

Colloquium du CERMICS



École des Ponts
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Structured equations in Biology

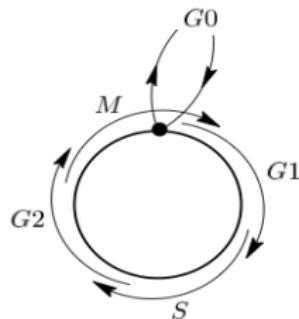
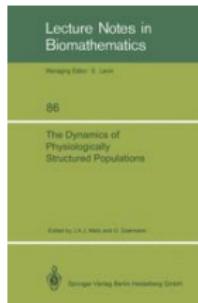
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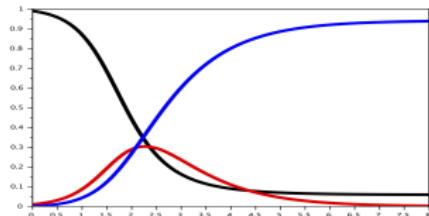
1. Why structured equations ?
2. Renewal and growth fragmentation eqs
3. Generalised relative entropy
4. Monge-Kantorovich distance

The standard SIR system

- $S(t)$ = Susceptible individuals
- $I(t)$ = Infectious individuals (who propagate the epidemic)
- $R(t)$ = Removed individuals

The SIR system (D. Bernouilli, 1760 ?) reads

$$\begin{cases} \dot{S}(t) = -\beta SI \\ \dot{I}(t) = \beta SI - \gamma I \\ \dot{R}(t) = \gamma I \end{cases}$$



It is easy to see that $S(t) \rightarrow S_\infty > S_{herd}$ (for $R_0 > 1$)

$$\ln S_\infty - \frac{\beta}{\gamma} S_\infty = \ln S^0 - \frac{\beta}{\gamma} [S^0 + I^0]$$

Examples of structured equations



Heterogeneous population (contact matrices)

- $x = \text{Social contacts}$
- $I(t, x) = \text{immune system status}$

$$\begin{cases} \dot{S}(t, x) = -S(t, x) \int_{\Omega} \beta(x, y) I(t, y) \, dy & (+\Delta S) \\ \dot{I}(t, x) = S(t, x) \int_{\Omega} \beta(x, y) I(t, y) \, dy - \gamma(x) I(t, x) \\ \dot{R}(t, x) = \gamma(x) I(t, x) \end{cases}$$

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Theorem (Almeida, Bliman, Nadin, BP, Vauchelet)

The solution converges to the UNIQUE steady state $S_{\infty} < S^0$ and

$$\ln S_{\infty}(x) - \int \frac{\beta(x, y)}{\gamma(y)} S_{\infty}(y) dy = \ln S^0(x) - \int \frac{\beta(x, y)}{\gamma(y)} (S^0(y) + I^0(y)) dy$$

Examples of structured equations



The Kermack-McKendrick renewal equation (1927)

$$\begin{cases} \frac{d}{dt}S(t) = B - \mu_S S(t) - I(t)S(t) \\ I(t) := \int_0^\infty \beta(s)n_I(t,s)ds \\ \frac{\partial}{\partial t}n_I(t,s) + \frac{\partial}{\partial s}n_I(t,s) + (\mu_I + \gamma(s))n_I(t,s) = 0 \\ n_I(t,s=0) = I(t)S(t) \end{cases}$$

Theorem (Magal, McCluskey, Webb, 2010). Define

$$\mathcal{E}(t) = \int_0^\infty \psi(s)\bar{n}_I(s)\left[\frac{n_I(t,s)}{\bar{n}_I(s)} - \ln\frac{n_I(t,s)}{\bar{n}_I(s)}\right]ds + \bar{S}\ln S(t) - S(t).$$

Then, we have

$$\frac{d}{dt}\mathcal{E}(t) \leq -D(t) \leq 0, \quad D(t) = \frac{\mu_S}{S(t)}(\bar{S} - S(t))^2.$$

