

# Annealed limit and quenched control for a diffusive disordered mean-field model with random jumps

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  - Point processes
  - Semigroup and generator
- 2 Model
  - Neural networks model
  - Definitions of the systems
  - Heuristics
- 3 Convergence
  - Result
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# Point process : definitions

## Point process (or counting process) $Z$ :

- a random countable set of  $\mathbb{R}_+$  :  $Z = \{T_i : i \in \mathbb{N}\}$
- a random point measure on  $\mathbb{R}_+$  :  $Z = \sum_{i \in \mathbb{N}} \delta_{T_i}$

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A process  $\lambda$  is the **stochastic intensity** of  $Z$  if :

$$\forall 0 \leq a < b, \mathbb{E} [Z([a, b]) | \mathcal{F}_a] = \mathbb{E} \left[ \int_a^b \lambda_t dt \middle| \mathcal{F}_a \right]$$

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## Characterization of $P$ by $A$

$$P_t g(x) = g(x) + \int_0^t P_s Ag(x) ds$$



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**Network of  $N$  neurons :**

- $Z_t^{N,i}$  = number of spikes of neuron  $i$  emitted in  $[0, t]$   
= point process with intensity  $f(X_{t-}^{N,i})$
- $X^{N,i}$  = potential of neuron  $i$

# $N$ -neurons network model

$$dX_t^{N,i} = -\alpha X_t^{N,i} dt + \frac{1}{N^\beta} \sum_{j=1}^N u_{ji}(t) dZ_t^{N,j}$$

with :

- $Z^{N,j}$  = **point process** with intensity  $f(X_{t-}^{N,j})$
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- $X_t^{N,i} = X_{t-}^{N,i} + \frac{u^{ji}(t)}{N^\beta}$  if a neuron  $j$  emits a spike at  $t$   
→  $u^{ji}(t)/N^\beta$  = synaptic weight

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- **linear scaling**  $\beta = 1$  (LLN) :  
[Delattre et al. (2016)] (Hawkes process,  $u_{ji}(t) = 1$ ),  
[Chevallier et al. (2019)] and [Agathe-Nerine (2022)]  
( $u_{ji}(t) = w(v_j, v_i)$  random)

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- **diffusive scaling**  $\beta = 1/2$  (CLT)  
→  $u_{ji}(t)$  random and centered

# Diffusive scaling

- **Marked point processes**
  
  
  
  
  
  
  
  
  
  
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**Marked model inconsistency :**

roles of synapses can change at every spike

# Diffusive scaling, random environment, dimension 1

$N$ -particle system  $U_j$  iid centered

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**Reference :** [Pfaffelhuber, Rotter, Stiefel (2022)] 2 differences :

- $X^N$  Hawkes process  $\Rightarrow$  not gentle process
- $\mathcal{L}(U_1) = 1/2\delta_1 + 1/2\delta_{-1} \Rightarrow$  gentle distribution



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**Our assumption :**  $U_j$  iid centered and

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**Limit system**  $W \sim \mathcal{N}(0, \sigma^2)$

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# Heuristics for the limit system

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## CLT coupling from KMT result

## Proposition

Let  $U_j$  ( $j \geq 1$ ) iid centered, for some  $\alpha > 0$ ,

$$\mathbb{E} \left[ e^{\alpha |U_1|} \right] < \infty \text{ and } \sigma^2 := \mathbb{E} [U_1^2]$$

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**Then** there exist  $W^{[M]}$  i.d.  $\sim \mathcal{N}(0, \sigma^2)$  and  $K$  such that :

$$\left| \frac{1}{\sqrt{N}} \sum_{j=1}^N U_j - W^{[M]} \right| \leq K \frac{\ln N}{\sqrt{N}} \text{ and } \mathbb{E} \left[ e^{\gamma K} \right] < \infty \text{ for some } \gamma > 0$$



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**Proof.** consequence of :

- KMT theorem [Komlós, Major, Tusnády (1976)]
- reasoning of [Ethier, Kurtz (2005)]

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**Remark** : same convergence speed for FIDI distribution

# Infinitesimal generator

## $N$ -particle system

$$dX_t^N = b(X_t^N)dt + N^{-1/2} \sum_{j=1}^N U_j \int_0^\infty \mathbb{1}_{\{z \leq f(X_{t-}^N)\}} d\pi^j(t, z)$$

