

# Continuous time Markov Chains: construction and basic tools

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# Outline

## Construction

Finite

Infinite

Interacting Particle Systems

Partial overview

## Tools

Kolmogorov equations

Martingales

Tightness

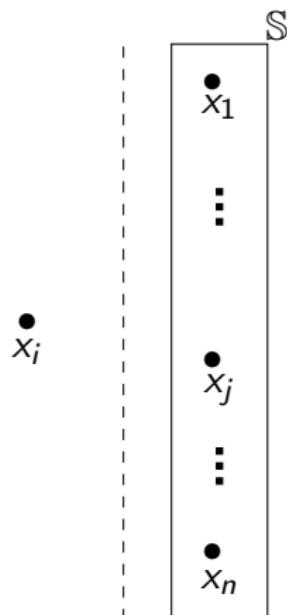
Martingale problems

Panorama

# Finite state spaces

$$\mathbb{S} = \{x_1, \dots, x_n\}$$

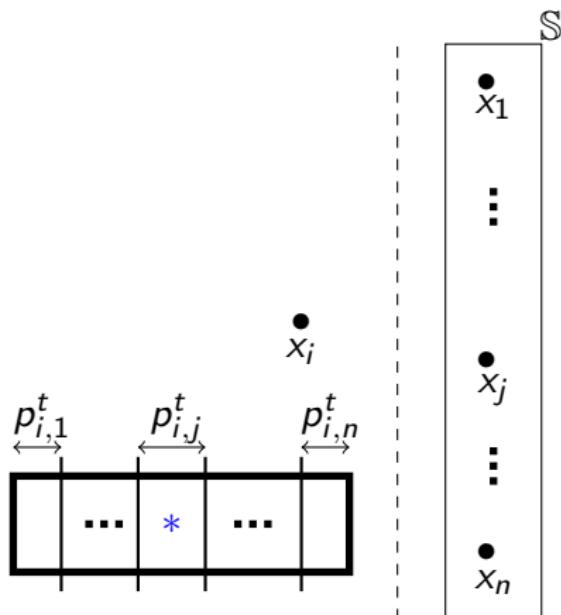
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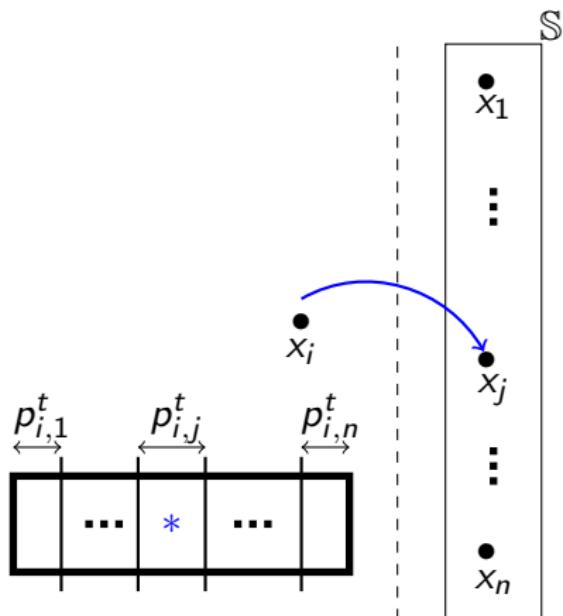
