

A pathwise construction of Birth-Death-Swap
systems leading to an averaging result in the
presence of two timescales

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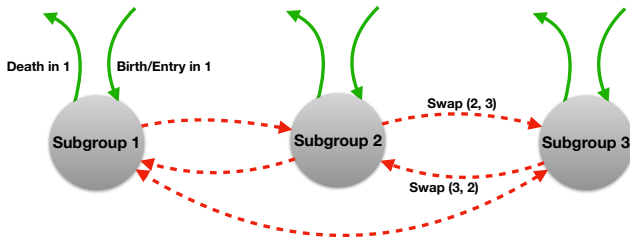
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Systèmes de Birth-Death-Swap

Dynamique stochastique de population structuré en sous-groupes discrets (lieu d'habitation, habitudes alimentaires, stratégie...):

- ▶ **Population:** $Z_t = (Z_t^1, \dots, Z_t^p) \in \mathbb{N}^p$.
- ▶ Composition de la population modifiée par les événements démographiques et les **swap** (changement de caractéristiques).



- ▶ **Population agrégée** $Z_t^h = \sum_{i=1}^p Z_t^i =$ taille de la population.

Cadre classique: Processus de naissance et mort multi-types Markoviens.

- 1 Pas de changements de caractéristiques.
- 2 Intensités des évènements démographiques ne dépendent que de l'état de la population.

Généralisation : Systèmes de Birth-Death-Swap (BDS).

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 - ↳ Applications: écologie (Auger et al.), biologie (Billiard et al. (2017)), botnets en interaction (Song et al. (2010)).
 - ↳ Génère des interactions au niveau agrégé/macroscopique.
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 - ↳ Prendre en compte la variabilité de l'environnement au cours du temps \Rightarrow Intensités stochastiques :

$$P(\text{ ev de type } \gamma \in]t, t + dt] | \mathcal{G}_t) = \mu^\gamma(\omega, t, Z_t) dt.$$

Questions

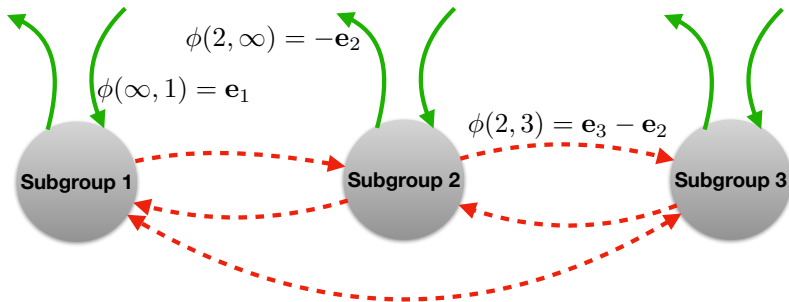
- 1 General conditions for existence of BDS systems?
 - 2 How does heterogeneity/swap events impact the global population dynamics?
 - ↳ Study of the **aggregated population** in the presence of **two timescales** (**fast swap** events).
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Stochastic intensities: classical results cannot be applied.

- ↳ Development of new tools for the study of BDS systems:
 - Realization of point processes by **strong domination**.
 - General averaging result by **stable convergence**.

- 1 Setup
- 2 Pathwise representation of BDS systems
- 3 Averaging result and aggregation in the presence of two timescales

Birth, Death and Swap events



Each event is characterized by a particular jump in the population:

- ▶ **Jumps function** $\phi : \mathcal{J}$ (set of all events) $\longrightarrow \mathbb{Z}^P$.
 - Swap event at t : $\Delta Z_t = \phi(i, j) = \mathbf{e}_j - \mathbf{e}_i$.
 - Birth event: $\Delta Z_t = \phi(\infty, i) = \mathbf{e}_i$, Death: $\Delta Z_t = \phi(i, \infty) = -\mathbf{e}_i$.

