

## **Theoretical Physics Seminar**

**From the Manhattan Project to Number Theory:  
How Nuclear Physics helps us understand primes**

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## Fundamental Problem: Spacing Between Events

General Formulation: Studying system, observe values at  $t_1, t_2, t_3, \dots$ .

Question: what rules govern the spacings between the  $t_i$ ?

Examples:

- Spacings b/w Energy Levels of Nuclei.
- Spacings b/w Eigenvalues of Matrices.
- Spacings b/w Primes.
- Spacings b/w Zeros of Functions.

## Goals of the Talk

- Determine correct scale to study spacings.
- See similar behavior in different systems.
- Discuss tools / techniques needed to prove the results.

# **PART I**

## **NORMALIZED SPACINGS**

## Normalized Spacing

Example: Fractional Parts

For  $\alpha \notin \mathbb{Q}$ , set  $x_n = n\alpha \bmod 1$ .

Order  $x_1, \dots, x_N$ :  $0 \leq y_1 \leq \dots \leq y_N \leq 1$ .

Expect spacings between adjacent  $y$ 's of size  $\frac{1}{N}$ .

Should study  $\frac{y_{n+1} - y_n}{1/N}$ .

# **PART II**

# **RANDOM MATRIX THEORY**

# Origins of Random Matrix Theory

Classical Mechanics: 3 Body Problem Intractable.

Heavy nuclei like Uranium (200+ protons / neutrons) even worse!

Get some info by shooting high-energy neutrons into nucleus, see what comes out.

Fundamental Equation:

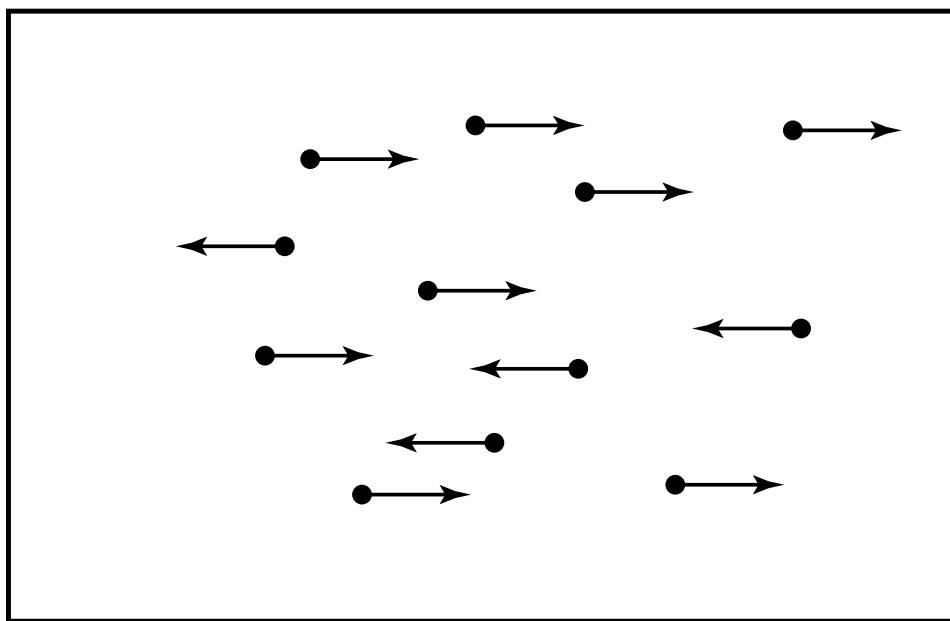
$$H\psi_n = E_n\psi_n$$

$H$  : matrix, entries depend on system

$E_n$  : energy levels

$\psi_n$  : energy eigenfunctions

## Origins (continued)



Statistical Mechanics: for each configuration, calculate quantity (say pressure).

Average over all configurations – most configurations close to system average.

Nuclear physics: choose matrix at random, calculate eigenvalues, average over matrices.

Look at: Real Symmetric ( $A = A^T$ ), Complex Hermitian ( $\overline{A}^T = A$ ).

# Random Matrix Ensembles

Real Symmetric Matrices:

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1N} \\ a_{12} & a_{22} & a_{23} & \cdots & a_{2N} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{1N} & a_{2N} & a_{3N} & \cdots & a_{NN} \end{pmatrix} = A^T, \quad a_{ij} = a_{ji}$$

Fix  $p$ , define

$$\text{Prob}(A) = \prod_{1 \leq i \leq j \leq N} p(a_{ij}).$$

This means

$$\text{Prob} (A : a_{ij} \in [\alpha_{ij}, \beta_{ij}]) = \prod_{1 \leq i \leq j \leq N} \int_{x_{ij}=\alpha_{ij}}^{\beta_{ij}} p(x_{ij}) dx_{ij}.$$

Want to understand eigenvalues of  $A$ .

## Eigenvalue Distribution

$\delta(x - x_0)$  is a unit point mass at  $x_0$ .

