rangle}{\partial x_i}(t, X_t^{j,N}) dt \\ \quad + D_U(t, X_t^{j,N}) dt + B_U(t, X_t^{j,N}) dW_t^{(i)j,N} \\ \quad + \text{MM5 guideline terms at the boundary, } \forall i \in \{1, 2, 3\} \\ \\ d\omega_t^{j,N} = D_\omega(t, X_t^{j,N}) dt + B_\omega(t, X_t^{j,N}) dW_t^{(4)j,N}. \end{array} \right.$$

The coefficients D_U , D_ω , B_U and B_ω depend on the particles approximations of $\langle \mathcal{U} \rangle$, $\langle \mathcal{U}^{(i)} \mathcal{U}^{(j)} \rangle$ and its derivatives, $\langle \omega \rangle$.

$-\frac{1}{\rho} \frac{\partial \langle \mathcal{P} \rangle}{\partial x_i}(t, X_t^{j,N})$ ensures that $\nabla \cdot \langle \mathcal{U} \rangle = 0$ and maintains the mass density constant.

A fractional step method : $n\Delta t \longrightarrow (n+1)\Delta t$ (Pope 85)

The N -Particles dynamic : for $j = 1, \dots, N$, for $n\Delta t \leq t \leq (n+1)\Delta t$,

$$\left\{ \begin{array}{l} d\tilde{X}_t^{j,N} = \tilde{U}_t^{j,N} dt, \\ d\tilde{U}_t^{(i),j,N} = -\frac{1}{\rho} \frac{\partial \langle \mathcal{P} \rangle}{\partial x_i}(t, \tilde{X}_t^{j,N}) dt \\ \quad + D_{\tilde{U}}(t, X_t^{j,N}) dt + B_{\tilde{U}}(t, X_t^{j,N}) dW_t^{(i),j,N} \\ \quad + \text{MM5 guideline terms at the boundary, } \forall i \in \{1, 2, 3\} \\ d\omega_t^{j,N} = D_\omega(t, X_t^{j,N}) dt + B_\omega(t, X_t^{j,N}) dW_t^{(4),j,N}. \\ X_{n\Delta t}^{j,N}, U_{n\Delta t}^{(i),j,N}, \omega_{n\Delta t}^{j,N} \text{ given.} \end{array} \right.$$

- Correction of the position of the particles $\tilde{X}_{(n+1)\Delta t}^{j,N} \longrightarrow X_{(n+1)\Delta t}^{j,N}$, in order to maintain the (discrete) uniform distribution.
- Correction of the particles velocity

$$\tilde{U}_{(n+1)\Delta t}^{j,N} \longrightarrow U_{(n+1)\Delta t}^{j,N}$$

such that $\nabla \cdot \langle \mathcal{U}^{(n+1)} \rangle = 0$.

Correction of the positions of the particles

Move the particles, such that the corresponding distribution becomes uniform.

Minimize the global amount of displacement.

The density $\rho(x)$ is an Eulerian quantity approximated thanks to the nearest grid point formula

$$\rho(x_i) = \frac{\#\{\text{particles in } C_i\}}{N_{ppc}}.$$

Can be viewed as a discretization of an optimal continuous transport problem (Brenier) :

Find a transport map $\phi : \mathbb{D} \rightarrow \mathbb{D}$, satisfying $\forall A \subset \mathbb{D}$

$$\int_{\phi^{-1}(A)} \rho(x) dx = \int_A \rho_0(x) dx$$

minimizing the L^2 -cost

$$K(\phi) = \int_{\mathbb{D}} |x - \phi(x)|^2 dx.$$

Correction of the positions of the particles

Well-known problem, having a well-know solution (see Benamou Brenier 2000 and ref. herein) : ϕ is unique and given by the Monge Ampère equation

$$\phi = \mathbf{1}_{\mathbb{D}} - \nabla\gamma$$

$$\text{with } \gamma \text{ satisfying } \rho(x) = \det \begin{pmatrix} 1 - \frac{\partial^2 \gamma}{\partial x_1^2} & -\frac{\partial^2 \gamma}{\partial x_1 \partial x_2} & -\frac{\partial^2 \gamma}{\partial x_1 \partial x_3} \\ -\frac{\partial^2 \gamma}{\partial x_1 \partial x_2} & 1 - \frac{\partial^2 \gamma}{\partial x_2^2} & -\frac{\partial^2 \gamma}{\partial x_2 \partial x_3} \\ -\frac{\partial^2 \gamma}{\partial x_1 \partial x_3} & -\frac{\partial^2 \gamma}{\partial x_2 \partial x_3} & 1 - \frac{\partial^2 \gamma}{\partial x_3^2} \end{pmatrix},$$

Numerical discretization : difficult

Pope 85 : neglect the nonlinear terms

$$\begin{cases} \Delta\gamma(x) = 1 - \rho(x), x \in \mathbb{D}, \\ \frac{\partial\gamma}{\partial n} \Big|_{\partial\mathbb{D}} = 0. \end{cases}$$

Eventually iterate the process until convergence.