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The renewal equation



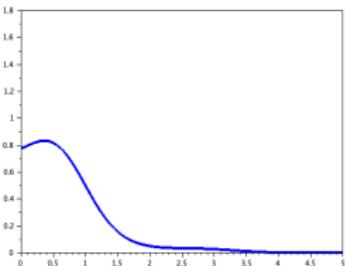
$$\begin{cases} \frac{\partial}{\partial t} n(t, x) + \frac{\partial}{\partial x} n(t, x) + d(x)n(t, x) = 0, & t \geq 0, x \geq 0 \\ N(t) := n(t, x=0) = \int_0^\infty b(y)n(t, y)dy \\ n(t=0, x) = n^0(x) \end{cases}$$
$$b, d \in L_+^\infty(0, \infty).$$

- Very useful (demography, cell cycle, anomalous diffusions)
- Very standard (Feller)
- Nonlinear versions are complex

The renewal equation



$$\begin{cases} \frac{\partial}{\partial t} n(t, x) + \frac{\partial}{\partial x} n(t, x) + d(x)n(t, x) = 0, & t \geq 0, x \geq 0 \\ N(t) := n(t, x=0) = \int_0^\infty b(y, I(t))n(t, y)dy \\ I(t) = \int q(x)n(t, x)dx \end{cases}$$



Another useful equation

$$\begin{cases} \frac{\partial n(t, x)}{\partial t} + \frac{\partial [g(x)n(t, x)]}{\partial x} + b(x)n(t, x) = 2 \int_x^\infty b(y)\kappa(x, y)n(t, y)dy \\ n(t, x=0) = 0, \quad g(0) > 0, \\ n(t=0, x) = n^0(x). \end{cases}$$

- $b(y)$ = is the division rate of cells/polymers/messages of sizes y
- $\kappa(x, y) = 0$ for $x > y$
- $\int_0^y \kappa(x, y)dx = 1, \quad \int_0^y x \kappa(x, y)dx = y/2$
- $\frac{d}{dt} \int_0^\infty n(t, x)dx = \int_0^\infty b(x)n(t, x)dx$
- $\frac{d}{dt} \int_0^\infty xn(t, x)dx = \int_0^\infty g(x)n(t, x)dx.$

Age and size structured

$$\begin{cases} \frac{\partial n(t,x,z)}{\partial t} + \frac{\partial n(t,x,z)}{\partial x} + \frac{\partial [g(z) n]}{\partial z} + d(x, z) n(t, x, z) = 0, & x > 0, z > 0, \\ n(t, x, z = 0) = 0, \\ n(t, x = 0, z) = \int_{x=0}^{\infty} \int_{z'=z}^{\infty} d(x', z') \kappa(z, z') n(t, x', z') dx' dz' \end{cases}$$

Age and space structured

$$\begin{cases} \frac{\partial n(t,x,z)}{\partial t} + \frac{\partial n(t,x,z)}{\partial x} + d(x) n(t, x, z) = 0, & x > 0, z \in \mathbb{R}^d \\ n(t, x = 0, z) = \int_0^{\infty} \int_{\mathbb{R}^d} d(x) n(t, x, z + \varepsilon \eta) k(\eta) dx d\eta \end{cases}$$

- Doumic, M. ; Hoffmann, M. ; Krell, N. ; Robert, L. et al (2015)
- Berry, H. ; Lepoutre, T. ; González, A. ; Acta Appl. Math. (2016)
- Calvez, V. ; Gabriel, P. ; Mateos G. ; Asymptot. Anal. (2019)
- Franck M. ; Goudon T. ; KRM (2018)

Which general tools ?