To each  $A$ , attach a probability measure:

$$\begin{aligned}\mu_{A,N}(x) &= \frac{1}{N} \sum_{i=1}^N \delta\left(x - \frac{\lambda_i(A)}{2\sqrt{N}}\right) \\ \int_a^b \mu_{A,N}(x) dx &= \frac{\#\left\{\lambda_i : \frac{\lambda_i(A)}{2\sqrt{N}} \in [a, b]\right\}}{N} \\ k^{\text{th}} \text{ moment} &= \frac{\sum_{i=1}^N \lambda_i(A)^k}{2^k N^{\frac{k}{2}+1}} = \frac{\text{Trace}(A^k)}{2^k N^{\frac{k}{2}+1}}.\end{aligned}$$

## Wigner's Semi-Circle Law

$N \times N$  real symmetric matrices, entries i.i.d.r.v. from a fixed  $p(x)$ .

**Semi-Circle Law:** Assume  $p$  has mean 0, variance 1, other moments finite.  
Then for almost all  $A$ , as  $N \rightarrow \infty$

$$\mu_{A,N}(x) \longrightarrow \begin{cases} \frac{2}{\pi} \sqrt{1 - x^2} & \text{if } |x| \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$

## SKETCH OF PROOF: Correct Scale

$$\text{Trace}(A^2) = \sum_{i=1}^N \lambda_i(A)^2.$$

By the Central Limit Theorem:

$$\text{Trace}(A^2) = \sum_{i=1}^N \sum_{j=1}^N a_{ij}a_{ji} = \sum_{i=1}^N \sum_{j=1}^N a_{ij}^2 \sim N^2$$

$$\sum_{i=1}^N \lambda_i(A)^2 \sim N^2$$

Gives  $N\text{Ave}(\lambda_i(A)^2) \sim N^2$  or  $\text{Ave}(\lambda_i(A)) \sim \sqrt{N}$ .

## SKETCH OF PROOF: Method of Moments

$$\text{Trace}(A^2) = \sum_{i=1}^N \sum_{j=1}^N a_{ij} a_{ji} = \sum_{i=1}^N \sum_{j=1}^N a_{ij}^2.$$

Substituting into expansion gives

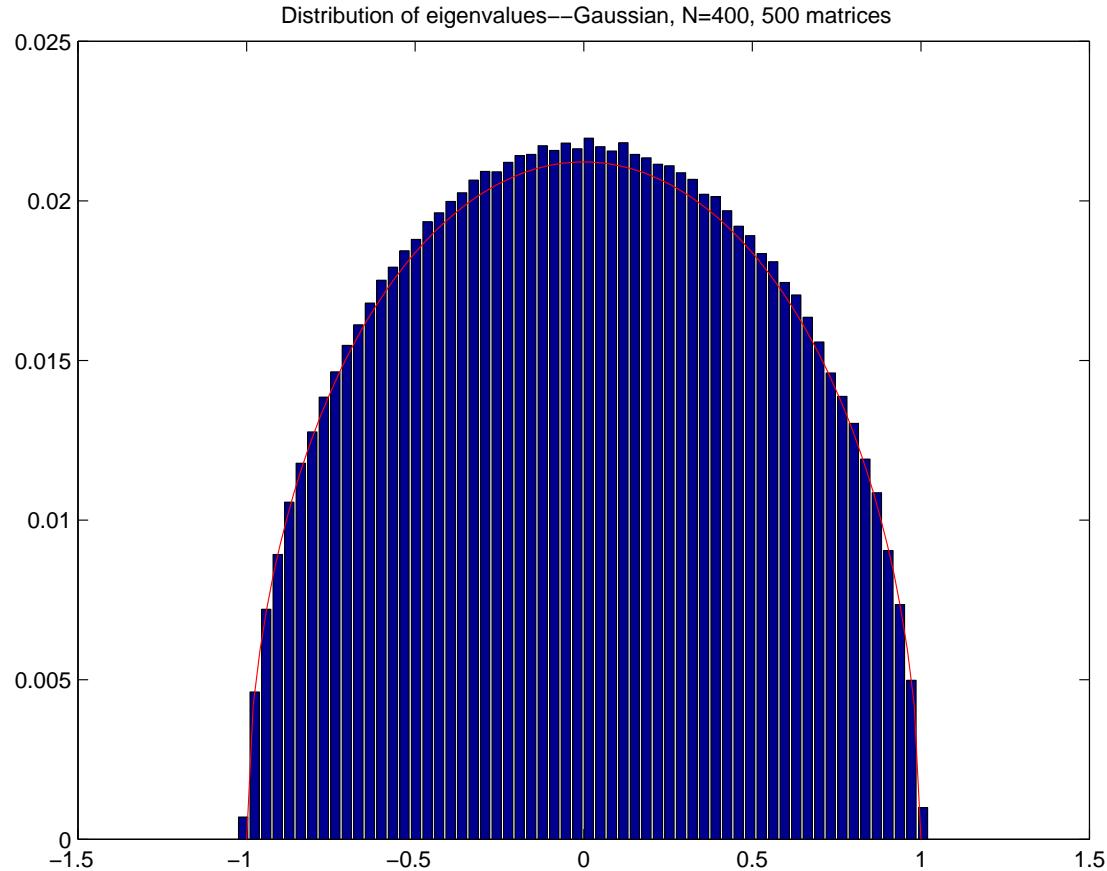
$$\frac{1}{2^2 N^2} \int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty} \sum_{i=1}^N \sum_{j=1}^N a_{ji}^2 \cdot p(a_{11}) da_{11} \cdots p(a_{NN}) da_{NN}$$