- ▶ Each type of event (birth, death, swap) $\gamma \in \mathcal{J}$ is associated with:

$$N_t^\gamma = \sum_{0 < s \leq t} \mathbb{1}_{\{\Delta Z_s = \phi(\gamma)\}} \quad (1)$$

- ▶ **Assumption:**

- $\forall \gamma \in \mathcal{J}$, N^γ has the \mathcal{G}_t -intensity $\mu(\omega, t, Z_{t-})$:

$$P(N_{t+dt}^\gamma - N_t^\gamma = 1 | \mathcal{G}_t) \simeq \mu^\gamma(t, Z_t) dt,$$

- $\mu(t, z) = (\mu^\gamma(t, z))_{\gamma \in \mathcal{J}}$ a predictable functional.

- ▶ **Vector notation:** the multivariate counting process $\mathbf{N} = (N^\gamma)_{\gamma \in \mathcal{J}}$ has the \mathcal{G}_t -multivariate intensity $\mu(t, Z_{t-})$.

Examples:

- ▶ Birth intensity functional: $\mu^{b,i}(\omega, t, z) = b_t^i(\omega)z^i + \underbrace{\lambda^i(t, Y_t)}_{\text{immigration rate}}$
 - ▶ Death intensity functional: $\mu^{d,i}(\omega, t, z) = d_t^i(\omega)z^i + \sum_{j \neq i} \underbrace{c(z^i, z^j)}_{\text{competition}}$
-

Population decomposition:

- ▶ Population process can be expressed as a linear function of \mathbf{N} :

$$Z_t = Z_0 + \sum_{\gamma \in \mathcal{J}} \phi(\gamma) N_t^\gamma = Z_0 + \phi \odot \mathbf{N}_t.$$

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Existence and uniqueness of BDS equation

- ▶ Driving $p(p+1)$ independent Poisson measures $\mathbf{Q} = (Q^\gamma)_{\gamma \in \mathcal{J}}$.
- ▶ **BDS multivariate differential system:**

$$\mathbf{N}_t = \int_0^t \int_{\mathbb{R}^+} \mathbb{1}_{\{\theta \leq \mu(s, Z_{s-})\}} \mathbf{Q}(ds, d\theta), \quad Z_t = Z_0 + \phi \odot \mathbf{N}_t. \quad (2)$$

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- ▶ **Idea:** control birth part \mathbf{N}^b of \mathbf{N} :

$$\mu^b(\omega, t, z) \leq k_t \mathbf{g}(z^{\mathfrak{h}}), \quad (3)$$

with $(k_t) \in \mathcal{P}(\mathcal{G}_t)$ locally bounded and \mathbf{g} verifying $\sum_{n \geq 1} \frac{1}{\mathbf{g}^{\mathfrak{h}}(n)} = \infty$.

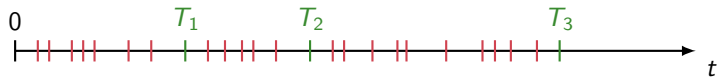
Proposition

*There exists a unique well-defined solution \mathbf{N} of (2), **strongly dominated** by a multivariate counting process \mathbf{G} : $\mathbf{G} - \mathbf{N}$ is a multivariate counting process.*

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BDS systems with fast swap events

Hyp: intensity of swap events $\sim O(\frac{1}{\epsilon}) \gg$ demographic events $\sim O(1)$.



The BDS system now depends on a small parameter ϵ :

$$Z_t^\epsilon = Z_0 + \phi^s \odot \mathbf{N}_t^{s,\epsilon} + \mathbf{N}_t^{b,\epsilon} - \mathbf{N}_t^{d,\epsilon}, \quad (4)$$

$$d\mathbf{N}_t^{s,\epsilon} = \mathbf{Q}^s(dt, [0, \frac{1}{\epsilon} \boldsymbol{\mu}^s(t, Z_{t-}^\epsilon)]), \quad d\mathbf{N}_t^{\text{dem},\epsilon} = \mathbf{Q}^{\text{dem}}(dt, [0, \boldsymbol{\mu}^{\text{dem}}(t, Z_{t-}^\epsilon)]).$$

