

On the Limit Curlicue Process for Theta Sums

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Joint work with Jens Marklof, University of Bristol.

Theta Sums and Curves

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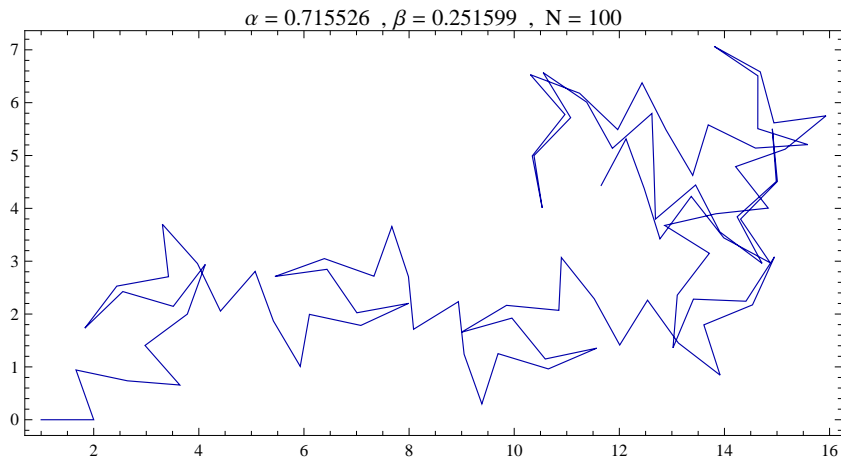
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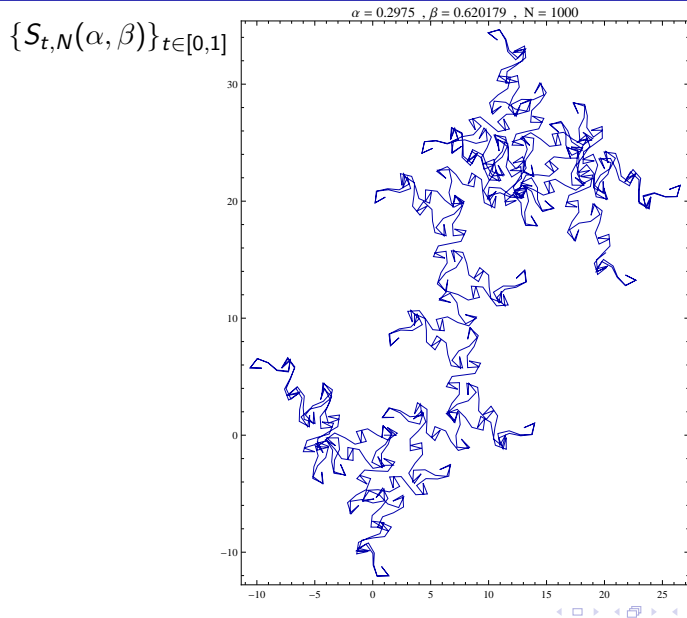
- ▶ We study the **curve** $[0, 1] \rightarrow \mathbb{C}$ obtained by linearly interpolating the N complex points $S_{\frac{1}{N},N}(\alpha, \beta), S_{\frac{2}{N},N}(\alpha, \beta), \dots, S_{\frac{N-1}{N},N}(\alpha, \beta), S_{1,N}(\alpha, \beta)$.

Example (1)

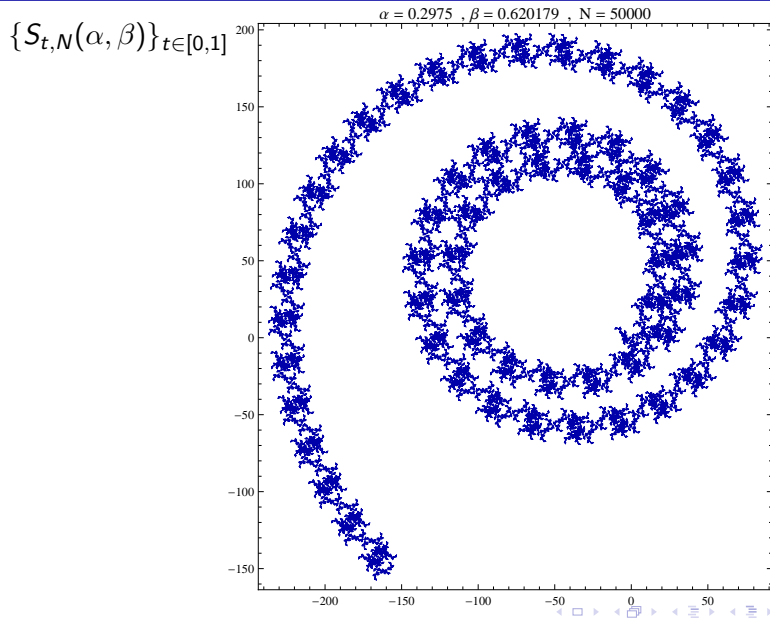
$$\{S_{t,N}(\alpha, \beta)\}_{t \in [0,1]}$$



