# Central limit theorems in number theory and graph theory

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- A "vertical" variant of the Sato-Tate distribution.
- General principles of asymptotic distribution of families in compact intervals of  $\mathbb{R}$ .
- Thematic similarities between the eigenvalues of Hecke operators and eigenvalues of families of regular graphs.

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• Let  $\{P_n(t)\}$ ,  $P_n(t)$  is a degree n polynomial. For all  $n \ge 0$ ,

$$\lim_{V\to\infty}\frac{1}{|A_V|}\sum_{\lambda\in A_V}P_n(\lambda)=\int_{\Omega}P_n(t)\mu(t)dt.$$



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• We denote  $Z_n(x) = X_n(x) - X_{n-2}(x), n \ge 2.$ 



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• For  $n \geq 1$ , let  $T_n: S_k(N) \to S_k(N)$  denote the n-th Hecke operator.



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- Any Hecke newform  $f(z) \in \mathcal{F}_k(N)$  has a Fourier expansion

$$f(z) = \sum_{n=1}^{\infty} n^{\frac{k-1}{2}} a_f(n) q^n,$$

where  $a_f(1) = 1$  and

$$\frac{T_n(f(z))}{n^{\frac{k-1}{2}}} = a_f(n)f(z), \ n \ge 1.$$

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$$\lim_{\substack{k+N\to\infty\\p\nmid N}}\frac{N_I(p,N,k)}{s_k(N)}=\int_I\nu_p(t)dt,$$

where

$$u_p(t) = rac{p+1}{\pi} rac{\sqrt{1-rac{t^2}{4}}}{(p^{1/2}+p^{-1/2})^2-t^2}.$$



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- Is there a central limit-type theorem that predicts the distribution of

$$\frac{\sum_{p \le x} (Y_p - \mu_p)}{\sqrt{\sum_{p \le x} \sigma_p^2}}$$

# Theorem (Conrey, Duke Farmer, 1997, Nagoshi, 2006)

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For any integer  $r \geq 0$ ,

$$\lim_{x \to \infty} \frac{1}{s_k(1)} \sum_{f \in \mathcal{F}_k(1)} \left( \frac{\sum_{p \le x} \chi_I(a_f(p)) - \pi(x)\mu(I)}{\sqrt{\pi(x)[\mu(I) - \mu(I)^2]}} \right)^r$$
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That is,  $\frac{\sum_{p\leq x}(Y_p-\mu_p)}{\sqrt{\sum_{p\leq x}\sigma_p^2}}$  converges to the normal distribution as  $x\to\infty$ . Our motivation was a result of Nagoshi, which showed that

$$\frac{\sum_{p\leq x}a_f(p)}{\sqrt{\pi(x)}}$$

converges to the normal distribution as  $x \to \infty$  (as  $\log k / \log x \to \infty$ ).



Let 
$$G(t)=\sum_{m\geq 0}U(m)X_{2m}(t),\ |U(m)|\ll e^{-\alpha m^\omega}.$$
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- $\sum_{p\leq x} \mathbb{E}[Y_p] \sim \pi(x) \int_0^1 G(t)\mu(t)dt$ ,
- $\sum_{p \le x} \sigma_p^2 \sim \pi(x) V_G$ , where

$$V_G=\int_0^1 G(t)^2 \mu(t)dt-\left(\int_0^1 G(t)\mu(t)dt
ight)^2.$$



• For any integer  $r \geq 0$ ,

$$\frac{1}{s_k(1)} \sum_{f \in \mathcal{F}_k(1)} \left( \frac{\sum_{p \le x} Y_p(f) - \pi(x) \int_0^1 G(t) \mu(t) dt}{\sqrt{\pi(x) V_G}} \right)^r$$
$$\sim \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} t^r e^{-t^2/2} dt.$$

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- Suppose we label the vertices of  $X \in \mathcal{G}(m,d)$  as  $v_1, v_2, \ldots, v_m$ . Its adjacency matrix  $\mathcal{A} = \mathcal{A}(X) \in \{0,1\}^{m \times m}$  is defined by setting  $\mathcal{A}_{ij} = 1$  if there is an edge  $(v_i, v_j) \in \mathcal{E}$ , and  $\mathcal{A}_{ij} = 0$  otherwise.