- Generalized relative entropy
- Modified Monge-Kantorovich distance

1. Why structured equations ?
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Generalized relative entropy



All (linear) equations preserving positivity satisfy the GRE

$$\mathcal{L}\varphi + \lambda_0\varphi = 0, \quad \varphi > 0, \quad \mathcal{L}^*\psi + \lambda_0\psi = 0, \quad \psi > 0,$$

Let

$$\frac{\partial n(t)}{\partial t} + \mathcal{L}n(t, x) = 0.$$

GRE Principle. For all $H(\cdot)$ convex, $u(t, x) = e^{-\lambda_0 t} \frac{n(t, x)}{\varphi(x)}$ satisfies

$$\frac{d}{dt} \int \psi(x)\varphi(x)H(u(t, x)) dx = -D_H(t) \leq 0$$

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- $\int \psi(x)n(t, x)dx = e^{\lambda_0 t} \int \psi(x)n^0(x)dx$ Conservation law
- $u^0 \leq C^0 \implies u(t, x) \leq C^0$
- $\frac{d}{dt} \ln n(t, x) \leq \max_x \frac{d}{dt} \ln n^0(x)$
- Explains Kermack-McKendrick $H(u) = u - \ln u$.

Generalized relative entropy



Examples. $u(t, x) = e^{-\lambda_0 t} \frac{n(t, x)}{\varphi(x)}$

$$\frac{\partial n(t, x)}{\partial t} - \Delta n + \operatorname{div}(n \nabla V) = 0 \quad (\text{Fokker-Planck})$$

$$\lambda_0 = 0, \quad \psi = 1, \quad \varphi = e^V, \quad D_H = \int \varphi H''(u) |\nabla u|^2$$

$$\begin{cases} \frac{\partial}{\partial t} n(t, x) + \frac{\partial}{\partial x} n(t, x) + d(x)n(t, x) = 0, & (\text{Renewal}) \\ N(t) := n(t, x=0) = \int_0^\infty b(y)n(t, y)dy \end{cases}$$

$$\varphi = e^{-\int_0^x d - \lambda_0 y}, \quad D_H = \int b\varphi H(u) - H\left(\int b\varphi u\right)$$

Examples. $u(t, x) = e^{-\lambda_0 t} \frac{n(t, x)}{\varphi(x)}$

In most of the cases, the spectral theory is not so easy

$$\frac{\partial n(t, x)}{\partial t} + \frac{\partial [g(x)n(t, x)]}{\partial x} + b(x)n(t, x) = 2 \int_x^\infty b(y)\kappa(x, y)n(t, y)dy$$

$$D_H = 2 \iint \psi(x)\varphi(y)b(y)\kappa(x, y) \\ [H(u(y)) - H(u(x)) - H'(u(x))(u(y) - u(x))] dx dy$$

Spectral gap. Aim is to find $\lambda_1 > 0$ such that

$$\|e^{-\lambda_0 t} n(t, x) - \rho^0 \varphi\| \leq C e^{-\lambda_1 t}$$

■ Poincaré inequality : when $\int \psi \varphi u(x) dx = 0$

$$\lambda_1 \int \psi \varphi H(u) dx \leq D_H(u)$$

- Doeblin's method
- Method of integral equation
- Bacry-Emery's method

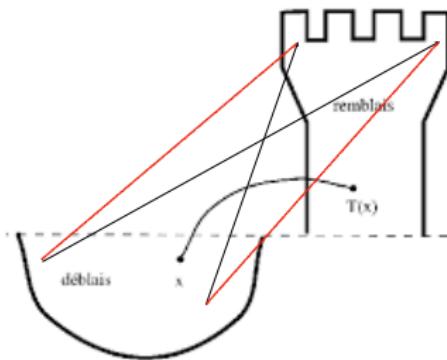
(Ryzhik, BP, Doumic, Gabriel, Mischler, Cañizo, Yoldas, Laurencot)

1. Why structured equations ?
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4. Monge-Kantorovich distance

- Ω open subset of \mathbb{R}^d
- Cost function $c(x, y) \geq 0$, $x \in \Omega$, $y \in \Omega$
- μ_1, μ_2 probability measures on Ω
- **Monge :** $T : \Omega \rightarrow \Omega$, $T_{\#}\mu_1 = \mu_2$

$$c(x, y) = |x - y|$$

$$d_{MK}(\mu_1, \mu_2) := \min_T \int_{\Omega} c(x, T(x)) \mu_1(x) dx$$



Monge-Kantorovich distance



- Ω open subset of \mathbb{R}^d
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- **Kantorovich :**

$$d_{MK}(\mu_1, \mu_2) := \min_{\mu} \int_{\Omega} c(x, y) \mu(x, y) dxdy$$

$$\int_{\Omega} \mu(x, y) dy = \mu_1(x), \quad \int_{\Omega} \mu(x, y) dx = \mu_2(y)$$

- Monge is equivalent to choose $\mu(x, y) = \mu_1(x)\delta(y = T(x))$.