Example (2)



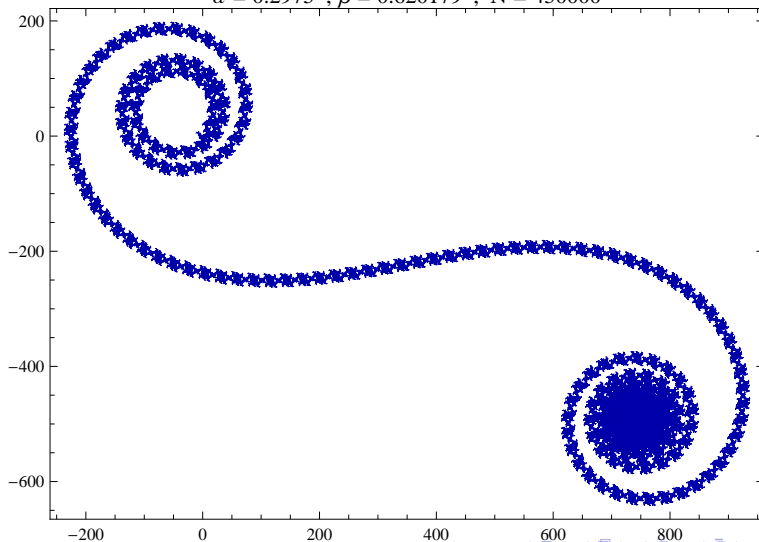
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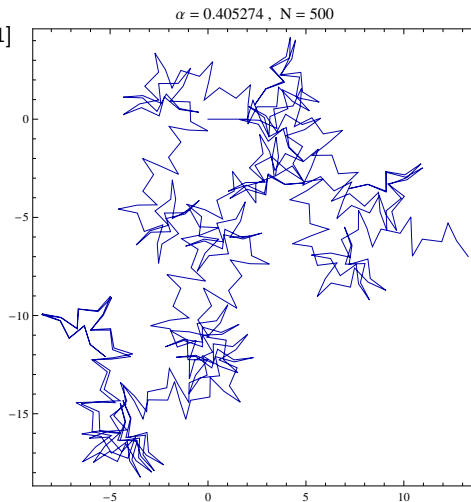
$$\{S_{t,N}(\alpha, \beta)\}_{t \in [0,1]}$$

$\alpha = 0.2975$, $\beta = 0.620179$, $N = 450000$



Example (3)

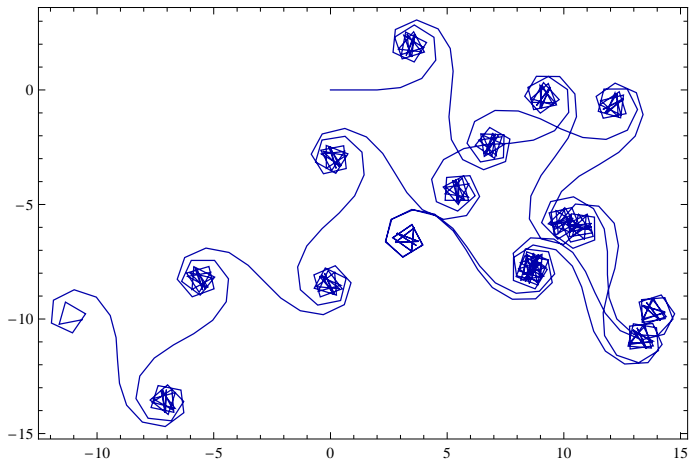
$$\{S_{t,N}(\alpha, 0)\}_{t \in [0,1]}$$