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- Let  $d \ge 3$  be fixed. By a 2008 result of Friedman, for X chosen uniformly at random from  $\mathcal{G}(m,d)$  and for any  $\epsilon > 0$ ,

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- Let  $\Lambda(x) := \frac{\lambda(X)}{\sqrt{d-1}}$ .
- Let  $d \ge 3$  be fixed. By a 2008 result of Friedman, for X chosen uniformly at random from  $\mathcal{G}(m,d)$  and for any  $\epsilon > 0$ ,

$$P[\Lambda(X) > 2 + \epsilon] \to 0 \text{ as } m \to \infty.$$

- Let us choose a sequence  $\{X_m\}$  of graphs such that  $X_m \in \mathcal{G}(m,d)$ . How are the families  $A(X_m)$  distributed as  $m \to \infty$ ?
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Note: if d = p + 1, we recover  $\nu_p(t)$ .



• For some function  $f: \mathbb{R} \to \mathbb{R}$ , we may define a random variable  $B_f: \mathcal{G}(m,d) \to \mathbb{R}$  by

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$$d(m) o \infty, \; d_m - 1 = m^{\epsilon_m} \; ext{for some} \; \epsilon_m = ext{o}(1),$$

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• Moreover, for an interval  $I \subset [-2, 2]$ ,

$$\mathbb{E}\left[B_{\chi_I}\right] \sim m \int_I \mu(t).$$



#### Theorem (Johnson, 2015)

Suppose f is an entire function of finite order. In this case, f can be expressed as

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## Theorem (Ctd.)

For every  $r \geq 0$ ,

$$\lim_{m\to\infty} \frac{1}{|\mathcal{G}(m,d(m))|} \sum_{Y\in\mathcal{G}(m,d(m))} \left( \frac{\left(B_f(Y) - m\int_{-2}^2 f(t)\mu(t)dt\right)}{\sqrt{V[B_f]}} \right)^r$$

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$$= \begin{cases} 0 & \text{if } r \text{ is odd} \\ \frac{r!}{(r/2)!2^{r/2}} & \text{if } r \text{ if even.} \end{cases}$$

Can we obtain asymptotics  $\mathbb{E}[B_f]$ ,  $V[B_f]$  and a central limit theorem under "looser" growth conditions for d in terms of m(possibly for a restricted class of test functions)?



• One takes a real-valued, even test function  $g \in C^{\infty}(\mathbb{R})$ .

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- Suppose the Fourier transform  $\hat{g}$  has compact support, without loss of generality, in [-1,1].
- $G_L(\theta)$  is a periodic function with Fourier expansion

$$G_L(\theta) = \frac{1}{L} \sum_{|m| \leq L} \widehat{g}\left(\frac{m}{L}\right) e(m\theta), \ e(x) = e^{2\pi i x}.$$

In fact,

$$H_L(t) := G_L(\theta) = \sum_{0 \le m \le L} U(m) X_{2m}(t), t = 2 \cos \pi \theta$$
, where

$$U(m) = \frac{1}{L} \left( \widehat{g} \left( \frac{m}{L} \right) - \widehat{g} \left( \frac{m+1}{L} \right) \right).$$



Let  $L \geq 1$ . Define  $B_{H_L}: \mathcal{G}(m,d) \to \mathbb{R}$  by

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## What next?

We should look at the asymptotics of the average

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Can we find suitable growth conditions for d under which the above  $\sim V_{H_I}$  and

$$\mathbb{E}\left[\left(\frac{\left(B_{H_L}(Y)-m\int_{-2}^2H_L(t)\mu(t)dt\right)}{\sqrt{V_{H_L}}}\right)^r\right]$$

converges to Gaussian moments as  $m \to \infty$ ?



### Proposition

Let  $Y \in \mathcal{G}(m,d)$ ,  $d \geq 3$ . Let A(Y) denote the family of eigenvalues of  $\mathcal{T}'$  (with respect to Y). For a positive integer n, let  $f_n(Y)$  denote the number of closed non-backtracking walks (CNBW)'s of length n in Y and let  $C_n(Y)$  denote the number of circuits of length n in Y.

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