

Monte-Carlo Methods for transport problems and kinetic equations

Remi Sentis (CEA/Bruyeres)

Introduction

Principle of the classical MC method for transport problems. + (Probabilist Interpretation)

Limit of the method (transport problem in collisional media)

Symbolic Monte-Carlo Method

1 Introduction

Evaluation of $I = \int f(x)p(x)dx$

According to the law of large numbers

$$I = \mathbf{E}(f(X)) \simeq \frac{1}{N} \sum_{j=1, N} f(X_j)$$

X_i are realizations of the r. v. X whose law is p .

Particle technique

Representation of a function $u(t, x)$ by particles

$$u(t, x)dx \simeq \sum_j w_j \delta_{X_j(t)}(dx)$$

Toy problem .

let g probability density and $u = u(t, z)$ verifying

$$\frac{\partial u}{\partial t} + \frac{\partial}{\partial z} \cdot (\vec{B} u) = 0, \quad u(0, \cdot) = g, \quad + \text{boundary cond } u = 0, (\vec{B}(z) \cdot \vec{n}_z > 0) \quad (1)$$

Evaluation of $M = \int \phi(z) u(t, z) dz$ for ϕ indicatrix f.

define the flow Z_t^z

$$\frac{d}{dt} Z_t^z = \vec{B}(Z_t^z), \quad Z_{t=0}^z = z$$

Denote \mathcal{Z}_t the r.p. $\frac{d}{dt} \mathcal{Z}_t = \vec{B}(\mathcal{Z}_t)$, whose law $\mathcal{Z}_{t=0}$ is $g(\cdot) dz$

Proposition. (proba. interpretation) For T positive :

$$\int \phi(z) u(T, z) dz = \int g(z) \phi(Z_T^z) dz = \mathbf{E}_g(\phi(\mathcal{Z}_T))$$

Proof. Let φ verifying :

$$-\frac{\partial \varphi}{\partial t} - \vec{B} \cdot \frac{\partial \varphi}{\partial z} = 0, \quad \varphi(T) = \phi, \quad \text{then : } \frac{\partial}{\partial t} (\varphi(t, Z_t^z)) = 0. \quad \text{But } \frac{\partial}{\partial t} \left(\int \varphi(t) u(t) \right) = 0$$

2 Principle of the classical MC method: Proba. Interpretation

■ Evolution transport equations

$$\frac{\partial u}{\partial t} + v \cdot \frac{\partial u}{\partial x} + ru - Ku = 0, \quad (2)$$

$$\begin{aligned} u(0, \cdot) &= G \\ u|_{\partial\mathcal{D}}(x, v) &= h(x, v), \quad v \in Z_x^+ = \{v \cdot n_x > 0\} \end{aligned} \quad (3)$$

where K is conservative (collision operator)

$$Ku(x, v) = \int_{\mathcal{V}} \sigma_x(v') k_x(v', v) u(x, v') dv' - \sigma_x(v) u(x, v)$$

such that $\int k(v', v) dv = 1$.

Set of velocity \mathcal{V} provided with dv s.t. $\int_{\mathcal{V}} dv = 1$

■ Stationary transport equations .

Let $u = u(x, v)$ be solution of

$$\begin{aligned} v \cdot \nabla u + ru - Ku &= f \\ u|_{\partial\mathcal{D}}(x, v) &= h(x, v), \quad \forall v \in Z_x^+ \end{aligned}$$

■ Evolution equations (2)

$u \longmapsto (v \cdot \frac{\partial}{\partial x} - K)u$ is associated to the processus $X(t), V(t)$

characterized by:

- Between the jump of V :

$$\frac{\partial}{\partial t} X(t) = V(t), \quad V \text{ constant}$$

- The stopping time η is such that

$$P(\eta \in [t, t + dt]) = e^{-\sigma(X,V)t} \sigma(X, V) dt$$

- At this stopping time η , the velocity jump from $V(\eta_-)$ to $V(\eta)$

according to prob. law $k(X(\eta), V(\eta_-), w) dw$

Remark :

$$\sigma^{-1} \quad \text{m. f. time}, \quad v\sigma^{-1} \quad \text{m. f. p.}$$

