

# Continued fractions, the Chen-Stein method and extreme value theory

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# Regular Continued Fractions: A Crash Course

# Regular continued fractions

$$\frac{7}{24}$$

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Note that  $\textcolor{red}{7} > \textcolor{red}{3} > \textcolor{red}{1}$ .

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Note that  $7 > 3 > 1$ .

Therefore by Euclidean Algorithm (of computing gcd), any rational number

$$\omega = p/q \in (0, 1)$$

(with  $\gcd(p, q) = 1$ ) will have a *terminating (regular) continued fraction expansion*.

# Conversely . . .

Whenever  $A_1, A_2, A_3, A_4 \in \mathbb{N}$ ,

$$[A_1, A_2, A_3, A_4] := \frac{1}{A_1 + \frac{1}{A_2 + \frac{1}{A_3 + \frac{1}{A_4}}}} \in (0, 1)$$

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More generally, by induction on  $n$ ,

$$\omega = [A_1, A_2, \dots, A_n]$$

(with  $A_1, A_2, \dots, A_n \in \mathbb{N}$ ) is a rational number in  $(0, 1)$ .

# Non-terminating continued fraction expansion

## Theorem

*A number  $\omega \in (0, 1)$  has a unique non-terminating continued fraction expansion*

$$\omega = \frac{1}{A_1 + \frac{1}{A_2 + \frac{1}{A_3 + \dots}}} =: [A_1, A_2, A_3, \dots]$$

*(with each  $A_i \in \mathbb{N}$ ) if and only if  $\omega \notin \mathbb{Q}$ .*

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**Examples:**  $\pi \approx \frac{22}{7}$  and  $\pi \approx \frac{355}{113}$ .

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See, for example, [Khintchine \(1964\)](#).

# Gauss Dynamical System

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**Quick Observation:**  $T, A_1$  measurable  $\Rightarrow$  each  $A_n$  measurable.

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# Extremes of Continued Fractions and the Melancholic Life of Wolfgang Doeblin

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- Since then: [Philipp \(1976\)](#), [Samur \(1989\)](#), [Nakada and Natsui \(2003\)](#), [Pollicott \(2009\)](#), [Tyran-Kamińska \(2010\)](#), [Bazarova, Berkes and Horváth \(2016\)](#), [Chang and Ma \(2017\)](#).



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- See also [Wolfgang Doeblin: A mathematician rediscovered](#).



# Doeblin-Iosifescu Asymptotics

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# A reformulation of Gauss's theorem

**Exercise (in *Probability Theory II*):** Suppose  $X$  is a random variable having probability density function

$$f_X(x) = \frac{1}{(1+x) \log 2}, \quad x \in (0, 1).$$

Then show that  $\{1/X\} \stackrel{\mathcal{L}}{=} X$ .

# A stationary process

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# Two easy observations

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Theorem (Doeblin (1940), Iosifescu (1977))

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Let  $M_n^{(1)} := \max\{A_i \log 2 : 1 \leq i \leq n\}$ ,  $n \in \mathbb{N}$ . Then for all  $u > 0$ ,

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# Our Contribution



# The main question

Theorem (Doeblin (1940), Iosifescu (1977))

For all  $u > 0$ ,

$$(DI) \quad \mathcal{E}_n^u := \#\{1 \leq j \leq n : A_j \log 2 > un\} \xrightarrow{\mathcal{L}} \mathcal{E}_\infty^u \sim \text{Poi}(u^{-1})$$

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Question

*What is the rate of convergence in (DI)?*

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- Rate of convergence of the scaled maxima for the geodesic flow on the modular surface.

# The main result

Theorem ([Ghosh, Kirsebom, R. \(2019\)](#))

*There exists  $\kappa > 0$  and a sequence  $1 \ll \ell_n \ll \log n$  such that for all  $u > 0$  and for all  $n \in \mathbb{N}$ ,*

$$d_{TV}(\mathcal{E}_n^u, \mathcal{E}_\infty^u) := \sup_{A \subseteq \mathbb{N} \cup \{0\}} |P(\mathcal{E}_n^u \in A) - P(\mathcal{E}_\infty^u \in A)| \leq \frac{\kappa}{\min\{u, u^2\}} \frac{\ell_n}{n}.$$

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Corollary

*Suppose  $M_n^{(k)} := k^{\text{th}}$  maximum of  $\{A_i \log 2 : 1 \leq i \leq n\}$ . For all  $u > 0$  and for all  $n \in \mathbb{N}$ ,*

$$\sup_{k \in \mathbb{N}} \left| P\left(\frac{M_n^{(k)}}{n} \leq u\right) - e^{-u^{-1}} \sum_{i=0}^{k-1} \frac{u^{-i}}{i!} \right| \leq \frac{\kappa}{\min\{u, u^2\}} \frac{\ell_n}{n}.$$

# Comparison with existing results

- **Resnick and de Haan (1989)**: If  $A_1, A_2, \dots$  were independent, then

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# Sketch of Proof of the Main Result:

Ergodic Theory + Hard Analysis + Applied Probability

# Sketch of proof

Recall  $\mathcal{E}_n^u = \sum_{j=1}^n \mathbb{1}_{(A_j \log 2 > un)} \stackrel{\text{approx}}{\sim} \text{Bin}(n, p_n = P(A_1 \log 2 > un))$ .

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- Estimate  $d_{TV}(\tilde{\mathcal{E}}_n^u, \mathcal{E}_\infty^u)$  using *second order regular variation*.

# How to estimate $d_{TV}(\tilde{\mathcal{E}}_n^u, \mathcal{E}_\infty^u)$ ?

Recall  $\tilde{\mathcal{E}}_n^u \sim \text{Poi}(np_n)$  and  $\mathcal{E}_\infty^u \sim \text{Poi}(u^{-1})$ .

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Theorem ([Philipp \(1970\)](#); see also [Galambos \(1972\)](#))

*There exists  $C > 0$  and  $\theta > 1$  such that for all  $m, n \in \mathbb{N}$ , for all  $F \in \sigma(A_1, A_2, \dots, A_m)$ , and for all  $H \in \sigma(A_{m+n}, A_{m+n+1}, \dots)$ ,*

$$|P(F \cap H) - P(F)P(H)| \leq C\theta^{-n} P(F)P(H).$$

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- Overall, application of Theorem 1 of [Smith \(1988\)](#), instead of Theorem 2 of [Arratia, Goldstein and Gordon \(1989\)](#), will result in an argument of similar length.
- However, we have not compared the rates obtained by these two results in our setup.

# Consequences and Future Work

# The central limit theorem

Theorem 3.1 of [Davis and Hsing \(1995\)](#) + extremal point process convergence yields

$$\frac{A_1 + A_2 + \cdots + A_n - E(A_1 \mathbb{1}_{(A_1 \leq n)})}{n} \xrightarrow{\mathcal{L}} S,$$

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This recovers a result of [Samur \(1989\)](#), who proved this using “direct frontal attack” with the help of exponential mixing.

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Regular continued fraction is egalitarian towards all (residue) classes.

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Our work yields the rate of convergence in Pollicott's result.

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- Extremes in the context of number theory and geometry.

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- Initiated during a visit by Maxim Sølund Kirsebom and P.R. at *Tata Institute of Fundamental Research, Mumbai*.
- Significant portion at *International Centre for Theoretical Sciences, Bangalore* during the program *Probabilistic Methods in Negative Curvature*.

Thank You Very Much