

Going beyond mean-field approximations

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Model

Interacting Markov chains, or diffusions, on a **sparse** graph $G = (V, E)$:

$$\begin{aligned} X_v(k+1) &= F(X_v(k), \mu_v(k), \xi_v(k+1)), & k \in \mathbb{N}_0, \\ dX_v(t) &= b(X_v(t), \mu_v(t)) dt + \sigma(X_v(t), \mu_v(t)) dW_v(t), & t \geq 0, \end{aligned}$$

where $\mu_v(k)$ is the neighborhood empirical measure given by

$$\mu_v(k) = \frac{1}{|N_v(G)|} \sum_{u \in N_v(G)} \delta_{X_u(k)},$$

and $N_v(G) := \{u \in V : (u, v) \in E\}$ denotes the neighborhood of v .

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Results to discuss:

- (a) **2nd order Markov random fields** \longrightarrow local-field equations
- (b) **Correlation decay** \longrightarrow convergence of empirical measures
- (c) **Dense graphs with heterogeneous limits** \longrightarrow graphon particle systems

2nd-order Markov random fields

Definition: A family of random variables $(Y_v)_{v \in G}$ is a **2nd-order Markov random field** if

$$(Y_v)_{v \in A} \perp (Y_v)_{v \in B} \mid (Y_v)_{v \in \partial^2 A},$$

for all sets $A, B \subset V$ with $B \cap (A \cup \partial^2 A) = \emptyset$.

Example:



Notation: For a set A of vertices in a graph $G = (V, E)$, define

$$\text{Boundary: } \partial A = \{u \in V : d(u, A) = 1\},$$

$$\text{Double-boundary: } \partial^2 A = \{u \in V : d(u, A) = 1, 2\} = \partial A \cup \partial(A \cup \partial A).$$

2nd-order Markov random fields

Theorem (Lacker-Ramanan-W. '25)

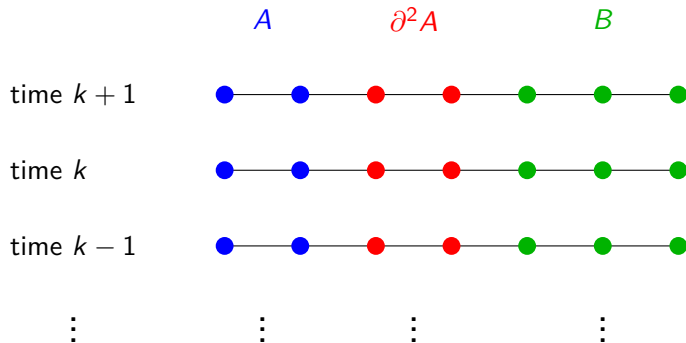
Assume the initial states $(X_v(0))_{v \in V}$ form a second-order MRF. For each time t , the particle *trajectories* $(X_v[k])_{v \in V}$ form a *second-order* MRF. Here $x[k] = (x(0), x(1), \dots, x(k))$.

This is sharp: in general, for $k \geq 1$,

- $(X_v[k])_{v \in V}$ is *not* a first-order MRF.
- $(X_v(k))_{v \in V}$ is *not* a MRF of any order.

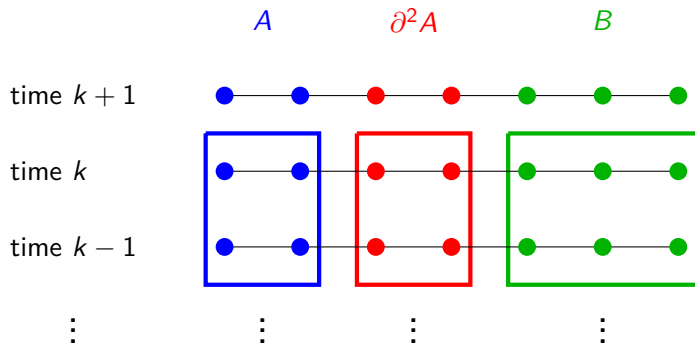
2nd-order Markov random fields

Why? We need to condition on the **paths** of **two nodes in the middle** to cut off the influence between A and B .