Example (4)

$$\{S_{t,N}(\alpha, 0)\}_{t \in [0,1]}$$

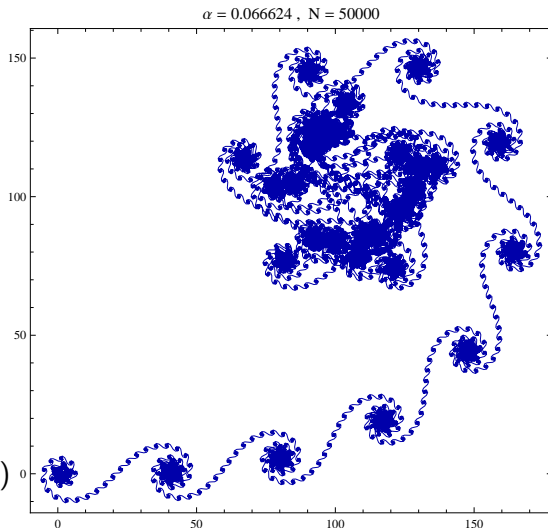
$\alpha = 0.0327471$, $N = 500$



Example (5)

$$\{S_{t,N}(\alpha, 0)\}_{t \in [0,1]}$$

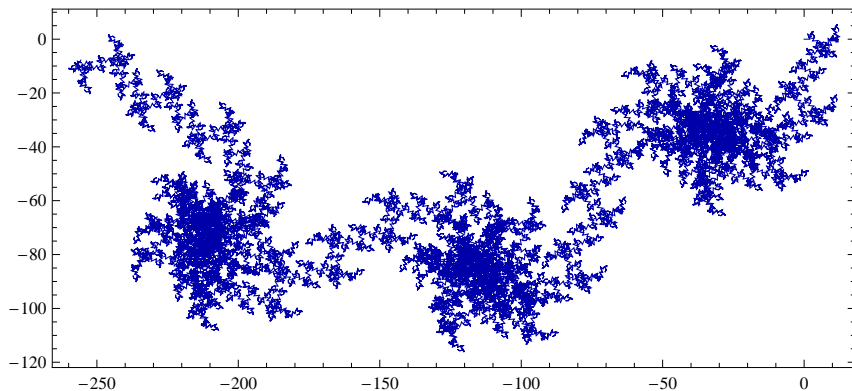
“curlicues”
(Berry-Goldberg 1988)



Example (6)

$$\{S_{t,N}(\alpha, 0)\}_{t \in [0,1]}$$

$$\alpha = 0.64825, \quad N = 29000$$



“Misshapen chaos of well-seeming forms”
(Romeo, *Romeo and Juliet*)

Growth of $|S_{1,N}(\alpha, \beta)|$

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- **Theorem A** (Hardy-Littlewood 1914). If α is bounded-type, then $|S_{1,N}(\alpha, 0)| \leq C\sqrt{N}$.

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- ▶ **Theorem B** (Friedler-Jurkat-Körner 1977; Flaminio-Forni 2006). For every increasing function $b : (1, \infty) \rightarrow (0, \infty)$ s.t. $\int_1^\infty u^{-1}b^{-4}(u)du < \infty$, there exists a full measure set \mathcal{G}_b such that for every $\alpha \in \mathcal{G}_b$ and every $\beta \in \mathbb{R}$ the following holds:

$$|S_{1,N}(\alpha, \beta)| \leq C\sqrt{N} b(N).$$

Limiting Distributions for $|S_{1,N}(\alpha, \beta)|$ and $S_{1,N}(\alpha, \beta)$

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- **Theorem C** (Jurkat-Van Horne 1981-3). There exists a function Ψ such that for all (except for countably many) $a, b \in \mathbb{R}_{\geq 0}$

$$\text{Leb} \left(\left\{ \alpha \in [0, 1] : a < N^{-\frac{1}{2}} |S_{1,N}(\alpha, 0)| < b \right\} \right) \xrightarrow{N \rightarrow \infty} \Psi_0(a, b).$$

The distribution function Ψ_0 is not Gaussian (it has only finitely many moments: $\Psi_0(R, \infty) \leq \frac{C_0}{R^4}$).

