

How Explicit is the Explicit Formula?

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Madison

- William Stein. Joint work with Barry Mazur.
(and some with Hao -)

① Birch & Swinnerton-Dyer's Hunch

E/\mathbb{Q} elliptic curve

$E(\mathbb{Q}) \rightsquigarrow E(\mathbb{F}_2), E(\mathbb{F}_3), \dots, E(\mathbb{F}_p)$

$$\text{sign}(a_p) = \begin{cases} 1 & \text{if } E(\mathbb{F}_p) \text{ small} \\ +1 & \text{if } E(\mathbb{F}_p) \text{ big} \end{cases}$$

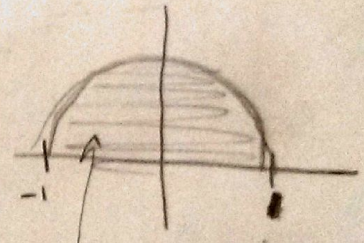
STATISTICAL
big \longleftrightarrow tend to be big

BIAS $a_p = p^{11} - \#E(\mathbb{F}_p)$ "big" means $a_p < 0$.

$$\text{rank}(E(\mathbb{Q})) \stackrel{\text{conj}}{=} \text{ord } L(E, s) \Big|_{s=1}, \quad L^*(E, s) = \prod \frac{p}{1 - a_p p^{-s} + p^{1-2s}}, \quad L^*(E, 1) = \prod \frac{1}{\#E(\mathbb{F}_p)}$$

Theorem ("Sato-Tate", Taylor et al.) (E not CM)

$\left\{ \frac{a_p}{2\sqrt{p}} : p \text{ prime} \right\}$ is distributed like this



perfect semicircle!

(show picture for $p < 10^9$).

Animation!
for 5071a.

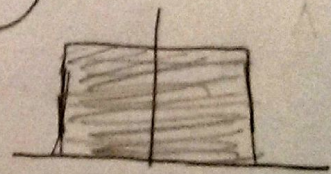
So: in the limit $\text{sign}(a_p) = \pm 1$ with equal prob!

② Chebyshev Bias

This talk: Explain what is really going on here.

$2 \neq p$ prime $p \equiv \begin{cases} 1 \\ 3 \end{cases} \pmod{4}$

Theorem (Dirichlet): Distribution of $p \pmod{4}$ is this



(show computer pic for $p < 10^9$)

Animation.

Chebyshev more refined question: Dirichlet (coef. of $L(\chi, s)$)

What about $\delta(x) = \sum_{2 < p \leq x} \left(\frac{-1}{p}\right)$ where $b_p \equiv \left(\frac{-1}{p}\right) = \begin{cases} -1 & \text{if } p \equiv 3 \\ 1 & \text{if } p \equiv 1 \end{cases} ?$

Prime race: show an animation.

Theorem (Rubinstein - Sarnak):

③ The Explicit Formula (Riemann)

$$(\text{Sum of local data}) = (\text{Global data}) + (\text{Oscillatory term}) + (\text{convergent term})$$

$$S(X) = F(X) \sum_{p \leq X} G(p) + \sum_{p \leq X} \left(\begin{array}{l} \text{real zeros} \\ \text{of } L\text{-functions} \\ \text{(so BSD, etc!)} \end{array} \right) + \left(\begin{array}{l} \text{infinite sum} \\ \text{over imaginary} \\ \text{zeros} \\ \text{(so GRH)} \end{array} \right) + \left(\begin{array}{l} ? \\ \end{array} \right)$$

Theorem (Riemann ζ -1)

$$\Delta(X) = \text{Von Mangoldt} = \begin{cases} 0 & \text{otherwise,} \\ \log(p) & \text{if } X=p^n. \end{cases}$$

$$\Psi_0(X) = \frac{1}{2} \Delta(X) + \sum_{n < X} \Delta(n)$$

Then

$$\frac{1}{X} \Psi_0(X) = 1 - \sum_{|y| \leq T} \frac{X^{-1/2 + iy}}{(\frac{1}{2} + iy)\sqrt{X}} + C(X, T)$$

