

Lower bound on spatial asymptotic of parabolic Anderson model

Fei Pu (Beijing Normal University)

July 23, 2024

Parabolic Anderson model

$$\begin{cases} \partial_t u(t, x) = \frac{1}{2} \partial_x^2 u(t, x) + u(t, x) \xi(t, x), & t > 0, x \in \mathbb{R}, \\ u(0, \cdot) \equiv 1, \end{cases} \quad (\text{PAM})$$

where ξ denotes space-time white noise.

Parabolic Anderson model

$$\begin{cases} \partial_t u(t, x) = \frac{1}{2} \partial_x^2 u(t, x) + u(t, x) \xi(t, x), & t > 0, x \in \mathbb{R}, \\ u(0, \cdot) \equiv 1, \end{cases} \quad (\text{PAM})$$

where ξ denotes space-time white noise.

- Integral equation

$$u(t, x) = 1 + \int_0^t \int_{\mathbb{R}} p_{t-s}(x-y) u(s, y) \xi(ds dy),$$

with $p_t(x) = (2\pi t)^{-1/2} e^{-x^2/(2t)}$.

Parabolic Anderson model

$$\begin{cases} \partial_t u(t, x) = \frac{1}{2} \partial_x^2 u(t, x) + u(t, x) \xi(t, x), & t > 0, x \in \mathbb{R}, \\ u(0, \cdot) \equiv 1, \end{cases} \quad (\text{PAM})$$

where ξ denotes space-time white noise.

- Integral equation

$$u(t, x) = 1 + \int_0^t \int_{\mathbb{R}} p_{t-s}(x-y) u(s, y) \xi(ds dy),$$

with $p_t(x) = (2\pi t)^{-1/2} e^{-x^2/(2t)}$.

- Conus-Joseph-Khoshnevisan-2013: there exist positive constants c_1, c_2 such that a.s.

$$c_1 \leq \liminf_{N \rightarrow \infty} \frac{\max_{0 \leq x \leq N} \log u(t, x)}{(\log N)^{2/3}} \leq \limsup_{N \rightarrow \infty} \frac{\max_{0 \leq x \leq N} \log u(t, x)}{(\log N)^{2/3}} \leq C_2.$$

Localization

$$u(t,x) = 1 + \int_0^t \int_{\mathbb{R}} p_{t-s}(x-y) u(s,y) \xi(ds dy).$$

Localization

$$u(t, x) = 1 + \int_0^t \int_{\mathbb{R}} p_{t-s}(x-y) u(s, y) \xi(ds dy).$$

- Fix $\beta > 0$. Let $\{U^{(\beta)}(t, x) : t \geq 0, x \in \mathbb{R}\}$ solve

$$U^{(\beta)}(t, x) = 1 + \int_0^t \int_{x-\sqrt{\beta}t}^{x+\sqrt{\beta}t} p_{t-s}(x-y) U^{(\beta)}(s, y) \xi(ds dy).$$

Localization

$$u(t, x) = 1 + \int_0^t \int_{\mathbb{R}} p_{t-s}(x-y) u(s, y) \xi(ds dy).$$

- Fix $\beta > 0$. Let $\{U^{(\beta)}(t, x) : t \geq 0, x \in \mathbb{R}\}$ solve

$$U^{(\beta)}(t, x) = 1 + \int_0^t \int_{x-\sqrt{\beta}t}^{x+\sqrt{\beta}t} p_{t-s}(x-y) U^{(\beta)}(s, y) \xi(ds dy).$$

- Let $U^{(\beta,n)}(t, x)$ be the n -th Picard-iteration approximation to $U^{(\beta)}(t, x)$:

$$\begin{cases} U^{(\beta,n+1)}(t, x) = 1 + \int_0^t \int_{x-\sqrt{\beta}t}^{x+\sqrt{\beta}t} p_{t-s}(x-y) U^{(\beta,n)}(s, y) \xi(ds dy), \\ U^{(\beta,1)}(t, x) \equiv 1. \end{cases}$$

Localization

$$u(t, x) = 1 + \int_0^t \int_{\mathbb{R}} p_{t-s}(x-y) u(s, y) \xi(ds dy).$$

- Fix $\beta > 0$. Let $\{U^{(\beta)}(t, x) : t \geq 0, x \in \mathbb{R}\}$ solve

$$U^{(\beta)}(t, x) = 1 + \int_0^t \int_{x-\sqrt{\beta}t}^{x+\sqrt{\beta}t} p_{t-s}(x-y) U^{(\beta)}(s, y) \xi(ds dy).$$