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- **Theorem D** (Marklof 1999). There exists a probability measure P_0 on \mathbb{C} such that for all *nice*, open $A \subset \mathbb{C}$

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- ▶ *Marklof's Thesis*: For $\beta \notin \mathbb{Q}$ the analogous statement for $S_{1,N}(\alpha, \beta)$ is true but $P_\beta = P \neq P_0$ does not depend on β and has slightly better decay: $\Psi_\beta(R, \infty) \leq \frac{C}{R^6}$ (and does not depend on β).

Limiting Distributions for $t \mapsto N^{-\frac{1}{2}} S_{t,N}(\alpha, 0)$

Consider the curve $\gamma_N^{\alpha,\beta} : [0, 1] \rightarrow \mathbb{C}$, $\gamma_N^{\alpha,\beta}(t) := N^{-\frac{1}{2}} S_{t,N}(\alpha, \beta)$.

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► **Theorem 1** (C. 2009, to appear in *Ann.Inst. Henri Poincaré*).

Fix $0 \leq t_1 < t_2 < \dots < t_k \leq 1$. There exists a probability measure $P_{0;t_1,\dots,t_k}^{(k)}$ (**curlicue measure**) on \mathbb{C}^k such that for all nice, open $A \subset \mathbb{C}^k$

$$\lambda \left(\left\{ \alpha \in [0, 1] : (\gamma_N^{\alpha,0}(t_j))_{j=1}^k \in A \right\} \right) \xrightarrow{N \rightarrow \infty} P_{0;t_1,\dots,t_k}^{(k)}(A),$$

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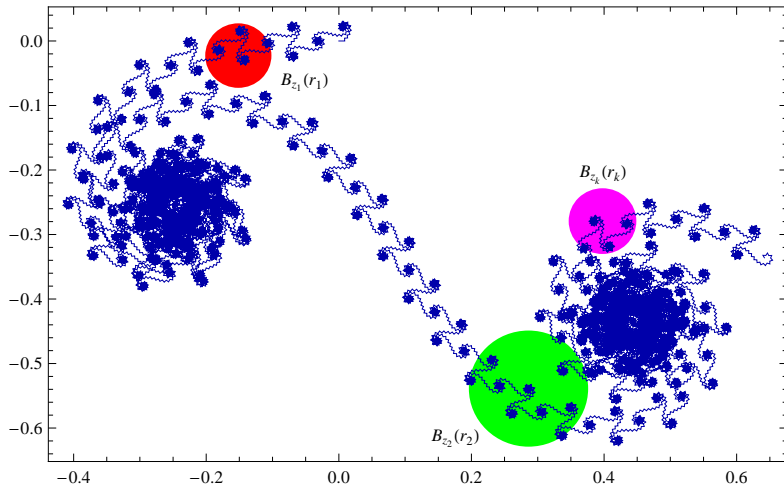
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- **Thm D** corresponds to $\lambda = \text{Leb}$, $k = 1$ and $t_1 = 1$. In particular $P_0 = P_{0;1}^{(1)}$.

$\alpha = 0.50512$, $N = 51000$



In other words:

- For every N , given the map $([0, 1], \mathcal{B}, \lambda) \ni \alpha \mapsto \gamma_{\alpha, N} \in (\mathcal{C}([0, 1], \mathbb{C}), \mathcal{B}_{\mathcal{C}})$, construct the induced measure

$$P_{\beta}^{(N)}(\mathcal{A}) := \lambda \left(\left\{ \alpha \in [0, 1] : \gamma_N^{\alpha, \beta} \in \mathcal{A} \right\} \right), \quad \mathcal{A} \in \mathcal{B}_{\mathcal{C}}.$$

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- We have the canonical projection $\pi_{t_1, \dots, t_k} : \mathcal{C}([0, 1], \mathbb{C}) \rightarrow \mathbb{C}^k$,

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- ▶ **Theorem 1** can be rephrased as:

$$P_0^{(N)} \pi_{t_1, \dots, t_k}^{-1} \implies P_{0; t_1, \dots, t_k}^{(k)} \quad \text{as } N \rightarrow \infty.$$

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- ▶ (Joint work with Jens Marklof). For $\beta \notin \mathbb{Q}$, we prove the analogous limit theorem for $t \mapsto N^{-\frac{1}{2}} S_{t, N}(\alpha, \beta)$. The measure $\tilde{P}_\beta = \tilde{P} \neq \tilde{P}_0$ does not depend on β . We want to understand how a \tilde{P}_0 (or \tilde{P})-typical curve “looks like”.