Monge-Kantorovich distance



- **Monge :** $T : \Omega \rightarrow \Omega, \quad T_{\#}\mu_1 = \mu_2$

$$d_{MK}(\mu_1, \mu_2) := \min_T \int_{\Omega} c(x, T(x)) \mu_1(x) dx$$

- **Kantorovich :** $d_{MK}(\mu_1, \mu_2) := \min_{\mu} \int_{\Omega} c(x, y) \mu(x, y) dx dy$

$$\int_{\Omega} \mu(x, y) dy = \mu_1(x), \quad \int_{\Omega} \mu(x, y) dx = \mu_2(y)$$

- **Brenier :** For $\Omega = \mathbb{R}^d$, $c(x, y) = |x - y|^2$, μ_i 'smooth', T is optimal if and only if

$$T(x) = \nabla \Phi(x) \quad \text{with} \quad \Phi : \mathbb{R}^d \rightarrow \mathbb{R} \text{ convex}$$

$$\det D^2 \Phi(x) = \frac{\mu_1(x)}{\mu_2(T(x))} \quad \text{Monge-Ampere eq.}$$

(Evans-Gangbo, Cafarelli-Feldman-McCann, Trudinger-Wang, Ambrosio)

And PDEs ?

- **Otto** : Formal Riemannian metric.

Theorem The porous media equation

$$\frac{\partial n}{\partial t} - \Delta A(n) = 0$$

is non-expansive for the quadratic cost.

Proof uses the Monge-Ampere equation (see also Bolley-Carrillo)

No proof known by 'coupling'

- **Tanaka** : Homogeneous Boltzmann (new proof by 'coupling', Rousset, Fournier-BP)

$$\partial_t n - \Delta n = 0, \quad x \in \mathbb{R}^d, \quad t \geq 0$$

Theorem : For costs $c(x - y)$, we have

$$d_{MK}(n_1(t), n_2(t)) \leq d_{MK}(n_1^0, n_2^0)$$

Proof : Consider v solution of

$$\frac{\partial v}{\partial t} - \Delta_x v - \Delta_y v - 2\nabla_x \cdot \nabla_y v = 0, \quad x, y \in \mathbb{R}^d, \quad t \geq 0$$

with a compatible initial data

$$\int v^0(x, y) dy = n_1^0(x) \quad \int v^0(x, y) dx = n_2^0(y)$$

Step 1. $v \geq 0$ because $\begin{pmatrix} I & -I \\ -I & I \end{pmatrix}$ is nonnegative

$$\partial_t n - \Delta n = 0, \quad x \in \mathbb{R}^d, \quad t \geq 0$$

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$$\int v^0(x, y) dy = n_1^0(x) \quad \int v^0(x, y) dx = n_2^0(y)$$

Step 2. Marginals are correct. Integrate in y :

$$\frac{\partial v_1(x, t)}{\partial t} - \Delta_x v_1 = 0$$

$$\partial_t n - \Delta n = 0, \quad x \in \mathbb{R}^d, t \geq 0$$

Theorem : For costs $c(x - y)$, we have

$$d_{MK}(n_1(t), n_2(t)) \leq d_{MK}(n_1^0, n_2^0)$$

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$$\frac{\partial v}{\partial t} - \Delta_x v - \Delta_y v - 2\nabla_x \cdot \nabla_y v = 0, \quad x, y \in \mathbb{R}^d, t \geq 0$$