(a statement for each T)
 $\lim_{X \rightarrow \infty} \lim_{T \rightarrow \infty} C(X, T) = 0$

where: $C(X, T) = \frac{-\log(2\pi) - \log(1 - 1/X^2)/2}{X} + \varepsilon(X, T)$ (Plot this all!)

$$|\varepsilon(X, T)| \ll \frac{\log(X)}{X} \cdot \min(1, \frac{X}{T \log(X)}) + \frac{\log^2(XT)}{T}$$

④ Raw, medium, and well-done sums - elliptic curves E/\mathbb{Q}

(on # field) or newform

Raw data:

$$\Delta_E(X) = \frac{\log X}{\sqrt{X}} \sum_{p \leq X} \text{sign}(a_E(p))$$

Plot each for $p \leq 10^9$ on a log scale for several E .

Medium-rare data:

$$D_E(X) = \frac{\log(X)}{\sqrt{X}} \sum_{p \leq X} \frac{a_E(p)}{\sqrt{p}}$$

Random walks

Histogram of "returns" is Sato-Tate

Plot frequency histogram: of values of each.

Well-done data:

$$D_E(X) = \frac{1}{\log(X)} \sum_{p \leq X} \frac{a_E(p) \cdot \log(p)}{p}$$

- raw: infinite variance
- medium: finite variance
- well-done: nice.

⑤ Sarnak's Letter - Conjectures

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Madison

Make histogram idea precise:

If $X \mapsto \delta(X)$ continuous function on \mathbb{R} , say $\delta(x)$ possesses limiting distribution μ_δ w.r.t multiplicative measure $\frac{dx}{x}$ for bounded cont f :

$$\lim_{X \rightarrow \infty} \frac{1}{\log X} \int_0^X f(\delta(x)) \frac{dx}{x} = \int_{\mathbb{R}} f(x) d\mu_\delta(x).$$

Mean: $\int_{\mathbb{R}} \delta(x) d\mu_\delta(x) = \lim_{X \rightarrow \infty} \frac{1}{\log X} \int_0^X \delta(x) \frac{dx}{x}$

Conj (Sarnak):

- $\mathcal{E}(D_E) = -r = -(\text{analytic rank of } E)$.

Show plot of
for each
on log scale.

- $\mathcal{E}(D_E) = 1 - 2r$

- $\mathcal{E}(\Delta_E) = \frac{2}{\pi} - \frac{16}{3\pi} r + \frac{4}{\pi} \sum_{k=1}^{\infty} (-1)^{k+1} \left[\frac{1}{2k+1} + \frac{1}{2k+3} \right] r_{(2k+1)}$

He deduces these using GRH + more + explicit formula.

$r(n) = \text{ord } L(Sym^n E)$
s-center.

⑥ Data, A Minimalist Conjecture.

Conj: 50% of elliptic curves have rank 0
50% _____ 1.

Data (which is against this)
RM heuristics - for this

Open Question: What about $r_E(n)$? Minimal 100% of time subject to f.i.e.?
Very little data - Watkins - _____ curves.

Idea: Assume minimalist and "compute" $\mathcal{E}(\Delta_E)$...

Examples: (various ranks)

⑦ Oscillatory Term, The Bayesian

Explicit formula for $D_E(X)$ (well-done):

$$D_E(X) = -r + \frac{1}{\log X} \cdot \underbrace{\sum_{0 < |\gamma| \leq T} \frac{X^{i\gamma}}{i\gamma}}_{S(X,T)} + \underbrace{C_E(X,T)}_{\text{goes to 0}}$$

What is $S(X,T)$? Crazy.

// calculation

$$\frac{2}{\log(X)} \sum_{0 < \gamma \leq T} \frac{1}{\gamma} \sin(\gamma \log X)$$

• sort of alternating harmonic series.
• locally crazy as $T \rightarrow \infty$, for X in interval
• plot it.

The Bayesian: Frequency histogram of $S(X,T)$ (fixed T , $X \rightarrow \infty$)

my ideal

Conj: Values are normally distributed: show examples

So we compute the dist, hence understand that term.

⑧ The convergent Term

Show some plots.

I don't understand $C_E(X,T)$ yet. but it might be the easiest (or hardest) to