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$$A_{\mathcal{E}}^N g(x) = b(x)g'(x) + f(x) \sum_{j=1}^N \left[ g\left(x + \frac{U_j}{\sqrt{N}}\right) - g(x) \right]$$

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## Difference of generators

$$\begin{aligned} \left| A_{\mathcal{E}}^N g(x) - \bar{A}_{\mathcal{E}}^N g(x) \right| \leq f(x) & \left( \left| \frac{1}{\sqrt{N}} \sum_{j=1}^N U_j - W^{[M]} \right| \cdot |g'(x)| \right. \\ & + \frac{1}{2} \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| \cdot |g''(x)| \\ & \left. + \frac{1}{6N\sqrt{N}} \sum_{j=1}^N |U_j|^3 \cdot \|g'''\|_{\infty} \right) \end{aligned}$$

## Difference of generators

$$\begin{aligned} \left| A_{\mathcal{E}}^N g(x) - \bar{A}_{\mathcal{E}}^N g(x) \right| \leq f(x) & \left( \left| \frac{1}{\sqrt{N}} \sum_{j=1}^N U_j - W^{[M]} \right| \cdot |g'(x)| \right. \\ & + \frac{1}{2} \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| \cdot |g''(x)| \\ & \left. + \frac{1}{6N\sqrt{N}} \sum_{j=1}^N |U_j|^3 \cdot \|g'''\|_{\infty} \right) \end{aligned}$$

## Difference of generators

$$\left| A_{\mathcal{E}}^N g(x) - \bar{A}_{\mathcal{E}}^N g(x) \right| \leq f(x) \|g\|_{3,\infty} \left( \left| \frac{1}{\sqrt{N}} \sum_{j=1}^N U_j - W^{[M]} \right| + \frac{1}{2} \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + \frac{1}{6N\sqrt{N}} \sum_{j=1}^N |U_j|^3 \right)$$

## Difference of generators

$$\begin{aligned} \left| A_{\mathcal{E}}^N g(x) - \bar{A}_{\mathcal{E}}^N g(x) \right| &\leq f(x) \|g\|_{3,\infty} \left( \left| \frac{1}{\sqrt{N}} \sum_{j=1}^N U_j - W^{[M]} \right| \right. \\ &\quad \left. + \frac{1}{2} \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| \right. \\ &\quad \left. + \frac{1}{6N\sqrt{N}} \sum_{j=1}^N |U_j|^3 \right) \\ &\leq \|f\|_{\infty} \|g\|_{3,\infty} \epsilon_N(\mathcal{E}) \end{aligned}$$

# Semigroup

$$\left(\bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N\right) g(x) = \int_0^t P_{\mathcal{E},t-s}^N \left(\bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N\right) \bar{P}_{\mathcal{E},s}^N g(x) ds$$

# Semigroup

$$\left(\bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N\right) g(x) = \int_0^t P_{\mathcal{E},t-s}^N \left(\bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N\right) \bar{P}_{\mathcal{E},s}^N g(x) ds$$

**Sketch of proof :**  $u(s) := P_{\mathcal{E},t-s}^N \bar{P}_{\mathcal{E},s}^N g(x)$



# Semigroup

$$\left(\bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N\right) g(x) = \int_0^t P_{\mathcal{E},t-s}^N \left(\bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N\right) \bar{P}_{\mathcal{E},s}^N g(x) ds$$

**Sketch of proof :**  $u(s) := P_{\mathcal{E},t-s}^N \bar{P}_{\mathcal{E},s}^N g(x)$

$$\left(\bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N\right) g(x) = u(t) - u(0)$$

# Semigroup

$$\left(\bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N\right) g(x) = \int_0^t P_{\mathcal{E},t-s}^N \left(\bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N\right) \bar{P}_{\mathcal{E},s}^N g(x) ds$$

**Sketch of proof :**  $u(s) := P_{\mathcal{E},t-s}^N \bar{P}_{\mathcal{E},s}^N g(x)$

$$\begin{aligned} \left(\bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N\right) g(x) &= u(t) - u(0) \\ &= \int_0^t u'(s) ds \end{aligned}$$

## Semigroup

$$\left(\bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N\right) g(x) = \int_0^t P_{\mathcal{E},t-s}^N \left(\bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N\right) \bar{P}_{\mathcal{E},s}^N g(x) ds$$