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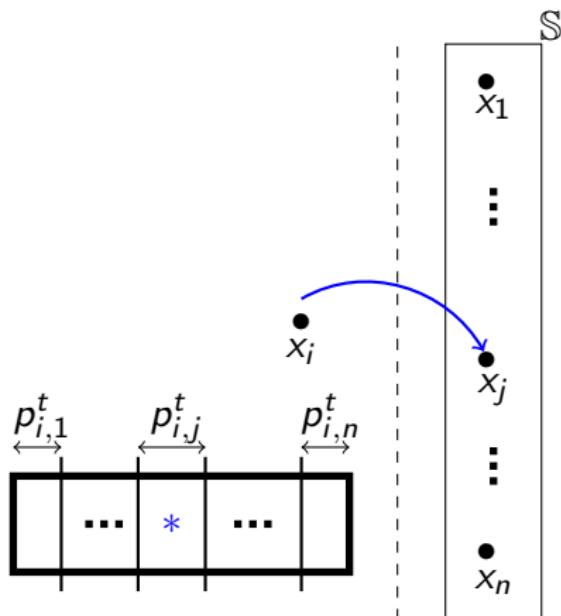
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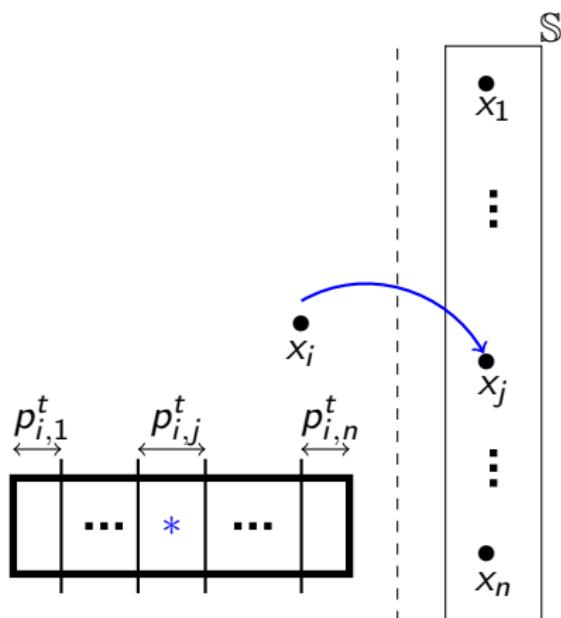
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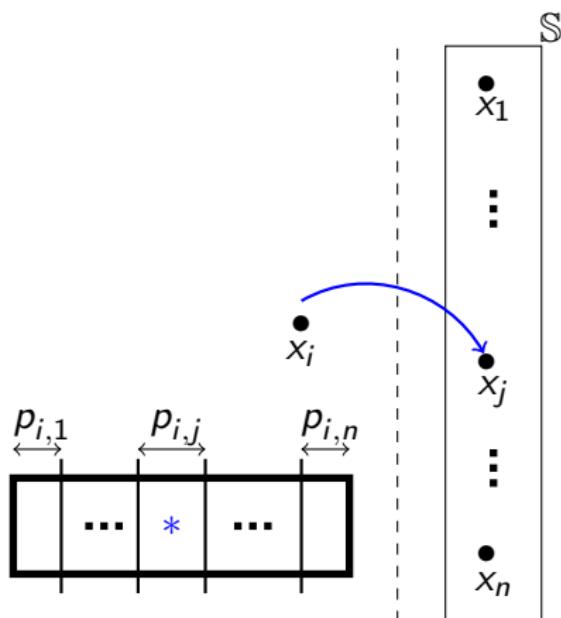
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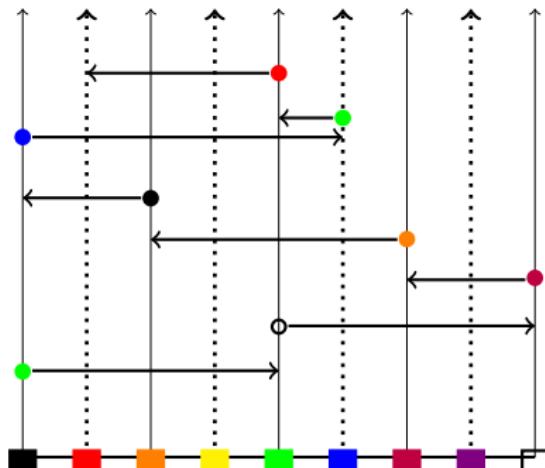
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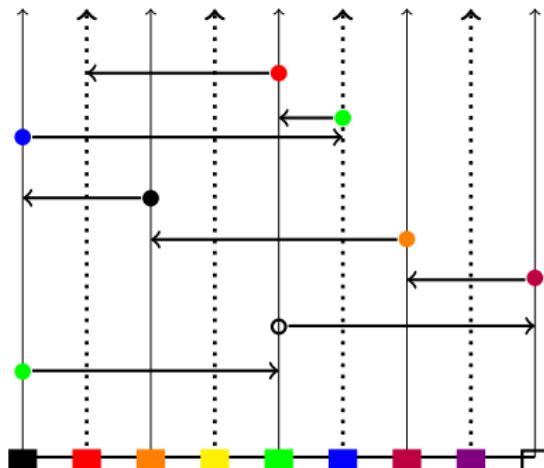
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Poisson point processes + colouring = Graphical construction