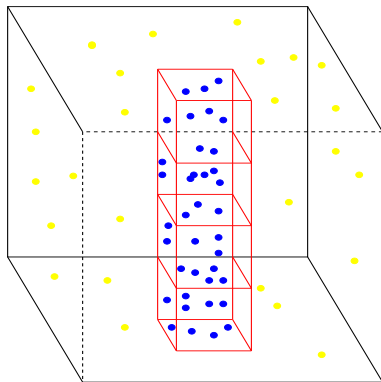
Correction of the positions of the particles : alternative strategy

Suppose $\mathbb{D} = (0, 1)$. The optimal transport is then entirely determined by the transfer condition :

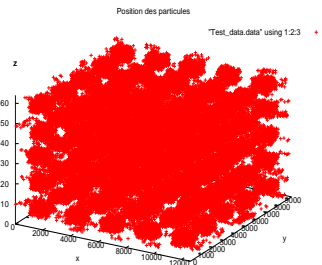
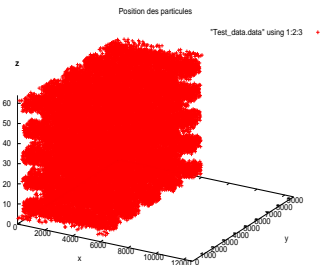
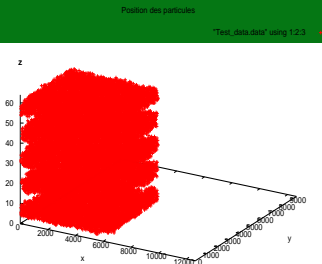
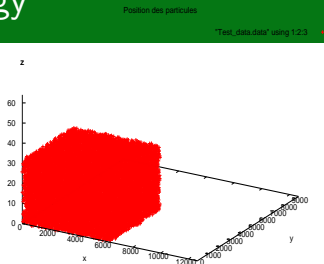
$$\forall x \in \mathbb{D}, \quad \phi(x) = \int_0^x \rho(y) dy,$$

We can solve directly the 1D discrete optimal transport problem.

The 3D case is a collection of 1D cases in the three directions.



Correction of the positions of the particles : alternative strategy



Sub-optimal transport procedure.

Correction of the positions of the particles : second alternative strategy

Birth-Death procedure in each cell, under the conservation of momentum constraint.

- If $\#\{\text{particles in } C_i\} < N_{ppc}$, add particles with the Eulerian characteristics at x_i .
- If $\#\{\text{particles in } C_i\} > N_{ppc}$, destroy $\#\{\text{particles in } C_i\} - N_{ppc} + 1$ particles and add a new one whose characteristics are their averages.

Correction of the particles velocity : Divergence free velocity

The field $\tilde{U}^{(n+1)\Delta t}$ computed from the SDEs does not match the following conservation law:

$$\nabla \cdot \langle \tilde{U}^{(n+1)\Delta t} \rangle = 0$$

We hence solve the following equation:

$$\begin{cases} \Delta P = -\frac{1}{\Delta t} \nabla \cdot \langle \tilde{U}^{(n+1)\Delta t} \rangle, & x \in \mathbb{D}, \\ \frac{\partial P}{\partial n} \Big|_{\partial \mathbb{D}} = 0, \end{cases}$$

and update the velocity field thanks to:

$$U^{(n+1)\Delta t} = \tilde{U}^{(n+1)\Delta t} - \Delta t \nabla P.$$

This insures the free divergence of $\langle U^{(n+1)\Delta t} \rangle$.

Compatibility condition for the Poisson problem :

$$\int_{\partial \mathbb{D}} \langle \tilde{U}^{(n+1)\Delta t} \rangle ds = \int_{\partial \mathbb{D}} V_{MM5} ds \simeq 0,$$

as *MM5* is a divergence free solver.