Step 3. The distance diminishes

$$\begin{aligned} \frac{d}{dt} \int c(x - y)v(x, y, t)dx dy &= \\ \int v(x, y, t) \left(\Delta_x c(x - y) + \Delta_y c(x - y) + 2\nabla_x \nabla_y c(x - y) \right) dx dy \\ &= \int v(x, y, t) (\Delta_{x+y} c(x - y)) dx dy = 0 \end{aligned}$$

$$\partial_t n - \Delta n = 0, \quad x \in \mathbb{R}^d, \quad t \geq 0$$

Theorem : For costs $c(x - y)$, we have

$$d_{MK}(n_1(t), n_2(t)) \leq d_{MK}(n_1^0, n_2^0)$$

Step 4. Conclusion. For all initial coupling v^0

$$d_{MK}(n_1(t), n_2(t)) \leq \int c(x - y) v(x, y, t) dx dy$$

$$\leq \int c(x - y) v^0(x, y) dx dy$$

$$\approx d_{MK}(n_1^0, n_2^0)$$

Challenge : $\partial_t n - \Delta(A(x)n) = 0$

Theorem. (N. Fournier, BP)

The Renewal and Growth-Fragmentation equations (and many others) are non-expansive for the cost type⁽¹⁾

$$c(x - y) = \min(|x - y|, a), \quad a \text{ related to Lipschitz bounds on coef.}$$

History.

Fournier and Locherbach (neuron networks)

Chafai, Malrieu, Paroux, Guillin, Zitt... (TCP connections)

Dobrushin, Vlasov equations. Functional anal. and its Appl. (1979)

Structured equation



Renewal equation as an example.

$$\begin{cases} \frac{\partial n(x,t)}{\partial t} + \frac{\partial [g(x)n]}{\partial x} + d(x)n = b(x)N(t), & t \geq 0, x \geq 0, \\ n(x=0, t) = 0, & N(t) = \int_0^\infty d(x)n(x, t)dx. \end{cases}$$

$$g' \leq 0, \quad g(0) \geq 0, \quad \int_0^\infty b = 1$$

$$0 < a < 1 \quad \text{and} \quad a = \inf_{|x-y|<1} \frac{|x-y| \max(d(x), d(y))}{|d(x) - d(y)|}$$

Example : $d(x) = \alpha + \beta x^p$, $p \geq 1$.

$$c(x-y) = \min(|x-y|, a)$$

$$d_{MK}(n_1(t), n_2(t)) \leq d_{MK}(n_1^0, n_2^0)$$

Structured equation



$$\begin{aligned} \frac{\partial v}{\partial t} + \frac{\partial g(x)v}{\partial x} + \frac{\partial g(y)v}{\partial y} + \max(d(x), d(y))v \\ = b(x)\delta(x - y) \int \min(d(x'), d(y')) v(dx', dy', t) \\ + b(x) \int (d(x') - d(y))_+ v(dx', y, t) \\ + b(y) \int (d(y') - d(x))_+ v(x, dy', t). \end{aligned}$$

with an initial data v^0 whose marginals are n_1^0 and n_2^0 .

$$\begin{aligned} c(x, y) \geq \int c(z, y) b(z) dz \frac{(d(x) - d(y))_+}{\max(d(x), d(y))} \\ + \int c(x, z) b(z) dz \frac{(d(y) - d(x))_+}{\max(d(x), d(y))} \end{aligned}$$

Entropy, Poincaré and Doeblin :

Perthame, B., Ryzhyk, L. Exponential decay for the fragmentation or cell-division equation. J. Diff. Eq. (2005)

Michel, P.; Mischler, S., Perthame, B. General relative entropy inequality : an illustration on growth models. J. Math. Pures et Appl. (2005)

Recent theory : Laurençot, Doumic, Gabriel, Mischler, Canizo, Yoldas, Cloez, Monmarche and coll.

Monge-Kantorovich :

Fournier, N., Perthame, B. Transport distances for PDEs : the coupling method, EMS Surveys in Math. Sc. (2021)

Fournier, N., Perthame, B. Monge-Kantorovich distance and structured equations. SIMA. In press

THANK YOU