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- ▶ An alternative approach uses equidistribution of long, closed horocycles in the unit tangent bundle of a suitably defined non-compact hyperbolic manifold. In particular, it provides an alternative probability space $(\mathcal{M}, \bar{\mu})$ (a homogeneous space) and realizes the curlicue process as an “explicit” function of the geodesic flow on \mathcal{M} .
This approach extends from $\beta = 0$ to general β .

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$$G = \widetilde{\mathrm{SL}}(2, \mathbb{R}) \ltimes \mathrm{H}(\mathbb{R})$$

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- ▶ $\mathcal{M} = \Gamma \backslash G$ is a bundle over $\mathcal{M}_0 = \tilde{\Gamma}_0 \backslash \widetilde{\mathrm{SL}}(2, \mathbb{R})$.
 \mathcal{M}_0 is a 4-fold cover of $T^1 M$, where $M = \Gamma_0 \backslash \mathfrak{H}$ is a non-compact hyperbolic surface with three cusps $(0, \pm \frac{1}{2}, \infty)$ and finite area. Let us normalize $\bar{\mu} = \mu / \mu(\mathcal{M})$.

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- ▶ The geodesic flow Φ^s on \mathcal{M} is given by right multiplication by $\left(\begin{pmatrix} e^{-s/2} & 0 \\ 0 & e^{s/2} \end{pmatrix}, 0; \mathbf{0}, 0 \right) \in G$.

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$$\Theta_f(z, \phi; \xi, \zeta) := y^{\frac{1}{4}} e(\zeta - \frac{1}{2}\xi_1\xi_2) \sum_{n \in \mathbb{Z}} f_{\phi}\left((n - \xi_2)y^{\frac{1}{2}}\right) e\left(\frac{1}{2}(n - \xi_2)^2x + n\xi_1\right),$$

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- ▶ where

$$f_\phi(w) := \begin{cases} e(-\sigma_\phi/8) f(w) & (\phi \equiv 0 \pmod{2\pi}) \\ e(-\sigma_\phi/8) f(-w) & (\phi \equiv \pi \pmod{2\pi}) \\ \frac{e(-\sigma_\phi/8)}{|\sin \phi|^{1/2}} \int_{\mathbb{R}} e\left(\frac{\frac{1}{2}(w^2 + w'^2) \cos \phi - ww'}{\sin \phi}\right) f(w') dw' & (\phi \not\equiv 0 \pmod{\pi}) \end{cases}$$

$$\sigma_\phi := \begin{cases} 2\nu, & \text{if } \phi = \nu\pi, \nu \in \mathbb{Z}; \\ 2\nu + 1, & \text{if } \nu\pi < \phi < (\nu + 1)\pi, \nu \in \mathbb{Z}. \end{cases}$$

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- ▶ We are interested in Θ_χ , where $\chi = 1_{(0,1)}$.
- ▶ The limit curlicue process for $\beta \notin \mathbb{Q}$ can be described as

$$[0, 1] \ni t \mapsto \gamma_g(t) = \frac{\sqrt{t}}{\mu(\mathcal{M})} \Theta_\chi(g\Phi^s), \quad s = 2\log t, \quad g \in \mathcal{M},$$

where $g \in \mathcal{M}$ is $\bar{\mu}$ -random.

For $\beta = 0$ we have

$$[0, 1] \ni t \mapsto \gamma_g(t) = \frac{\sqrt{t}}{\mu(\mathcal{M}_0)} \Theta_\chi(g\Phi^s), \quad s = 2\log t, \quad g \in \mathcal{M}_0.$$

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- ▶ (time inversion) $t\gamma_g(1/t)$ and $\gamma_g(t)$ have the same distribution.
- ▶ In other words: the curlicue process shares many properties with the Brownian motion, even if it has dependent increments and its finite dimensional distribution are not Gaussian.

Thank You!