The fractional step method : $n\Delta t \longrightarrow (n+1)\Delta t$

The N -Particles dynamic : for $j = 1, \dots, N$, for $n\Delta t \leq t \leq (n+1)\Delta t$,

$$\left\{ \begin{array}{l} d\tilde{X}_t^{j,N} = \tilde{U}_t^{j,N} dt, \\ d\tilde{U}_t^{(i),j,N} = -\frac{1}{\rho} \frac{\partial \langle \mathcal{P} \rangle}{\partial x_i}(t, \tilde{X}_t^{j,N}) dt \\ \quad + D_{\tilde{U}}(t, X_t^{j,N}) dt + B_{\tilde{U}}(t, X_t^{j,N}) dW_t^{(i),j,N} \\ \quad + \text{MM5 guideline terms at the boundary, } \forall i \in \{1, 2, 3\} \\ d\omega_t^{j,N} = D_\omega(t, X_t^{j,N}) dt + B_\omega(t, X_t^{j,N}) dW_t^{(4),j,N}. \\ X_{n\Delta t}^{j,N}, U_{n\Delta t}^{(i),j,N}, \omega_{n\Delta t}^{j,N} \text{ given.} \end{array} \right.$$

- Correction of the particles position : Optimal Transport Problem.
- Correction of the particles velocity : Poisson Equation.

Mathematical study of a very simplified Langevin model

$$\begin{cases} dX_t = U_t dt, \\ dU_t = \mathbb{E} [b(U_t)/X_t] dt + dW_t, \quad t \in [0, T] \end{cases}$$

Nonlinear but smooth drift term $b : \mathbb{R}^d \rightarrow \mathbb{R}^d$, bounded continuous.
No divergence free condition.

Related works

Sznitman 86 : PoC for the Burgers' equation :

$$X_t = X_0 + W_t + 2 \int_0^t u(s, X_s) ds$$

$u(t, x) dx$ is the law of X_t .

Dermoune 03 : PoC and conditional PoC for pressurless gas equations

$$X_t = X_0 + W_t + \int_0^t \mathbb{E}[v(X_0)/X_s] ds.$$

Definition (Sznitman 89)

Let E be a separable metric space and ν a probability measure on E . A sequence of symmetric probabilities ν^N on E^N is ν -chaotic if for any $\phi_1, \dots, \phi_k \in C_b(E; \mathbb{R})$, $k \geq 1$,

$$\lim_{N \rightarrow \infty} \left\langle \nu^N, \phi_1 \otimes \dots \otimes \phi_k \otimes 1 \dots \otimes 1 \right\rangle = \prod_{l=1}^k \langle \nu, \phi_l \rangle.$$

Propagation of chaos property, prototypic example :

$$X_t^{i,N} = X_0^i + \int_0^t \frac{1}{N} \sum_{j=1}^N \sigma(X_s^{i,N}, X_s^{j,N}) dW_s^i + \int_0^t \frac{1}{N} \sum_{j=1}^N b(X_s^{i,N}, X_s^{j,N}) ds, \quad t \in [0, T],$$

Theorem

Suppose that $b(\cdot, \cdot)$ and $\sigma(\cdot, \cdot)$ are Lipschitz continuous kernels on \mathbb{R}^{2d} . Let P^N be the joint law on $(C([0, T]; \mathbb{R}^d))^N$ of the particle system $(X^{1,N}, \dots, X^{N,N})$. The sequence (P^N) is P -chaotic, where P is given by a nonlinear martingale problem.

The P -chaoticity is equivalent to the convergence of the laws of the

empirical measures $\mu^N = \frac{1}{N} \sum_{i=1}^N \delta_{X^{i,N}}$ to δ_P .

$$X_t = X_0 + \int_0^t \int_{\mathbb{R}^d} \sigma(X_s, x) P_s(dx) dW_s + \int_0^t \int_{\mathbb{R}^d} b(X_s, x) P_s(dx) ds, \quad t \in [0, T]$$

$$\begin{cases} dX_t = U_t dt, \\ dU_t = \mathbb{E} [b(U_t)/X_t] dt + dW_t, \quad t \in [0, T] \end{cases}$$

Theorem

Let $b : \mathbb{R}^d \rightarrow \mathbb{R}$ be a bounded continuous function and let (X_0, U_0) be such that $\mathbb{E}_{\mathbb{P}} [\|X_0\|_{\mathbb{R}^d} + \|U_0\|_{\mathbb{R}^d}^4] < +\infty$. Then the system admits a unique weak solution.