- Let $U^{(\beta,n)}(t, x)$ be the n -th Picard-iteration approximation to $U^{(\beta)}(t, x)$:

$$\begin{cases} U^{(\beta,n+1)}(t, x) = 1 + \int_0^t \int_{x-\sqrt{\beta}t}^{x+\sqrt{\beta}t} p_{t-s}(x-y) U^{(\beta,n)}(s, y) \xi(ds dy), \\ U^{(\beta,1)}(t, x) \equiv 1. \end{cases}$$

- $u(t, x) \asymp U^{(\beta, [\log \beta])}(t, x)$ when β is large. Fix $t, \beta > 0$. $U^{(\beta, [\log \beta])}(t, x)$ and $U^{(\beta, [\log \beta])}(t, y)$ are independent if $|x - y| \geq \sqrt{\beta}t \log \beta$.

Localization

$$u(t, x) = 1 + \int_0^t \int_{\mathbb{R}} p_{t-s}(x-y) u(s, y) \xi(ds dy).$$

- Fix $\beta > 0$. Let $\{U^{(\beta)}(t, x) : t \geq 0, x \in \mathbb{R}\}$ solve

$$U^{(\beta)}(t, x) = 1 + \int_0^t \int_{x-\sqrt{\beta}t}^{x+\sqrt{\beta}t} p_{t-s}(x-y) U^{(\beta)}(s, y) \xi(ds dy).$$

- Let $U^{(\beta,n)}(t, x)$ be the n -th Picard-iteration approximation to $U^{(\beta)}(t, x)$:

$$\begin{cases} U^{(\beta,n+1)}(t, x) = 1 + \int_0^t \int_{x-\sqrt{\beta}t}^{x+\sqrt{\beta}t} p_{t-s}(x-y) U^{(\beta,n)}(s, y) \xi(ds dy), \\ U^{(\beta,1)}(t, x) \equiv 1. \end{cases}$$

- $u(t, x) \asymp U^{(\beta, [\log \beta])}(t, x)$ when β is large. Fix $t, \beta > 0$. $U^{(\beta, [\log \beta])}(t, x)$ and $U^{(\beta, [\log \beta])}(t, y)$ are independent if $|x - y| \geq \sqrt{\beta}t \log \beta$.

- Chen-2016

$$\lim_{N \rightarrow \infty} \frac{\max_{0 \leq x \leq N} \log u(t, x)}{(\log N)^{2/3}} = \frac{3}{4} \left(\frac{2t}{3} \right)^{1/3}, \quad \text{a.s.}$$

One of the key ingredients is

$$\lim_{m \rightarrow \infty} m^{-3} \log \mathbb{E}[u(t, 0)^m] = \frac{t}{24}.$$

Parabolic Anderson Model with narrow wedge initial data

$$\begin{cases} \partial_t u(t, x) = \frac{1}{2} \partial_x^2 u(t, x) + u(t, x) \xi(t, x), & t > 0, x \in \mathbb{R}, \\ u(0, \cdot) = \delta_0. \end{cases}$$

Parabolic Anderson Model with narrow wedge initial data

$$\begin{cases} \partial_t u(t, x) = \frac{1}{2} \partial_x^2 u(t, x) + u(t, x) \xi(t, x), & t > 0, x \in \mathbb{R}, \\ u(0, \cdot) = \delta_0. \end{cases}$$

- Integral equation

$$u(t, x) = p_t(x) + \int_0^t \int_{\mathbb{R}} p_{t-s}(x-y) u(s, y) \xi(ds dy).$$

Let $U(t, x) = \frac{u(t, x)}{p_t(x)}$. Then

$$U(t, x) = 1 + \int_0^t \int_{\mathbb{R}} p_{s(t-s)/t} \left(y - \frac{s}{t} x \right) U(s, y) \xi(ds dy).$$

Parabolic Anderson Model with narrow wedge initial data

$$\begin{cases} \partial_t u(t, x) = \frac{1}{2} \partial_x^2 u(t, x) + u(t, x) \xi(t, x), & t > 0, x \in \mathbb{R}, \\ u(0, \cdot) = \delta_0. \end{cases}$$

- Integral equation

$$u(t, x) = p_t(x) + \int_0^t \int_{\mathbb{R}} p_{t-s}(x-y) u(s, y) \xi(ds dy).$$