Proposition (probabilist interpretation).

$$\text{Assume that } g \geq 0, \quad \int_D \int_{\mathcal{V}} g(x, v) dx dv = 1$$

For any test fonction $\varphi \in C_b(\mathcal{D} \times \mathcal{V})$,

$$\begin{aligned} \ll u(t, \cdot) \varphi \gg &= \int \int_{\mathcal{D}\mathcal{V}} u(t, x, v) \varphi(x, v) dx dv = \\ &= E_g \left[\varphi(X(t), V(t)) \exp\left(-\int_0^t r(X, V) d\zeta\right) \right] \end{aligned}$$

$E_g =$ expectation knowing the law of $(X(0), V(0))$ is $g(\cdot, \cdot) dx dv$.

2.1 Principle of the method

■ Pb without boundary conditions

* Generation of N processes, ind. equid. : $(X_p, V_p), p = 1, \dots, N$.

knowing that

$$\frac{\partial}{\partial t} X_p = V_p,$$

V_p jump process

at time $t = 0$, (X_p, V_p) distribu a. to g

* Initialization of the weights w_p^0 s.t. in D

$$\sum_{p=1}^N w_p^0 = 1, \quad \sup |w_p^0| \rightarrow 0$$

* Each weight solves

$$\frac{\partial}{\partial t} w_p + w_p r(X_p, V_p) = 0.$$

■ **Corollary.** $N \rightarrow \infty$, we get

$$\begin{aligned}\Phi^N(t) &= \sum_{p=1, \dots, N} w_p(t) \delta_{X_p(t)}(dx) \delta_{V_p(t)}(dv) \\ \ll \Phi^N(t) \varphi \gg &\rightarrow \ll u(t, \cdot) \varphi \gg\end{aligned}$$

2.2 Implementation of the method.

* Discretization into sub-domains D' such that the coef. are constant

* In each sub-domain D' , initialize the weights w_p^0 s.t.

$$\sum_{p/X_p \in D'} w_p^0 = \int_{D'} \int_{\mathcal{V}} g(x, v) dx dv$$

$X_p(0), V_p(0)$ generated a. to $g(\cdot, \cdot)|_{D'}$

* Tracking of the particles.

$$\frac{\partial}{\partial t} X_p = V_p,$$

V_p jump process

* Compute the outgoing time τ_p out of D' and the collision time η_p .

If $\eta_p < \tau_p$, velocity jumps a. to the law k .

* The weight varies a. to

$$\frac{\partial}{\partial t} w_p + w_p r(X_p, V_p) = 0.$$

* Tally of the weights of the particles and evaluate at final time

$$w_p(t)\varphi(X_p(t), V_p(t))$$

.Boundary condition.

Assume V with spherical symmetry

Consider transport equation with h independent of v .

Particles are generated on each face Γ of the boundary in the half-sphere

$$Z^+ = v \quad / \quad v \cdot n_\Gamma > 0$$

s. t. the prob. law of $\Omega = v/|v|$ is :

$$\frac{1}{\int_{Z^+} \Omega' \cdot n_\Gamma d\Omega'} \Omega \cdot n_\Gamma d\Omega$$

i.e. for $\mu = \Omega \cdot n_\Gamma$ on $[0, 1]$, the prob. law is (Lambert law)

$$2\mu d\mu.$$

the azimuthal angle is equidistributed.

For a general function $h(\mu)$, the proba. law is $\mu h(\mu)$ for each particle.