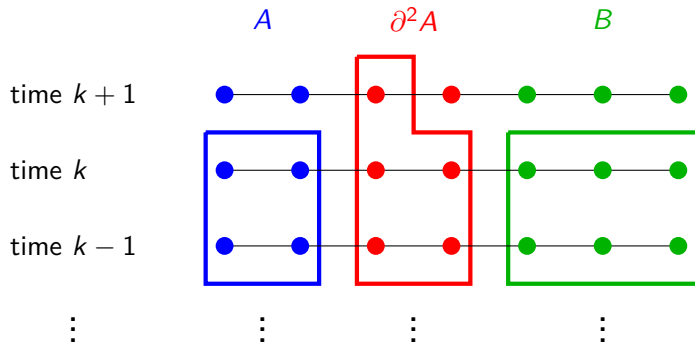
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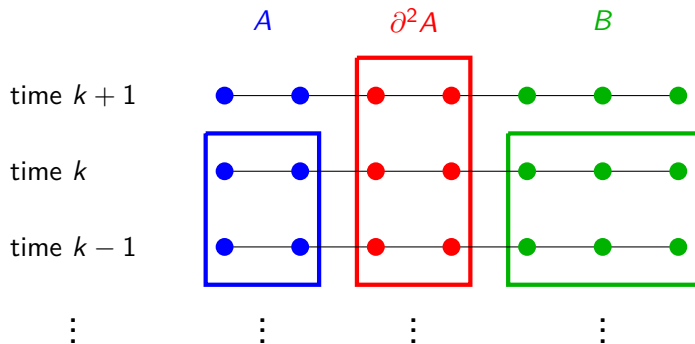
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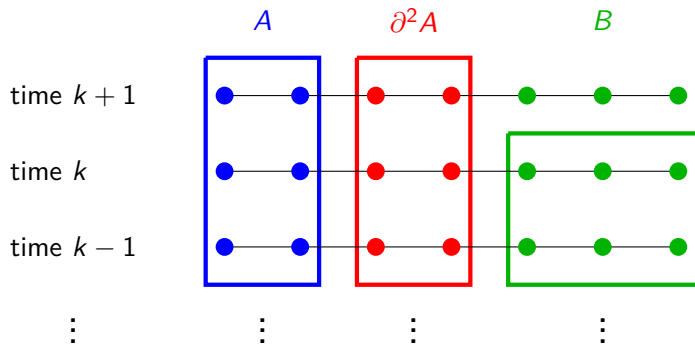
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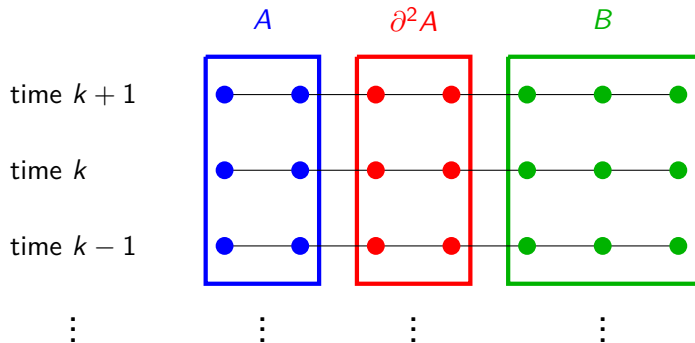
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Correlation decay

Theorem (Lacker-Ramanan-W. '23)

Under suitable conditions, there exists $c: \mathbb{N}_0 \rightarrow \mathbb{R}_+$ with $\lim_{n \rightarrow \infty} c(n) = 0$ such that

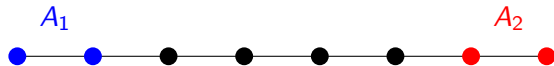
$$|\text{Cov}(f_1(X_{A_1}[T]), f_2(X_{A_2}[T]))| \leq (|A_1| + |A_2|) c(d_G(A_1, A_2))$$

for all functions f_1, f_2 that are 1-Lipschitz and bounded by 1.

Essential for proofs of convergence of empirical measures.

Correlation decay

Argue via coupling: $\text{Cov}(X_{A_1}, X_{A_2}) = \mathbb{E}[X_{A_1} X_{A_2}] - \mathbb{E}[X_{A_1}] \mathbb{E}[X_{A_2}]$



Will construct Y and Z such that

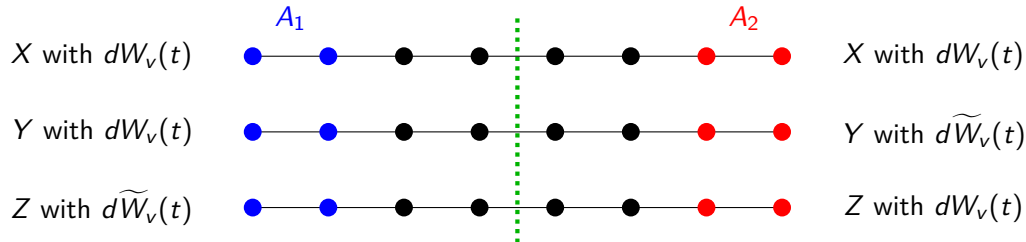
- Y and Z are independent;
- $Y \stackrel{d}{=} X \stackrel{d}{=} Z$;
- $\mathbb{E}|Y_{A_1} - X_{A_1}|$ and $\mathbb{E}|Z_{A_2} - X_{A_2}|$ decay as $d_G(A_1, A_2)$ grows.