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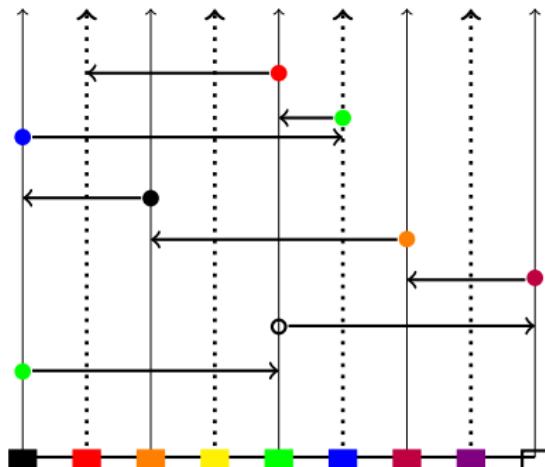
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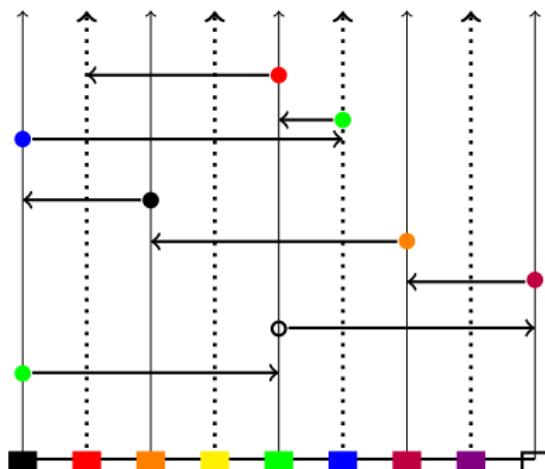


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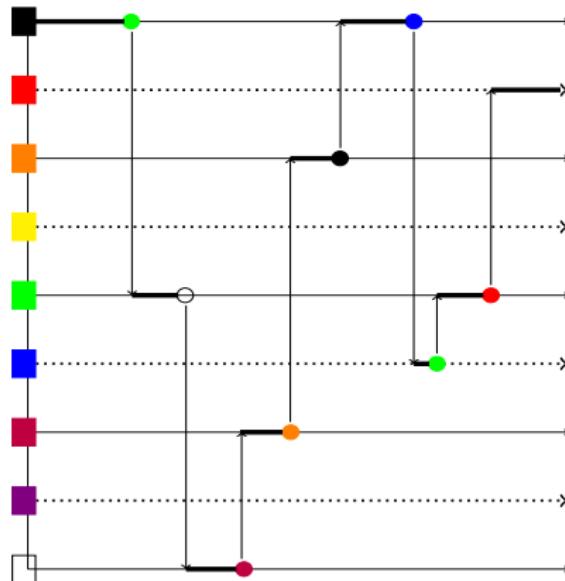
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To complete the construction: extension and regularization.

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This is the case of Interacting Particle Systems (IPS)

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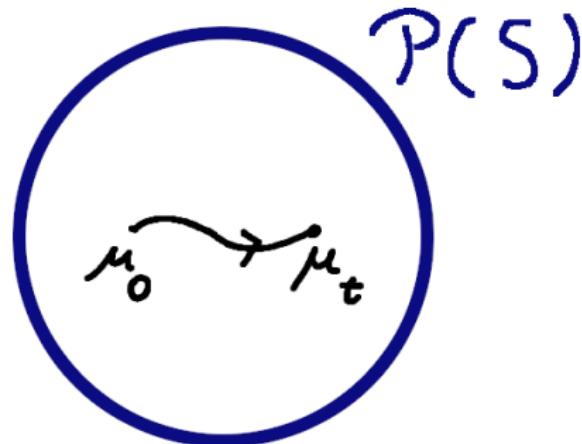
$V = \mathbb{Z}$ : uncountable state space