3 Limit of the method

Monte-Carlo algorithms are non efficient if

$$\lambda = \frac{\overline{|v|}}{\overline{\sigma}} \ll L_{char}, \quad L_{char} \text{ char. length,} \quad \overline{\sigma} \text{ char. value}$$

The Monte-Carlo method becomes prohibitive (to much jumps in each cell)

Then diffusion approximation is necessary

Hypothesis: k is symmetric, V spherical symmetry and

$$\varepsilon = \lambda/L_{char}, \quad r \ll \sigma, \quad r = \varepsilon r_0,$$

$$Ku = \sigma(x)Mu, \quad \sigma = \frac{\sigma_0}{\varepsilon} \quad Mu(v) = \int_{\mathcal{V}} k(v', v)u(v')dv' - u(v)$$

$$\tilde{\bullet} = \int_{\mathcal{V}} \bullet(v)dv$$

Let u_ε be the solution of

$$\begin{aligned} v \cdot \nabla u_\varepsilon + \varepsilon r_0 u_\varepsilon - \frac{\sigma_0}{\varepsilon} M u_\varepsilon &= \varepsilon f_0 \\ u_\varepsilon|_{\partial \mathcal{D}}(x, v) &= h_x(v), \quad \forall v \in Z_x \end{aligned}$$

Theorem (classical result ~1960)

If $h(x, \cdot)$ independent of v ; when $\varepsilon \rightarrow 0$, we get $u_\varepsilon(x, v) \rightarrow U(x)$ where

$$r_0 U - \nabla \cdot \left(\frac{1}{3\sigma_0} \nabla U \right) = \tilde{f}_0, \quad \text{with boundary cond. } U(x)|_\Gamma = h_x$$

So the behavior of the solution of (2) is like a diffusion one.

Remark. If $h(x, \cdot)$ depends on v , there exists boundary layers (Chandrasekhar)

Transport problem in a zoom region near a boundary point x_b . (V is the sphere, $\mu = v \cdot n_{x_b}$).

$$\text{Set } y = \frac{x - x_b}{\varepsilon} \cdot n_{x_b} \text{ and } u_\varepsilon(x) = U_0(x) + \phi_{x_b} \left(\frac{x - x_b}{\varepsilon} \cdot n_{x_b}, \mu \right) + \varepsilon u_1 + \varepsilon^2 u_2 + \dots$$

ϕ is a boundary layer, solution to a transport equation.

Theorem (G. Papanicolaou ~1975, R.S. 1981).

The solution u_ε satisfies $u_\varepsilon(x, v) \simeq U(x)$

$$r_0 U - \nabla \cdot \left(\frac{1}{3\sigma_0} \nabla U \right) = \tilde{f}_0, \quad \text{with boundary cond. } U(x)|_\Gamma = C(h_x)$$

where the operator C is defined by the way. Let $\psi(y, \mu)$ be the unique bounded solution of

$$\begin{aligned} \mu \frac{\partial}{\partial y} \psi - M\psi &= 0 \quad \text{with } y \in [0, +\infty] \\ \psi(0, \mu) &= g(\mu), \text{ on } Z^+ \end{aligned} \tag{4}$$

then there exists a fonction H (Chandrasekhar) such that

$$\lim_{y \rightarrow \infty} \psi(y, \mu) = C(g) = \int_0^1 g(\mu) H(\mu) \mu \frac{d\mu}{2}. \blacksquare$$

4 The symbolic Monte-Carlo method

Stationary problem with $r > 0$. ($q = \sigma + r$)

$$\begin{aligned}\Omega.\nabla u + qu &= \sigma\tilde{u} \\ u(x, \Omega) &= h, \text{ on } Z^+.\end{aligned}$$

Denote now $v = v(x, \Omega)$, the solution

$$\begin{aligned}\Omega.\nabla v + qv &= 0 \\ v(x, \Omega) &= h, \text{ on } Z^+\end{aligned}$$

For each cell C_j , denote ζ_j the solutions

$$\begin{aligned}\Omega.\nabla\zeta_j + q\zeta_j &= q\mathbf{1}_{C_j} \\ \zeta_j &= 0 \text{ on } Z^+\end{aligned} \tag{5}$$

We introduce the coefficients M_{kj} ,

$$M_{kj} = \int_{C_k} \sigma_k \tilde{\zeta}_j(x) \frac{dx}{|C_k|}$$

and the data

$$V_k = \int_{C_k} \tilde{v}(x) \frac{dx}{|C_k|}$$

We can check that

$$\sum_j M_{kj} \leq \sigma_k \tag{6}$$

Proposition.