Then

$$\mathbb{E}[X_{A_1} X_{A_2}] - \mathbb{E}[X_{A_1}] \mathbb{E}[X_{A_2}] = \mathbb{E}[X_{A_1} X_{A_2}] - \mathbb{E}[Y_{A_1} Z_{A_2}] \leq C \mathbb{E}|Y_{A_1} - X_{A_1}| + C \mathbb{E}|Z_{A_2} - X_{A_2}|.$$

Correlation decay

Replace some of the driving Brownian motion W by independent copies \widetilde{W} :



By construction: y and Z are independent; $y \stackrel{d}{=} X \stackrel{d}{=} Z$. Similar to Picard iteration estimates:

$$\max_{v: d(v, A_1) \leq k} \mathbb{E} [\|X_v - Y_v\|_{*,t}^2] \leq C \int_0^t \max_{v: d(v, A_1) \leq k+1} \mathbb{E} [\|X_v - Y_v\|_{*,s}^2] ds$$

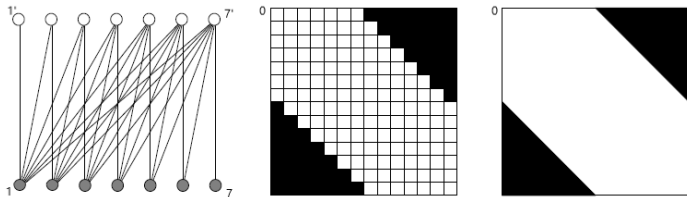
$$\implies \max_{v \in A_1} \mathbb{E} [\|X_v - Y_v\|_{*,t}^2] \leq (Ct)^L / L!, \quad L = \lceil d_G(A_1, A_2) / 2 \rceil.$$

Dense graphs \rightarrow graphon

Particles with **heterogeneous interaction** $\xi_{ij}^n \in [0, 1]$ corresponding to $G_n: [0, 1]^2 \rightarrow [0, 1]$

$$X_i^n(t) = x_0 + \int_0^t \frac{1}{n} \sum_{j=1}^n \xi_{ij}^n b(X_i^n(s), X_j^n(s)) ds + W_i(t).$$

Example: Half-graph and its limit.



Limit of half-graphs in the
cut norm $\|\cdot\|_{\square}$.
(Lovász '12)

The cut metric is weak: $\|G\|_{\square} \leq \|G\|_{L_1} \leq \|G\|_{L_2}$, but convenient for convergence of graphs.

E.g., $G_n = \text{Erdős-Rényi}(n, \frac{1}{2}) \rightarrow G \equiv \frac{1}{2}$ **in** $\|\cdot\|_{\square}$ **but not in** L_p , since $|G_n(u, v) - G(u, v)| \equiv \frac{1}{2}$.

Dense graphs \rightarrow graphon

We expect and can show

$$X_i^n(t) = x_0 + \int_0^t \frac{1}{n} \sum_{j=1}^n \xi_{ij}^n b(X_i^n(s), X_j^n(s)) ds + B_{\frac{i}{n}}(t),$$

$$\begin{aligned} \Rightarrow \quad X_u(t) &= x_0 + \int_0^t \int_{[0,1]} \int_{\mathbb{R}^d} b(X_u(s), x) G(u, v) \mu_{v,s}(dx) dv ds + B_u(t), \\ \mu_{v,t} &= \mathcal{L}(X_v(t)), \quad v \in [0, 1]. \end{aligned}$$

Theorem (Bayraktar-Chakraborty-W. '23)

Under conditions, $\mu^n := \frac{1}{n} \sum_{i=1}^n \delta_{X_i^n} \rightarrow \bar{\mu} := \int_0^1 \mathcal{L}(X_u) du$ in $\mathcal{P}(\mathcal{C}_d)$ in probability as $n \rightarrow \infty$.

No continuity of the limiting graph G is needed.

Dense graphs \rightarrow graphon

Graphon particle systems are also analyzed for ...

- jump processes
 - SIR model with heterogeneous connection
 - JSQ- d with dispatchers and servers connected via heterogeneous bipartite graphs
- games
- control problems
- etc.

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