Let $U(t, x) = \frac{u(t, x)}{p_t(x)}$. Then

$$U(t, x) = 1 + \int_0^t \int_{\mathbb{R}} p_{s(t-s)/t} \left(y - \frac{s}{t} x \right) U(s, y) \xi(ds dy).$$

- Huang-Lê-2019 proved

$$\limsup_{N \rightarrow \infty} \frac{\max_{0 \leq x \leq N} \log U(t, x)}{(\log N)^{2/3}} \leq \frac{3}{4} \left(\frac{2t}{3} \right)^{1/3} \quad \text{a.s.}$$

and they conjectured that

$$\lim_{N \rightarrow \infty} \frac{\max_{0 \leq x \leq N} \log U(t, x)}{(\log N)^{2/3}} = \frac{3}{4} \left(\frac{2t}{3} \right)^{1/3} \quad \text{a.s.}$$

Theorem 1. (P.-2023+)

$$\liminf_{N \rightarrow \infty} \frac{\max_{0 \leq x \leq N} \log U(t, x)}{(\log N)^{2/3}} \geq \frac{1}{4} \left(\frac{t}{2} \right)^{1/3} \quad \text{a.s.}$$

Theorem 1. (P.-2023+)

$$\liminf_{N \rightarrow \infty} \frac{\max_{0 \leq x \leq N} \log U(t, x)}{(\log N)^{2/3}} \geq \frac{1}{4} \left(\frac{t}{2} \right)^{1/3} \quad \text{a.s.}$$

- Corwin-Ghosal-2020 give lower and upper bounds on the moments of $U(t, 0)$, which leads to

$$\lim_{m \rightarrow \infty} m^{-3} \log \mathbb{E}[U(t, 0)^m] = \frac{t}{24}.$$

- Ganguly-Hegde-2022 estimate the upper tail probability, which leads to

$$\lim_{\theta \rightarrow +\infty} \frac{\log \mathbb{P}(\log U(t, 0) \geq \theta)}{\theta^{3/2}} = -\frac{4}{3} \sqrt{\frac{2}{t}}.$$

Association

- Chen-Khoshnevisan-Nualart-P.-2023: $\{U(t, x) : t > 0, x \in \mathbb{R}\}$ is associated, i.e.,

$$\text{Cov}(f(U(t_1, x_1), \dots, U(t_n, x_n)), g(U(t_1, x_1), \dots, U(t_n, x_n))) \geq 0, \quad (\text{FKG})$$

for all coordinatewise nondecreasing functions $f, g : \mathbb{R}^n \rightarrow \mathbb{R}$. (Esary-Proshchan-Walkup-1967)

Association

- Chen-Khoshnevisan-Nualart-P.-2023: $\{U(t, x) : t > 0, x \in \mathbb{R}\}$ is associated, i.e.,

$$\text{Cov}(f(U(t_1, x_1), \dots, U(t_n, x_n)), g(U(t_1, x_1), \dots, U(t_n, x_n))) \geq 0, \quad (\text{FKG})$$

for all coordinatewise nondecreasing functions $f, g : \mathbb{R}^n \rightarrow \mathbb{R}$. (Esary-Proshchan-Walkup-1967)

- Using Clark-Ocone formula and Ito's isometry

$$\text{Cov}(f(U(t_1, x_1), \dots, U(t_n, x_n)), g(U(t_1, x_1), \dots, U(t_n, x_n)))$$

$$= \mathbb{E} \left[\int_0^\infty \int_{\mathbb{R}} \mathbb{E}[D_{s,y} f(U(t_1, x_1), \dots, U(t_n, x_n)) | \mathcal{F}_s] \mathbb{E}[D_{s,y} g(U(t_1, x_1), \dots, U(t_n, x_n)) | \mathcal{F}_s] dy ds \right]$$

Association

- Chen-Khoshnevisan-Nualart-P.-2023: $\{U(t, x) : t > 0, x \in \mathbb{R}\}$ is associated, i.e.,

$$\text{Cov}(f(U(t_1, x_1), \dots, U(t_n, x_n)), g(U(t_1, x_1), \dots, U(t_n, x_n))) \geq 0, \quad (\text{FKG})$$

for all coordinatewise nondecreasing functions $f, g : \mathbb{R}^n \rightarrow \mathbb{R}$. (Esary-Proshchan-Walkup-1967)