# Overview

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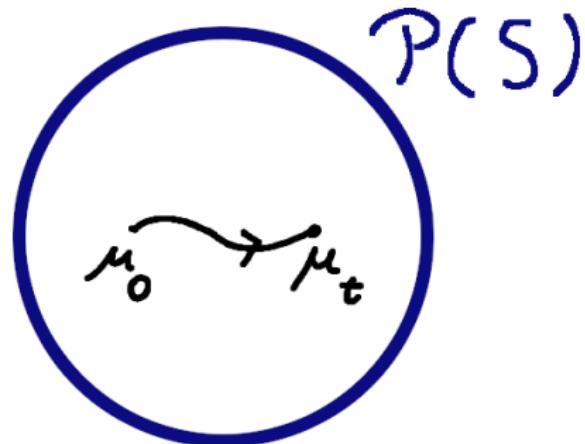
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Goal: To understand the behavior of  $\mu_t$  in the relevant scales.

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$$\begin{cases} \partial_t P(s, x; t, \bullet) = L^* P(s, x; t, \bullet) \\ \lim_{t \downarrow s} P(s, x; t, \bullet) = \delta_x(\bullet) \end{cases}$$

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$$F : \mathbb{R}_+ \times S \rightarrow \mathbb{R} \quad \sup_{(s,x)} |\partial_s^j F(s,x)| < C$$

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# Tightness criterion

Family of processes  $(X^N, N \in \mathbb{N})$

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Then there is a subsequence  $\{N_k, k \in \mathbb{N}\}$  and  $X^* \in D$  s.t.

$$X^{N_k} \rightarrow X^*$$

# Martingale problems

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A continuous adapted process  $(M_t, t \geq 0)$  is a  $d$ -dimensional Brownian Motion if and only if

$$f(M_t) - f(M_0) - \frac{1}{2} \int_0^t \Delta f(M_s) ds$$

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**Theorem:(Lévy)** If  $(M_t, t \geq 0)$  is a continuous real valued local martingale with

$$\langle M^k, M^j \rangle_t = t\delta_{k,t}$$

then  $M_t$  is a Brownian motion.

## Proof (Lévy)

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$$e^{i\lambda M_t} = e^{i\lambda M_s} + i\lambda \int_s^t e^{i\lambda M_u} dM_u - \frac{\lambda^2}{2} \int_s^t e^{i\lambda M_u} du.$$

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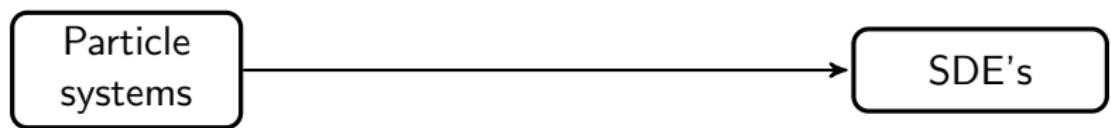
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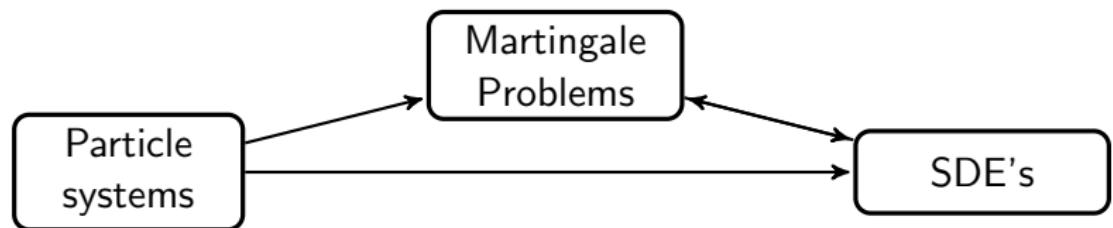
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# Panorama