Well posedness of the simplified Langevin model

$$\begin{cases} dX_t = U_t dt, \\ dU_t = \mathbb{E} [b(U_t)/X_t] dt + dW_t, \quad t \in [0, T]. \end{cases}$$

The smoothed system in the space variable x :
for a given regularization ϕ_ε of the Dirac mass in \mathbb{R}^d

$$\begin{cases} dX_t^\varepsilon = U_t^\varepsilon dt, \\ dU_t^\varepsilon = \frac{\mathbb{E} [b(U_t^\varepsilon)\phi_\varepsilon(x - X_t^\varepsilon)] \Big|_{x=X_t^\varepsilon}}{\mathbb{E} [\phi_\varepsilon(x - X_t^\varepsilon)] \Big|_{x=X_t^\varepsilon} + \varepsilon} dt + dW_t, \quad t \in [0, T] \end{cases}$$

$$\mathcal{L}aw(X_t^\varepsilon, U_t^\varepsilon) = p^\varepsilon(t, x, u) dx du.$$

$$\begin{cases} dX_t^\varepsilon = U_t^\varepsilon dt, \\ dU_t^\varepsilon = G_\varepsilon[X_t^\varepsilon, U_t^\varepsilon, p_t^\varepsilon] dt + dW_t, \quad t \in [0, T]. \end{cases}$$

Uniqueness results for the laws of (X, U) and $(X^\varepsilon, U^\varepsilon)$

Uniqueness result in $L^1(\mathbb{R}^d \times \mathbb{R}^d)$ for the mild equations

$$p_t = S_t(p_0) + \int_0^t S'_{t-s}(p_s(\cdot))B[s, \cdot]ds, \text{ where } B[t, x] = \frac{\int_{\mathbb{R}^d} b(u)p_t(x, u)du}{\int_{\mathbb{R}^d} p_t(x, u)du}.$$

$$p_t^\varepsilon = S_t(p_0) + \int_0^t S'_{t-s}(p_s^\varepsilon B_\varepsilon[s, \cdot])ds,$$

$$\text{where } B_\varepsilon[s, x] = \frac{\int_{\mathbb{R}^d \times \mathbb{R}^d} b(u')\phi_\varepsilon(x - x')p_s^\varepsilon(x', u')dx' du'}{\int_{\mathbb{R}^d \times \mathbb{R}^d} \phi_\varepsilon(x - x')p_s^\varepsilon(x', u')dx' du' + \varepsilon}$$

$$\text{with } S_t(f)(x, u) = \mathbb{E}_{x, u}[f(\mathcal{X}_t, \mathcal{U}_t)]$$

$$\text{and } S'_t(f)(x, u) = \mathbb{E}_{x, u}[\nabla_u \cdot f(\mathcal{X}_t, \mathcal{U}_t)].$$

$$\mathcal{U}_t = \mathcal{U}_0 + W_t, \quad \mathcal{X}_t = \mathcal{X}_0 + \int_0^t \mathcal{U}_s ds.$$

$$\left\{ \begin{array}{l} dX_t^{\varepsilon,i,N} = U_t^{\varepsilon,i,N} dt, \\ dU_t^{\varepsilon,i,N} = \frac{1}{N-1} \sum_{i \neq j} b(U_t^{\varepsilon,i,N}) \phi_\varepsilon(X_t^{\varepsilon,j,N} - X_t^{\varepsilon,i,N}) \\ \quad \frac{1}{N-1} \sum_{i \neq j} [\phi_\varepsilon(X_t^{\varepsilon,j,N} - X_t^{\varepsilon,i,N})] + \varepsilon dt + dW_t^i, \quad t \in [0, T] \end{array} \right.$$

- Tightness result.
- Propagation of Chaos result.
- Convergence when ε tends to 0.

$$\begin{cases} dX_t^{\varepsilon,i,N} = U_t^{\varepsilon,i,N} dt, \\ dU_t^{\varepsilon,i,N} = \frac{\frac{1}{N-1} \sum_{i \neq j} b(U_t^{\varepsilon,i,N}) \phi_\varepsilon(X_t^{\varepsilon,j,N} - X_t^{\varepsilon,i,N})}{\frac{1}{N-1} \sum_{i \neq j} [\phi_\varepsilon(X_t^{\varepsilon,j,N} - X_t^{\varepsilon,i,N})] + \varepsilon} dt + dW_t^i, \quad t \in [0, T] \end{cases}$$

Theorem

Let $P^{\varepsilon,N}$ the joint law on $(C([0, T]; \mathbb{R}^{2d}))^N$ of the particle system $(X^{\varepsilon,1,N}, U^{\varepsilon,1,N}, \dots, X^{\varepsilon,N,N}, U^{\varepsilon,N,N})$.