- Using Clark-Ocone formula and Ito's isometry

$$\begin{aligned} & \text{Cov}(f(U(t_1, x_1), \dots, U(t_n, x_n)), g(U(t_1, x_1), \dots, U(t_n, x_n))) \\ &= E \left[\int_0^\infty \int_{\mathbb{R}} E[D_{s,y} f(U(t_1, x_1), \dots, U(t_n, x_n)) | \mathcal{F}_s] E[D_{s,y} g(U(t_1, x_1), \dots, U(t_n, x_n)) | \mathcal{F}_s] dy ds \right] \\ &= \sum_{j,\ell=1}^n E \left[\int_0^\infty \int_{\mathbb{R}} E[\partial_j f(U(t_1, x_1), \dots, U(t_n, x_n)) \mathbf{D}_{s,y} U(t_j, x_j) | \mathcal{F}_s] \right. \\ & \quad \left. E[\partial_\ell g(U(t_1, x_1), \dots, U(t_n, x_n)) \mathbf{D}_{s,y} U(t_\ell, x_\ell) | \mathcal{F}_s] \right] \\ &\geq 0. \end{aligned}$$

- Lebowitz's inequality (1972)

$$\begin{aligned} & P\{\log U(t, x_1) \leq a_1, \log U(t, x_2) \leq a_2, \dots, \log U(t, x_n) \leq a_n\} - \prod_{j=1}^n P\{\log U(t, x_j) \leq a_j\} \\ & \leq \sum_{1 \leq j < k \leq n} (P\{\log U(t, x_j) \leq a_j, \log U(t, x_k) \leq a_k\} - P\{\log U(t, x_j) \leq a_j\}P\{\log U(t, x_k) \leq a_k\}) \end{aligned}$$

- Lebowitz's inequality (1972)

$$\begin{aligned}
& \mathrm{P}\{\log U(t, x_1) \leq a_1, \log U(t, x_2) \leq a_2, \dots, \log U(t, x_n) \leq a_n\} - \prod_{j=1}^n \mathrm{P}\{\log U(t, x_j) \leq a_j\} \\
& \leq \sum_{1 \leq j < k \leq n} (\mathrm{P}\{\log U(t, x_j) \leq a_j, \log U(t, x_k) \leq a_k\} - \mathrm{P}\{\log U(t, x_j) \leq a_j\} \mathrm{P}\{\log U(t, x_k) \leq a_k\})
\end{aligned}$$

- $\log U(t, 0)$ has a bounded and continuous probability density function by Malliavin calculus.
- By an inequality by Bagai-Prakasa Rao-1991 for associated random variables with bounded pdf,

$$\begin{aligned}
& \sup_{a,b \in \mathbb{R}} (\mathrm{P}\{\log U(t, x) \leq a, \log U(t, y) \leq b\} - \mathrm{P}\{\log U(t, x) \leq a\} \mathrm{P}\{\log U(t, y) \leq b\}) \\
& \leq K [\mathrm{Cov}(\log U(t, x), \log U(t, y))]^{1/3}.
\end{aligned}$$

- Lebowitz's inequality (1972)

$$\begin{aligned}
& \mathrm{P}\{\log U(t, x_1) \leq a_1, \log U(t, x_2) \leq a_2, \dots, \log U(t, x_n) \leq a_n\} - \prod_{j=1}^n \mathrm{P}\{\log U(t, x_j) \leq a_j\} \\
& \leq \sum_{1 \leq j < k \leq n} (\mathrm{P}\{\log U(t, x_j) \leq a_j, \log U(t, x_k) \leq a_k\} - \mathrm{P}\{\log U(t, x_j) \leq a_j\} \mathrm{P}\{\log U(t, x_k) \leq a_k\})
\end{aligned}$$

- $\log U(t, 0)$ has a bounded and continuous probability density function by Malliavin calculus.
- By an inequality by Bagai-Prakasa Rao-1991 for associated random variables with bounded pdf,

$$\begin{aligned}
\sup_{a,b \in \mathbb{R}} (\mathrm{P}\{\log U(t, x) \leq a, \log U(t, y) \leq b\} - \mathrm{P}\{\log U(t, x) \leq a\} \mathrm{P}\{\log U(t, y) \leq b\}) \\
\leq K [\mathrm{Cov}(\log U(t, x), \log U(t, y))]^{1/3}.
\end{aligned}$$

- By Poincaré inequality,

$$\begin{aligned}
\mathrm{Cov}(\log U(t, x), \log U(t, y)) & \leq \int_0^t \int_{\mathbb{R}} \|D_{r,z} \log U(t, x)\|_2 \|D_{r,z} \log U(t, y)\|_2 dz dr \\
& \lesssim \frac{1}{|x-y|}.
\end{aligned}$$