**Sketch of proof :**  $u(s) := P_{\mathcal{E},t-s}^N \bar{P}_{\mathcal{E},s}^N g(x)$

$$\left(\bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N\right) g(x) = u(t) - u(0)$$

$$= \int_0^t u'(s) ds$$

$$= \int_0^t \left[ -\frac{d}{du} \left( P_{\mathcal{E},u}^N \bar{P}_{\mathcal{E},s}^N g(x) \right) \Big|_{u=t-s} + \frac{d}{du} \left( P_{\mathcal{E},t-s}^N \bar{P}_{\mathcal{E},u}^N g(x) \right) \Big|_{u=s} \right] ds$$

## Semigroup

$$\left(\bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N\right) g(x) = \int_0^t P_{\mathcal{E},t-s}^N \left(\bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N\right) \bar{P}_{\mathcal{E},s}^N g(x) ds$$

**Sketch of proof :**  $u(s) := P_{\mathcal{E},t-s}^N \bar{P}_{\mathcal{E},s}^N g(x)$

$$\begin{aligned} \left(\bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N\right) g(x) &= u(t) - u(0) \\ &= \int_0^t u'(s) ds \\ &= \int_0^t \left[ -\frac{d}{du} \left( P_{\mathcal{E},u}^N \bar{P}_{\mathcal{E},s}^N g(x) \right) \Big|_{u=t-s} + \frac{d}{du} \left( P_{\mathcal{E},t-s}^N \bar{P}_{\mathcal{E},u}^N g(x) \right) \Big|_{u=s} \right] ds \\ &= \int_0^t \left[ -P_{\mathcal{E},t-s}^N A_{\mathcal{E}}^N \bar{P}_{\mathcal{E},s}^N g(x) + P_{\mathcal{E},t-s}^N \bar{A}_{\mathcal{E}}^N \bar{P}_{\mathcal{E},s}^N g(x) \right] ds \end{aligned}$$

# Semigroup convergence

$$\left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| \leq \int_0^t \left| P_{\mathcal{E},t-s}^N \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(x) \right| ds$$

## Semigroup convergence

$$\begin{aligned} \left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| &\leq \int_0^t \left| P_{\mathcal{E},t-s}^N \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(x) \right| ds \\ &\leq \int_0^t \mathbb{E}_{\mathcal{E}, X_0^N = x} \left[ \left| \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(X_{t-s}^N) \right| \right] ds \end{aligned}$$

## Semigroup convergence

$$\begin{aligned} \left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| &\leq \int_0^t \left| P_{\mathcal{E},t-s}^N \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(x) \right| ds \\ &\leq \int_0^t \mathbb{E}_{\mathcal{E}, X_0^N = x} \left[ \left| \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(X_{t-s}^N) \right| \right] ds \\ &\leq \|f\|_{\infty} \left( \int_0^t \left\| \bar{P}_{\mathcal{E},s}^N g \right\|_{3,\infty} ds \right) \epsilon_N(\mathcal{E}) \end{aligned}$$

## Semigroup convergence

$$\begin{aligned} \left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| &\leq \int_0^t \left| P_{\mathcal{E},t-s}^N \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(x) \right| ds \\ &\leq \int_0^t \mathbb{E}_{\mathcal{E}, X_0^N = x} \left[ \left| \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(X_{t-s}^N) \right| \right] ds \\ &\leq \|f\|_{\infty} \left( \int_0^t \left\| \bar{P}_{\mathcal{E},s}^N g \right\|_{3,\infty} ds \right) \epsilon_N(\mathcal{E}) \end{aligned}$$

Control of  $\left\| \bar{P}_{\mathcal{E},s}^N g(x) \right\|_{3,\infty}$

$$\bar{P}_{\mathcal{E},s}^N g(x) = \mathbb{E}_{\mathcal{E}, \bar{X}_0^N = x} \left[ g(\bar{X}_s^N) \right]$$



## Semigroup convergence

$$\begin{aligned} \left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| &\leq \int_0^t \left| P_{\mathcal{E},t-s}^N \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(x) \right| ds \\ &\leq \int_0^t \mathbb{E}_{\mathcal{E}, X_0^N=x} \left[ \left| \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(X_{t-s}^N) \right| \right] ds \\ &\leq \|f\|_{\infty} \left( \int_0^t \left\| \bar{P}_{\mathcal{E},s}^N g \right\|_{3,\infty} ds \right) \epsilon_N(\mathcal{E}) \end{aligned}$$