The sequence $(P^{\varepsilon,N})$ is P^ε -chaotic, where P^ε is the law on $C([0, T]; \mathbb{R}^{2d})$ of $(X^\varepsilon, U^\varepsilon)$.

Equivalently the random measure $\mu^{\varepsilon,N} = \frac{1}{N} \sum_{i=1}^N \delta_{\{X^{\varepsilon,i,N}, U^{\varepsilon,i,N}\}}$ converges in law to the deterministic value P^ε .

Spatially Confined Langevin model $\mathbb{D} \in \mathbb{R}^d$

Impact problem with stochastic forcing.

(Deterministic motions, see e.g. Schatzman 98 , Ballard 01).

A Dirichlet condition for the impact problem.

Given a velocity field V on $[0, T] \times \partial\mathbb{D}$, $\forall t \in [0, T]$

$$\left\{ \begin{array}{l} X_t = X_0 + \int_0^t U_s ds, \\ U_t = U_0 + \int_0^t \langle b(X_s, U_s) \rangle ds + W_t \\ \quad - \sum_{0 \leq s \leq t} 2 (U_{s-} \cdot n(X_s)) n(X_s) \mathbb{1}_{\{X_s \in \partial\mathbb{D}\}} \\ \quad - \sum_{0 \leq s \leq t} 2 (U_{s-} \cdot n^\perp(X_s)) n^\perp(X_s) \mathbb{1}_{\{X_s \in \partial\mathbb{D}\}} \\ \quad + \sum_{0 \leq s \leq t} V(s, X_s) \mathbb{1}_{\{X_s \in \partial\mathbb{D}\}}. \end{array} \right.$$

We have to show that, for any $x \in \partial\mathbb{D}$,

$$\langle \mathcal{U} \rangle(t, x) = \mathbb{E} [U_t / X_t = x] = V(t, x).$$

Confined Langevin model in $\mathbb{D} = \mathbb{R}^{d-1} \times \mathbb{R}^+$

Averaged no-permeability condition

(X, U) valued on $C([0, T]; \mathbb{D}) \times \mathbb{D}([0, T]; \mathbb{R}^d)$ s.t.

$$\begin{cases} X_t = X_0 + \int_0^t U_s ds, \\ U_t = U_0 + \int_0^t \langle b(\mathcal{U}) \rangle (s, X_s) ds + W_t \\ \quad - \sum_{0 < s \leq t} 2(U_{s-} \cdot n_{\mathbb{D}}(X_s)) n_{\mathbb{D}}(X_s) \mathbb{1}_{\{X_s \in \partial \mathbb{D}\}} \end{cases}$$

Lemma

The joint law of (X_t, U_t) has a density $p(t, x, V)$, satisfying $p(t, x, V) = p(t, x, V - 2(V \cdot n_{\mathbb{D}}(x))n_{\mathbb{D}}(x), \forall x \in \partial \mathbb{D}$. Then we have

$$\mathbb{E}_{\mathbb{P}} (U_t / X_t = x) \cdot n_{\mathbb{D}}(x) = 0, \quad \forall x \in \partial \mathbb{D}.$$

Moreover $p(t, x, V)$ is the unique weak solution of the following Vlasov-Fokker-Planck Eq, with a specular boundary condition

The Vlasov-Fokker-Planck Equation

$$\left\{ \begin{array}{l} \frac{\partial p}{\partial t} = -V \nabla_x p - \nabla_V \cdot \left(\frac{\int_{\mathbb{R}^2} b(V) p(t, x, V) dV}{\int_{\mathbb{R}^2} p(t, x, V) dV} \right) + \frac{1}{2} \Delta_V p, \\ p_0(x, V) \text{ given,} \\ p(t, 0, V) = p(t, 0, V - 2(V \cdot n_{\mathbb{D}}(x)) n_{\mathbb{D}}(x)). \end{array} \right.$$