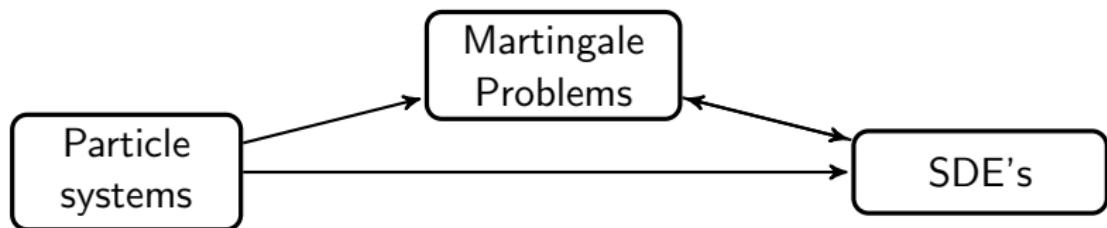


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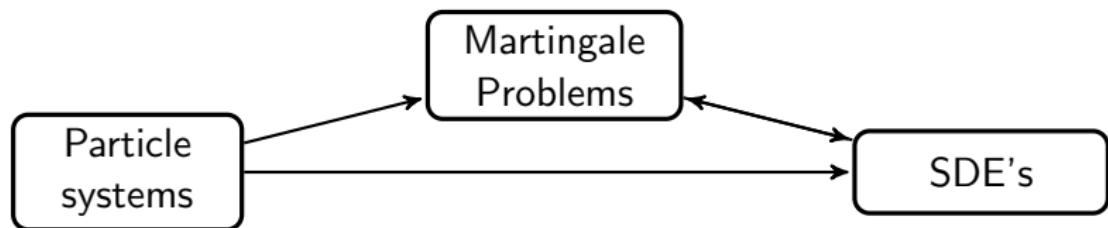
# Panorama

$$X_t^n - X_0^n - \int_0^t L_n(X_s^n) ds = M_t^n$$



# Panorama

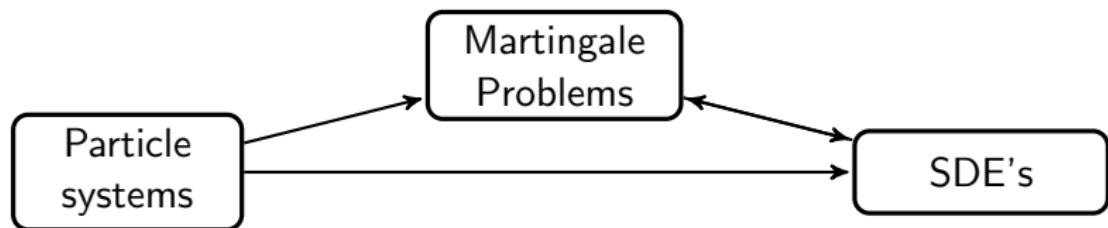
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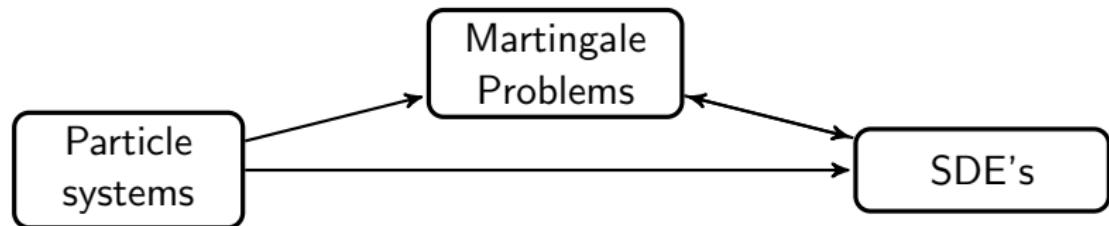
$$M_t = \int_0^t \sigma(X_s) dB_s$$



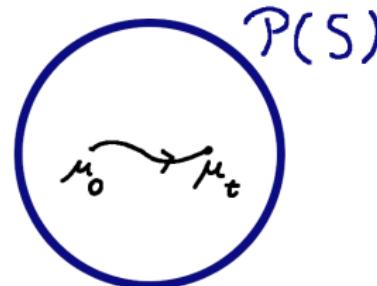
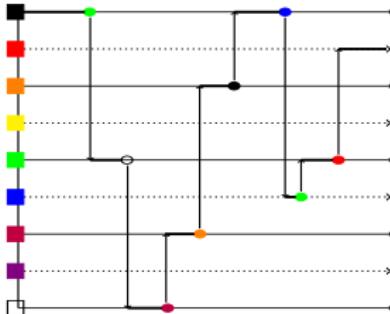
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$$X_t = X_0 + \int_0^t b(X_s) ds + \int_0^t \sigma(X_s) dB_s$$

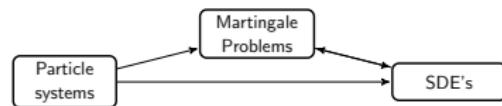


Thank you!

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