Control of  $\left\| \bar{P}_{\mathcal{E},s}^N g(x) \right\|_{3,\infty}$

$$\bar{P}_{\mathcal{E},s}^N g(x) = \mathbb{E}_{\mathcal{E}, \bar{X}_0^N=x} \left[ g(\bar{X}_s^N) \right] = \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_s^N(x)) \right]$$

## Semigroup convergence

$$\begin{aligned} \left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| &\leq \int_0^t \left| P_{\mathcal{E},t-s}^N \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(x) \right| ds \\ &\leq \int_0^t \mathbb{E}_{\mathcal{E}, X_0^N = x} \left[ \left| \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(X_{t-s}^N) \right| \right] ds \\ &\leq \|f\|_{\infty} \left( \int_0^t \left\| \bar{P}_{\mathcal{E},s}^N g \right\|_{3,\infty} ds \right) \epsilon_N(\mathcal{E}) \end{aligned}$$

Control of  $\left\| \bar{P}_{\mathcal{E},s}^N g(x) \right\|_{3,\infty}$

$$\begin{aligned} \bar{P}_{\mathcal{E},s}^N g(x) &= \mathbb{E}_{\mathcal{E}, \bar{X}_0^N = x} \left[ g(\bar{X}_s^N) \right] = \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_s^N(x)) \right] \\ \partial_x \bar{P}_{\mathcal{E},s}^N g(x) &= \mathbb{E}_{\mathcal{E}} \left[ (\partial_x \bar{X}_s^N(x)) g'(\bar{X}_s^N(x)) \right] \end{aligned}$$

## Semigroup convergence

$$\begin{aligned}
\left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| &\leq \int_0^t \left| P_{\mathcal{E},t-s}^N \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(x) \right| ds \\
&\leq \int_0^t \mathbb{E}_{\mathcal{E}, X_0^N=x} \left[ \left| \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(X_{t-s}^N) \right| \right] ds \\
&\leq \|f\|_{\infty} \left( \int_0^t \left\| \bar{P}_{\mathcal{E},s}^N g \right\|_{3,\infty} ds \right) \epsilon_N(\mathcal{E})
\end{aligned}$$

**Control of**  $\left\| \bar{P}_{\mathcal{E},s}^N g(x) \right\|_{3,\infty}$

$$\bar{P}_{\mathcal{E},s}^N g(x) = \mathbb{E}_{\mathcal{E}, \bar{X}_0^N=x} \left[ g(\bar{X}_s^N) \right] = \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_s^N(x)) \right]$$

$$\partial_x \bar{P}_{\mathcal{E},s}^N g(x) = \mathbb{E}_{\mathcal{E}} \left[ (\partial_x \bar{X}_s^N(x)) g'(\bar{X}_s^N(x)) \right]$$

$$|\partial_x \bar{P}_{\mathcal{E},s}^N g(x)| \leq \|g'\|_{\infty} \sup_x \mathbb{E}_{\mathcal{E}} \left[ |\partial_x \bar{X}_s^N(x)| \right]$$

## Semigroup convergence

$$\begin{aligned}
\left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| &\leq \int_0^t \left| P_{\mathcal{E},t-s}^N \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(x) \right| ds \\
&\leq \int_0^t \mathbb{E}_{\mathcal{E}, X_0^N=x} \left[ \left| \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(X_{t-s}^N) \right| \right] ds \\
&\leq \|f\|_{\infty} \left( \int_0^t \left\| \bar{P}_{\mathcal{E},s}^N g \right\|_{3,\infty} ds \right) \epsilon_N(\mathcal{E})
\end{aligned}$$

Control of  $\left\| \bar{P}_{\mathcal{E},s}^N g(x) \right\|_{3,\infty}$

$$\bar{P}_{\mathcal{E},s}^N g(x) = \mathbb{E}_{\mathcal{E}, \bar{X}_0^N=x} \left[ g(\bar{X}_s^N) \right] = \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_s^N(x)) \right]$$

$$\partial_x \bar{P}_{\mathcal{E},s}^N g(x) = \mathbb{E}_{\mathcal{E}} \left[ (\partial_x \bar{X}_s^N(x)) g'(\bar{X}_s^N(x)) \right]$$

$$|\partial_x \bar{P}_{\mathcal{E},s}^N g(x)| \leq \|g'\|_{\infty} \sup_x \mathbb{E}_{\mathcal{E}} \left[ |\partial_x \bar{X}_s^N(x)| \right] \leq C_s \|g'\|_{\infty} e^{C_s |W^M|}$$