- Lebowitz's inequality (1972)

$$\begin{aligned} & P\{\log U(t, x_1) \leq a_1, \log U(t, x_2) \leq a_2, \dots, \log U(t, x_n) \leq a_n\} - \prod_{j=1}^n P\{\log U(t, x_j) \leq a_j\} \\ & \leq \sum_{1 \leq j < k \leq n} (P\{\log U(t, x_j) \leq a_j, \log U(t, x_k) \leq a_k\} - P\{\log U(t, x_j) \leq a_j\} P\{\log U(t, x_k) \leq a_k\}) \end{aligned}$$

- $\log U(t, 0)$ has a bounded and continuous probability density function by Malliavin calculus.
- By an inequality by Bagai-Prakasa Rao-1991 for associated random variables with bounded pdf,

$$\begin{aligned} \sup_{a,b \in \mathbb{R}} (P\{\log U(t, x) \leq a, \log U(t, y) \leq b\} - P\{\log U(t, x) \leq a\} P\{\log U(t, y) \leq b\}) \\ \leq K [\text{Cov}(\log U(t, x), \log U(t, y))]^{1/3}. \end{aligned}$$

- By Poincaré inequality,

$$\begin{aligned} \text{Cov}(\log U(t, x), \log U(t, y)) & \leq \int_0^t \int_{\mathbb{R}} \|D_{r,z} \log U(t, x)\|_2 \|D_{r,z} \log U(t, y)\|_2 dz dr \\ & \lesssim \frac{1}{|x-y|}. \end{aligned}$$

- Remark: $\text{Cov}(U(t, x), U(t, y)) \asymp \frac{1}{|x-y|}.$

Proof of lower bound

Proof of lower bound

Let $\beta \in (0, \frac{1}{8}\sqrt{\frac{t}{2}})$. Choose and fix $a \in (0, \frac{1}{6})$ and $\epsilon \in (0, 1)$ such that $\beta < \frac{a}{\frac{4}{3}\sqrt{\frac{2}{t}} + \epsilon}$.

Let $x_j = jN/\lfloor N^a \rfloor$ for $j = 1, \dots, \lfloor N^a \rfloor$.

Proof of lower bound

Let $\beta \in (0, \frac{1}{8}\sqrt{\frac{t}{2}})$. Choose and fix $a \in (0, \frac{1}{6})$ and $\epsilon \in (0, 1)$ such that $\beta < \frac{a}{\frac{4}{3}\sqrt{\frac{2}{t}} + \epsilon}$.

Let $x_j = jN/\lfloor N^a \rfloor$ for $j = 1, \dots, \lfloor N^a \rfloor$. Then

$$P \left\{ \max_{0 \leq x \leq N} \log U(t, x) \leq (\beta \log N)^{2/3} \right\} \leq P \left\{ \max_{1 \leq j \leq \lfloor N^a \rfloor} \log U(t, x_j) \leq (\beta \log N)^{2/3} \right\}$$

Proof of lower bound

Let $\beta \in (0, \frac{1}{8}\sqrt{\frac{t}{2}})$. Choose and fix $a \in (0, \frac{1}{6})$ and $\epsilon \in (0, 1)$ such that $\beta < \frac{a}{\frac{4}{3}\sqrt{\frac{2}{t}} + \epsilon}$.

Let $x_j = jN/\lfloor N^a \rfloor$ for $j = 1, \dots, \lfloor N^a \rfloor$. Then

$$\begin{aligned} P \left\{ \max_{0 \leq x \leq N} \log U(t, x) \leq (\beta \log N)^{2/3} \right\} &\leq P \left\{ \max_{1 \leq j \leq \lfloor N^a \rfloor} \log U(t, x_j) \leq (\beta \log N)^{2/3} \right\} \\ &= P \left\{ \max_{1 \leq j \leq \lfloor N^a \rfloor} \log U(t, x_j) \leq (\beta \log N)^{2/3} \right\} - \prod_{j=1}^{\lfloor N^a \rfloor} P \left\{ \log U(t, x_j) \leq (\beta \log R)^{2/3} \right\} \\ &\quad + \left(1 - P \left\{ \log U(t, 0) > (\beta \log N)^{2/3} \right\} \right)^{\lfloor N^a \rfloor} \end{aligned}$$

Proof of lower bound

Let $\beta \in (0, \frac{1}{8} \sqrt{\frac{t}{2}})$. Choose and fix $a \in (0, \frac{1}{6})$ and $\epsilon \in (0, 1)$ such that $\beta < \frac{a}{\frac{4}{3} \sqrt{\frac{2}{t}} + \epsilon}$.