The linear system

$$q_k \phi_k - \sum_j M_{kj} \phi_j = \sigma_k V_k$$

has a unique solution (ϕ_j) . If we set

$$\eta(x, \Omega) = \sum_k \zeta_k(x, \Omega) \phi_k$$

then $v + \eta$ is a good approximation of u , when δx tends to 0.

Sketch of the proof

The matrix of the linear system is diagonal dominant.

Since the V_j are positive, the solution (ϕ_j) exists, is positive.

. η is solution to

$$\begin{aligned}\Omega \cdot \nabla \eta + q\eta &= \sum_k \mathbf{1}_{C_k} q \phi_k \\ \eta &= 0 \text{ on } Z^+\end{aligned}$$

According to the definition of ϕ_j , we get

$$\sum_k \mathbf{1}_{C_k} q \phi_k = \sum_k \mathbf{1}_{C_k} \sigma_k V_k + \sum_k \sum_j \mathbf{1}_{C_k} \sigma_k \int_{C_k} \tilde{\zeta}_j(x) \frac{dx}{|C_k|} \phi_j = \sum_k \mathbf{1}_{C_k} \sigma_k \int_{C_k} (\tilde{v} + \tilde{\eta})(x) \frac{dx}{|C_k|}$$

$$\text{Thus } (\Omega \cdot \nabla + q)(v + \eta) = \sum_k \mathbf{1}_{C_k} \sigma_k \int_{C_k} (\tilde{v} + \tilde{\eta})(x) \frac{dx}{|C_k|}. \quad \square$$

$$\text{Remark : } \int_{\partial C_k} \tilde{\zeta}_j \widetilde{\Omega \cdot \mathbf{n}} d\gamma(x) = q_k \int_{C_k} \tilde{\zeta}_j dx, \quad \Rightarrow M_{kj} |C_k| = \frac{\sigma_k}{q_k} \int_{\partial C_k} \tilde{\zeta}_j \widetilde{\Omega \cdot \mathbf{n}} d\gamma(x)$$

this corresponds to the particles born in C_j , absorbed in cell C_k .

Numerical method .

1. Free fly. We first generate particles on the boundary

Track them according to a free fly technique (without coll.)

2. Evaluation of the coefficients M_{kj} .

For each C_j , generate particles (X_p, Ω_p) such that their weights w_p^0

$$\sum_p w_p^0 = |C_j|q_j$$

Track these particles according to free fly without collisions

Estimation of M_{kj} by using the flux on the boundary of C_k

$$M_{kj}|C_k| = \frac{\sigma_k}{q_k} \sum_{p, \text{born in } j} (w_{p,C_k}^{in} - w_{p,C_k}^{out}) \quad (7)$$

w_{p,C_k}^{in} and w_{p,C_k}^{out} = weights of p when it goes in and out C_k . Thus (6) holds.

3. Finally, we solve the linear system

$$q_k \phi_k - \sum_j M_{kj} \phi_j = \sigma_k V_k$$

The method is very adaptable. Example of generalization

$$\begin{aligned}\Omega \cdot \nabla u + ru + \sigma(u - \tilde{u}) + \tau u &= \tau \theta \\ u(x, \Omega) &= h, \text{ on } Z^+. \\ \frac{\partial}{\partial t} \theta + \tau \theta - \tau \tilde{u} - \nabla(\lambda \nabla \theta) &= 0. \\ &+ B.C \text{ for } \theta\end{aligned}$$

5 Conclusion

The Monte Carlo method are very adaptable, powerfull
but are prohibitive in collisional domain.

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The Symbolic MC method for stationary problems is a variant of a well known method in neutronics :
the collision probability method.

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The mathematical analysis allows to adapt this method to other situations for example to radiative transfer
problems
and to perform coupling with other methods.