Well posedness of the confined SDE in the half plane

Starting from (X_0, U_0) with $X_0^{(2)} > 0$, and a 2D-Brownian motion (B_t)

$$\mathcal{Y}_t = X_0 + \int_0^t \mathcal{V}_s ds, \quad \mathcal{V}_t = U_0 + B_t.$$

$$\text{Set} \quad \left(X_t^{(1)}, U_t^{(1)} \right) = \left(\mathcal{Y}_t^{(1)}, \mathcal{V}_t^{(1)} \right)$$

$$\text{with} \quad X_t^{(2)} = |\mathcal{Y}_t^{(2)}|$$

$$U_t^{(2)} = \mathcal{V}_t^{(2)} \mathcal{S}_t, \quad \text{with } \mathcal{S}_t := \text{sign}(\mathcal{Y}_t^{(2)}).$$

Lemma

\mathcal{S}_t jump a countable many times and

$$U_t^{(2)} = U_0^{(2)} + W_t^{(2)} - 2 \sum_{0 < s \leq t} U_{s-}^{(2)} \mathbb{1}_{\{X_s^{(2)}=0\}}, \quad \mathbb{P}.a.s.$$

where $W_t^{(2)}$ is a Brownian motion.

McKean 63, Lachal 97.

Euler scheme for confined models, $\mathbb{D} = \mathbb{R}^+$

$$\begin{cases} X_t = X_0 + \int_0^t U_s ds, \\ U_t = U_0 + \int_0^t b(U_s) ds + W_t - \sum_{0 < s \leq t} 2U_{s-} \mathbb{1}_{\{X_s=0\}} \end{cases}$$

Euler scheme : $\Delta t > 0$ and $K \in \mathbb{N}$ s.t. $T = K\Delta t$; $t_k := k\Delta t$, $1 \leq k \leq K$, \bar{X}_0, \bar{U}_0 given.

if $\bar{X}_{t_k} + \Delta t \bar{U}_{t_k} \geq 0$ then
$$\begin{aligned} \bar{X}_{t_{k+1}} &= \bar{X}_{t_k} + \Delta t \bar{U}_{t_k} \\ \bar{U}_{t_{k+1}} &= \bar{U}_{t_k} + \Delta t b(\bar{U}_{t_k}) + (W_{t_{k+1}} - W_{t_k}). \end{aligned}$$

else
$$\begin{aligned} \tau_k &= t_k + \bar{X}_{t_k} / \bar{U}_{t_k}. \\ \bar{X}_{t_{k+1}} &= -(t_{k+1} - \tau_k) \bar{U}_{t_k} \\ \bar{U}_{t_{k+1}} &= \underbrace{-\bar{U}_{t_k} - (\tau_k - t_k) b(\bar{U}_{t_k})}_{-\bar{U}_{\tau_k}} + (t_{k+1} - \tau_k) b(-\bar{U}_{t_k}) + (W_{t_{k+1}} - W_{t_k}). \end{aligned}$$

Weak convergence of the Euler scheme

Lemma

If $b(u) = -cu$ then $h(t, x, u)$ have bounded spatial derivatives up to the order 4 and

$$|\mathbb{E}f(X_T) - \mathbb{E}f(\bar{X}_T)| \leq C\Delta t.$$

$h(t, x, u) = \mathbb{E}(f(X_T^{t,x,u}))$ solves

$$\begin{cases} \frac{\partial h}{\partial t} + u\nabla_x h + b(u)\nabla_u h + \frac{1}{2}\Delta_u h = 0, \\ h(t, 0, u) = h(t, 0, -u), \\ h(T, x, u) = f(x). \end{cases}$$

Go back to MM5 and meteorology

- \mathbb{D} is a MM5 cell of size $12\,636\text{m} \times 9\,305\text{m} \times 64$ (x, y, z), located near the coast or on the sea ($z = \sigma$).
- The wind is about 6 m/s in the dominant direction.
- Time scale : for MM5, the time step is 50 s. The time step for DSM must be smaller : we work with $\Delta t = 1$.
- DSM Cell size near 500 m for the horizontal mesh. (1000 sub-cells, with $N_{ppc} = 100$)
- Parametrisation of the pseudo-dissipation $\varepsilon(t, x) = \frac{C}{L} k^{3/2}(t, x)$.

With a stationary forcing (in time), we observe a stabilization of $k(t, x)$, $\omega(t, x)$ around values compatible with the meteorology.

Conclusion