## Semigroup convergence

$$\begin{aligned}
\left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| &\leq \int_0^t \left| P_{\mathcal{E},t-s}^N \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(x) \right| ds \\
&\leq \int_0^t \mathbb{E}_{\mathcal{E}, X_0^N=x} \left[ \left| \left( \bar{A}_{\mathcal{E}}^N - A_{\mathcal{E}}^N \right) \bar{P}_{\mathcal{E},s}^N g(X_{t-s}^N) \right| \right] ds \\
&\leq \|f\|_{\infty} \left( \int_0^t \left\| \bar{P}_{\mathcal{E},s}^N g \right\|_{3,\infty} ds \right) \epsilon_N(\mathcal{E}) \\
&\leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})
\end{aligned}$$

Control of  $\left\| \bar{P}_{\mathcal{E},s}^N g(x) \right\|_{3,\infty}$

$$\bar{P}_{\mathcal{E},s}^N g(x) = \mathbb{E}_{\mathcal{E}, \bar{X}_0^N=x} \left[ g(\bar{X}_s^N) \right] = \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_s^N(x)) \right]$$

$$\partial_x \bar{P}_{\mathcal{E},s}^N g(x) = \mathbb{E}_{\mathcal{E}} \left[ (\partial_x \bar{X}_s^N(x)) g'(\bar{X}_s^N(x)) \right]$$

$$|\partial_x \bar{P}_{\mathcal{E},s}^N g(x)| \leq \|g'\|_{\infty} \sup_x \mathbb{E}_{\mathcal{E}} \left[ |\partial_x \bar{X}_s^N(x)| \right] \leq C_s \|g'\|_{\infty} e^{C_s |W^{[M]}|}$$

## Annealed convergence and quenched control

$$\left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

## Annealed convergence and quenched control

$$\left| \left( \bar{P}_{\mathcal{E},t}^N - P_{\mathcal{E},t}^N \right) g(x) \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

## Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$



# Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

## Annealed convergence

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \mathbb{E} \left[ \epsilon_N(\mathcal{E})^2 \right]^{1/2}$$

## Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

**Annealed convergence**

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \mathbb{E} \left[ \epsilon_N(\mathcal{E})^2 \right]^{1/2}$$

$$\text{with } \epsilon_N(\mathcal{E}) \leq C \left( \left| \frac{1}{\sqrt{N}} \sum_{j=1}^N U_j - W^{[M]} \right| + \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + N^{-3/2} \sum_{j=1}^N |U_j|^3 \right)$$

## Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

**Annealed convergence**

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \mathbb{E} \left[ \epsilon_N(\mathcal{E})^2 \right]^{1/2}$$

$$\text{with } \epsilon_N(\mathcal{E}) \leq C \left( \left| \frac{1}{\sqrt{N}} \sum_{j=1}^N U_j - W^{[M]} \right| + \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + N^{-3/2} \sum_{j=1}^N |U_j|^3 \right)$$

## Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

**Annealed convergence**

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \mathbb{E} \left[ \epsilon_N(\mathcal{E})^2 \right]^{1/2}$$

$$\text{with } \epsilon_N(\mathcal{E}) \leq C \left( K \frac{\ln N}{\sqrt{N}} + \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + N^{-3/2} \sum_{j=1}^N |U_j|^3 \right)$$

## Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

**Annealed convergence**

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \mathbb{E} \left[ \epsilon_N(\mathcal{E})^2 \right]^{1/2}$$

$$\text{with } \epsilon_N(\mathcal{E}) \leq C \left( K \frac{\ln N}{\sqrt{N}} + \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + N^{-3/2} \sum_{j=1}^N |U_j|^3 \right)$$

## Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

## Annealed convergence

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \frac{\ln N}{\sqrt{N}}$$

$$\text{with } \epsilon_N(\mathcal{E}) \leq C \left( K \frac{\ln N}{\sqrt{N}} + \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + N^{-3/2} \sum_{j=1}^N |U_j|^3 \right)$$

## Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

## Annealed convergence

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \frac{\ln N}{\sqrt{N}}$$

with  $\epsilon_N(\mathcal{E}) \leq C \left( K \frac{\ln N}{\sqrt{N}} + \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + N^{-3/2} \sum_{j=1}^N |U_j|^3 \right)$