- ▶ $\mathbf{N}^{s,\epsilon}$ is a "fast" counting system of intensity functional $\frac{1}{\epsilon} \boldsymbol{\mu}^s(t, z)$: explosion when $\epsilon \rightarrow 0$.
- ▶ **But** $\mathbf{N}^{\text{dem},\epsilon}$ only depends on ϵ through Z^ϵ and is strongly dominated by a multivariate counting process which **doesn't depend on ϵ**

$$\forall \epsilon > 0, \quad \mathbf{N}^{\text{dem},\epsilon} < \mathbf{G}^{\text{dem}}.$$

- ▶ **Aggregated process:**

$$Z_t^{h,\epsilon} = Z_0^h + N_t^{b,h,\epsilon} - N_t^{d,h,\epsilon} = F(Z_0, \mathbf{N}_t^{\text{dem},\epsilon})$$

- ▶ **Strong domination** $\Rightarrow (\mathbf{N}^{\text{dem},\epsilon})$ is tight in $\mathcal{A}^{2p} \subset D(\mathbb{R}^+, \mathbb{N}^{2p})$.

Identification of limit points of $(\mathbf{N}^{\text{dem},\epsilon})$

- ▶ \mathcal{G}_t -local martingale $\mathbf{N}_t^{\text{dem},\epsilon} - \int_0^t \boldsymbol{\mu}^{\text{dem}}(\omega, s, Z_{s^-}^\epsilon) ds$.
- ▶ Deterministic intensity functional (Markov framework) \Rightarrow Averaging result of Kurtz (1992).
- ▶ Here: $\boldsymbol{\mu}^{\text{dem}}(\omega, t, z)$. Need convergence of random functionals preserving martingale properties \Rightarrow **Stable convergence**
 - \hookrightarrow Averaging result for stable limits of $\mathbf{N}^{\text{dem},\epsilon}$.

Particular case: deterministic swap intensity function $\mu^s(z)$.

$$Z_t^\epsilon = Z_0 + \phi^s \odot \mathbf{N}_t^{s,\epsilon} + \mathbf{N}_t^{b,\epsilon} - \mathbf{N}_t^{d,\epsilon},$$

$$d\mathbf{N}_t^{s,\epsilon} = \mathbf{Q}^s(dt,]0, \frac{1}{\epsilon} \mu^s(Z_{t-}^\epsilon)], \quad d\mathbf{N}_t^{\text{dem},\epsilon} = \mathbf{Q}^{\text{dem}}(dt,]0, \mu^{\text{dem}}(\omega, t, Z_{t-}^\epsilon)].$$

- ▶ **Pure swap processes** X of \mathcal{G}_t -intensity μ^s :
 - Population with **NO** demographic events.
 - **Constant size:** $X_0^{\natural} = d \Rightarrow X_t = d$ ($X \in \mathcal{U}_d$, populations of size d).
- ▶ Deterministic intensity $\mu^s \Rightarrow X$ **continuous time Markov chain**.

Assumption: $\forall d \in \mathbb{N}$, the swap process restricted to \mathcal{U}_d admits a unique stationary distribution $(\pi(d, dx))_{x \in \mathcal{U}_d}$.

Convergence of the demographic system

Hyp: pure swap on \mathcal{U}_d admits a unique stationary distribution $(\pi(d, dx))_{x \in \mathcal{U}_d}$.

- ▶ Aggregated process $(Z^{\epsilon, \natural})$ birth and death intensities:

$$\mu^{b, \natural}(t, Z_t) = \sum_{i=1}^p \mu^{b, i}(t, Z_t), \quad \mu^{d, \natural}(t, Z_t) = \sum_{i=1}^p \mu^{d, i}(t, Z_t)$$