Let $x_j = jN/\lfloor N^a \rfloor$ for $j = 1, \dots, \lfloor N^a \rfloor$. Then

$$\begin{aligned} \mathbb{P} \left\{ \max_{0 \leq x \leq N} \log U(t, x) \leq (\beta \log N)^{2/3} \right\} &\leq \mathbb{P} \left\{ \max_{1 \leq j \leq \lfloor N^a \rfloor} \log U(t, x_j) \leq (\beta \log N)^{2/3} \right\} \\ &= \mathbb{P} \left\{ \max_{1 \leq j \leq \lfloor N^a \rfloor} \log U(t, x_j) \leq (\beta \log N)^{2/3} \right\} - \prod_{j=1}^{\lfloor N^a \rfloor} \mathbb{P} \left\{ \log U(t, x_j) \leq (\beta \log N)^{2/3} \right\} \\ &\quad + \left(1 - \mathbb{P} \left\{ \log U(t, 0) > (\beta \log N)^{2/3} \right\} \right)^{\lfloor N^a \rfloor} \\ &\lesssim \sum_{1 \leq j < k \leq \lfloor N^a \rfloor} [\text{Cov}(\log U(t, x_j), \log U(t, x_k))]^{1/3} + \left(1 - e^{-(\frac{4}{3} \sqrt{\frac{2}{t}} + \epsilon) \beta \log N} \right)^{\lfloor N^a \rfloor} \\ &\lesssim N^{2a - \frac{1}{3}} + e^{-\frac{1}{2} N^a - \frac{4}{3} (\sqrt{\frac{2}{t}} + \epsilon) \beta}. \end{aligned}$$

Proof of lower bound

Let $\beta \in (0, \frac{1}{8}\sqrt{\frac{t}{2}})$. Choose and fix $a \in (0, \frac{1}{6})$ and $\epsilon \in (0, 1)$ such that $\beta < \frac{a}{\frac{4}{3}\sqrt{\frac{2}{t}} + \epsilon}$.

Let $x_j = jN/\lfloor N^a \rfloor$ for $j = 1, \dots, \lfloor N^a \rfloor$. Then

$$\begin{aligned} \mathbb{P} \left\{ \max_{0 \leq x \leq N} \log U(t, x) \leq (\beta \log N)^{2/3} \right\} &\leq \mathbb{P} \left\{ \max_{1 \leq j \leq \lfloor N^a \rfloor} \log U(t, x_j) \leq (\beta \log N)^{2/3} \right\} \\ &= \mathbb{P} \left\{ \max_{1 \leq j \leq \lfloor N^a \rfloor} \log U(t, x_j) \leq (\beta \log N)^{2/3} \right\} - \prod_{j=1}^{\lfloor N^a \rfloor} \mathbb{P} \left\{ \log U(t, x_j) \leq (\beta \log N)^{2/3} \right\} \\ &\quad + \left(1 - \mathbb{P} \left\{ \log U(t, 0) > (\beta \log N)^{2/3} \right\} \right)^{\lfloor N^a \rfloor} \\ &\lesssim \sum_{1 \leq j < k \leq \lfloor N^a \rfloor} [\text{Cov}(\log U(t, x_j), \log U(t, x_k))]^{1/3} + \left(1 - e^{-(\frac{4}{3}\sqrt{\frac{2}{t}} + \epsilon)\beta \log N} \right)^{\lfloor N^a \rfloor} \\ &\lesssim N^{2a - \frac{1}{3}} + e^{-\frac{1}{2}N^a - \frac{4}{3}(\sqrt{\frac{2}{t}} + \epsilon)\beta}. \end{aligned}$$

Remark: by the ergodicity of $\{U(t, x) : x \in \mathbb{R}\}$ (Chen-Khoshnevisan-Nualart-P.-2022), there exist finite positive constants α_1, α_2 such that a.s.

$$\liminf_{N \rightarrow \infty} \frac{\max_{0 \leq x \leq N} \log U(t, x)}{(\log N)^{2/3}} = \alpha_1, \quad \limsup_{N \rightarrow \infty} \frac{\max_{0 \leq x \leq N} \log U(t, x)}{(\log N)^{2/3}} = \alpha_2.$$