## Quenched control

- $\epsilon_N(\mathcal{E}) = \mathcal{O} \left( \frac{\ln N}{\sqrt{N}} \right) + \mathcal{O} \left( \sqrt{\frac{\ln \ln N}{N}} \right) + \mathcal{O} \left( \frac{1}{\sqrt{N}} \right)$

# Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

## Annealed convergence

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \frac{\ln N}{\sqrt{N}}$$

with  $\epsilon_N(\mathcal{E}) \leq C \left( K \frac{\ln N}{\sqrt{N}} + \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + N^{-3/2} \sum_{j=1}^N |U_j|^3 \right)$

## Quenched control

- $\epsilon_N(\mathcal{E}) = \mathcal{O} \left( \frac{\ln N}{\sqrt{N}} \right) + \mathcal{O} \left( \sqrt{\frac{\ln \ln N}{N}} \right) + \mathcal{O} \left( \frac{1}{\sqrt{N}} \right) = \mathcal{O} \left( \frac{\ln N}{\sqrt{N}} \right)$



## Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

## Annealed convergence

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \frac{\ln N}{\sqrt{N}}$$

with  $\epsilon_N(\mathcal{E}) \leq C \left( K \frac{\ln N}{\sqrt{N}} + \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + N^{-3/2} \sum_{j=1}^N |U_j|^3 \right)$

## Quenched control

- $\epsilon_N(\mathcal{E}) = \mathcal{O} \left( \frac{\ln N}{\sqrt{N}} \right) + \mathcal{O} \left( \sqrt{\frac{\ln \ln N}{N}} \right) + \mathcal{O} \left( \frac{1}{\sqrt{N}} \right) = \mathcal{O} \left( \frac{\ln N}{\sqrt{N}} \right)$
- There exists  $\beta$  BM,  $|W^{[M]}| = |\beta_N|/\sqrt{N} = \mathcal{O}(\sqrt{\ln \ln N})$

## Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

## Annealed convergence

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \frac{\ln N}{\sqrt{N}}$$

with  $\epsilon_N(\mathcal{E}) \leq C \left( K \frac{\ln N}{\sqrt{N}} + \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + N^{-3/2} \sum_{j=1}^N |U_j|^3 \right)$

## Quenched control

- $\epsilon_N(\mathcal{E}) = \mathcal{O} \left( \frac{\ln N}{\sqrt{N}} \right) + \mathcal{O} \left( \sqrt{\frac{\ln \ln N}{N}} \right) + \mathcal{O} \left( \frac{1}{\sqrt{N}} \right) = \mathcal{O} \left( \frac{\ln N}{\sqrt{N}} \right)$
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## Annealed convergence and quenched control

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} e^{C_t |W^{[M]}|} \epsilon_N(\mathcal{E})$$

## Annealed convergence

$$\mathbb{E} \left[ \left| \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] \right| \right] \leq C_{g,t} \frac{\ln N}{\sqrt{N}}$$

with  $\epsilon_N(\mathcal{E}) \leq C \left( K \frac{\ln N}{\sqrt{N}} + \left| \frac{1}{N} \sum_{j=1}^N U_j^2 - \sigma^2 \right| + N^{-3/2} \sum_{j=1}^N |U_j|^3 \right)$

## Quenched control

- $\epsilon_N(\mathcal{E}) = \mathcal{O} \left( \frac{\ln N}{\sqrt{N}} \right) + \mathcal{O} \left( \sqrt{\frac{\ln \ln N}{N}} \right) + \mathcal{O} \left( \frac{1}{\sqrt{N}} \right) = \mathcal{O} \left( \frac{\ln N}{\sqrt{N}} \right)$
- There exists  $\beta$  BM,  $|W^{[M]}| = |\beta_N|/\sqrt{N} = \mathcal{O}(\sqrt{\ln \ln N})$

$$\left| \mathbb{E}_{\mathcal{E}} \left[ g(\bar{X}_t^N) \right] - \mathbb{E}_{\mathcal{E}} \left[ g(X_t^N) \right] \right| \leq C_{g,t} \mathcal{O} \left( (\ln N)^{C_t} \cdot (\ln N) / \sqrt{N} \right)$$

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Thank you for your attention !

Questions ?