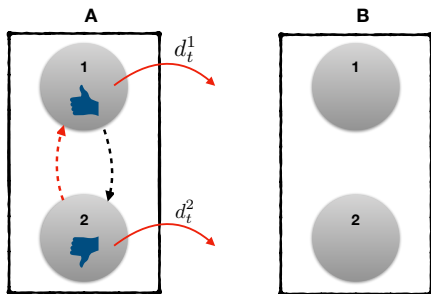
Theorem

The aggregated processes $Z^{\epsilon, \natural}$ converge to the true *Birth-Death process* \bar{Z}^{\natural} of intensity:

$$\lambda^b(t, \bar{Z}_t^{\natural}) = \int_{\mathcal{U}_{Z_t^{\natural}}} \mu^{b, \natural}(t, z) \pi(\bar{Z}_t^{\natural}, dz), \quad \lambda^d(t, \bar{Z}_t^{\natural}) = \int_{\mathcal{U}_{Z_t^{\natural}}} \mu^{d, \natural}(t, z) \pi(\bar{Z}_t^{\natural}, dz).$$

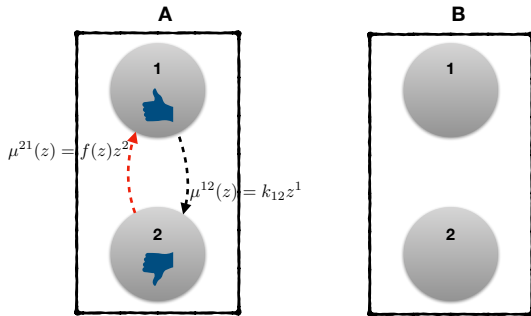
- ▶ **Averaging effect:** aggregated intensities depend *non-linearly* of the number of individuals in the population.

A toy example (I)

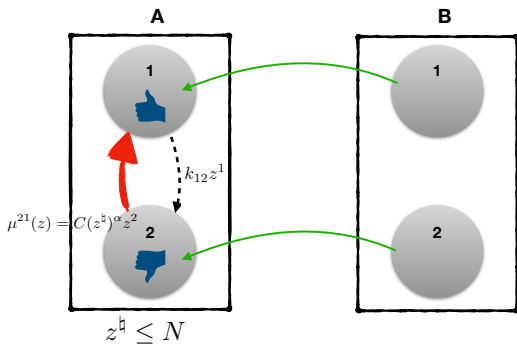


- ▶ Linear death functionals: $\mu^{d,i}(t, Z_t) = d_t^i Z_t^i$, $d_t^1 \leq d_t^2$
(Aggregated death intensity) $\mu^{d,b}(t, Z_t) = d_t^1 Z_t^1 + d_t^2 Z_t^2$.
- ▶ If $Z_t^b = n$, individual death rate is $\frac{\mu^{d,b}(t, Z_t)}{n}$.

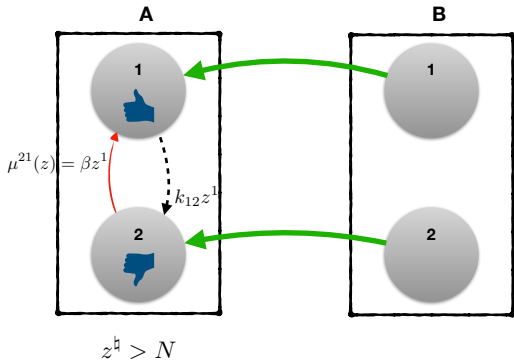
A toy example (II)



A toy example (II)



A toy example (II)



A toy example (II)

- ▶ Individual death rate in the limit aggregated population:

$$d(t, \bar{Z}_t^h) = \frac{1}{\bar{Z}_t^h} \lambda^d(t, \bar{Z}_t^h) = \frac{1}{\bar{Z}_t^h} (d_t^1 \pi(n, z^1) + d_t^2 \pi(n, z^2)).$$

- ▶ Aggregated death rate depend **non-linearly** on the **population size n** :

- Small population ($n \leq N$):

$$d(t, n) = d_t^1 \frac{C n^\alpha}{k_{12} + C n^\alpha} + d_t^2 \frac{k_{12}}{k_{12} + C n^\alpha}$$

- Large population ($n > N$):

$$d(t, n) = d_t^1 \frac{\beta}{\beta + k_{12}} + d_t^2 \frac{k_{12}}{k_{12} + \beta}$$



Thank You For Your Attention