Integration factors as

$$\int_{a_{ij}=-\infty}^{\infty} a_{ij}^2 p(a_{ij}) da_{ij} \cdot \prod_{\substack{(k,l) \neq (i,j) \\ k < l}} \int_{a_{kl}=-\infty}^{\infty} p(a_{kl}) da_{kl} = 1.$$

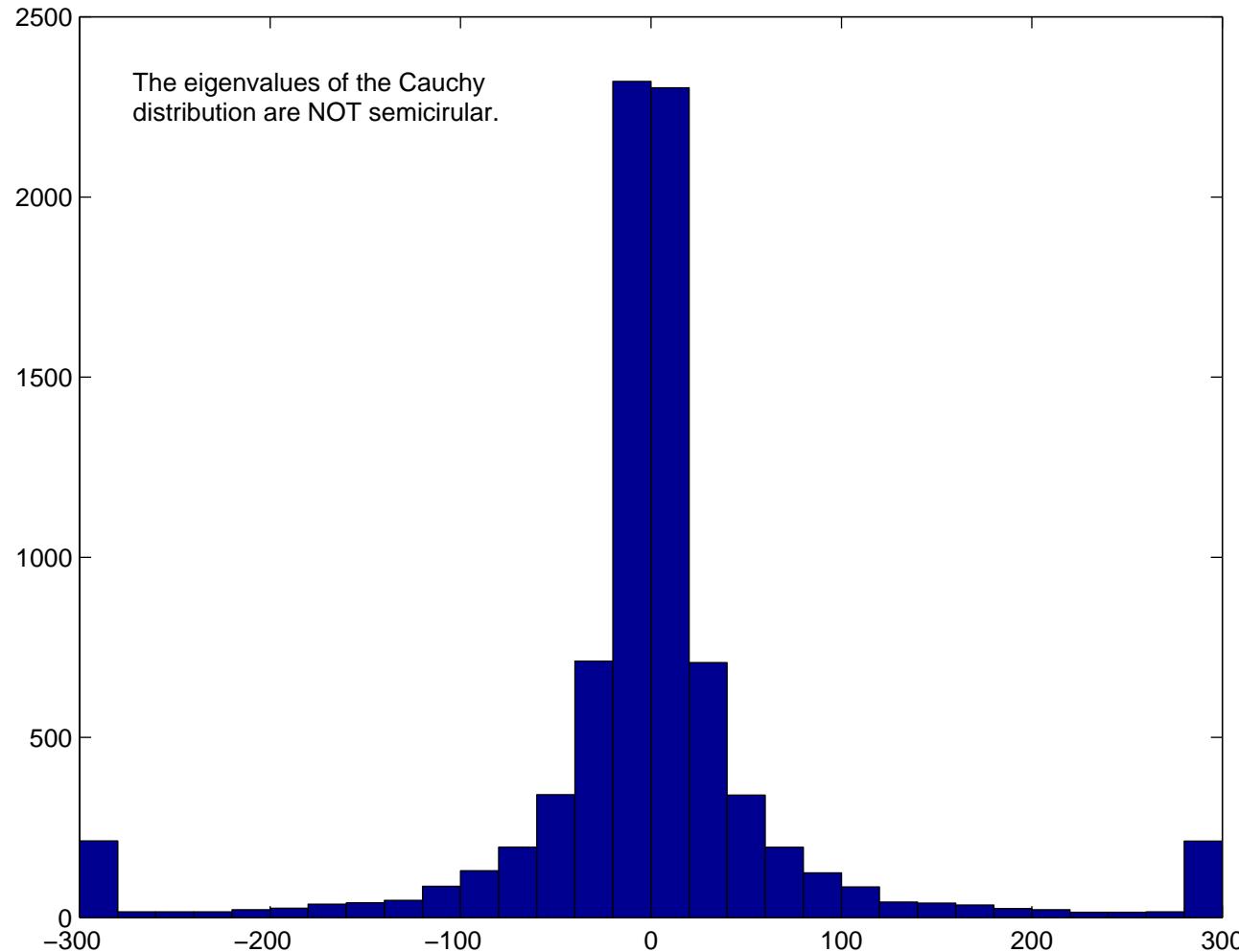
Have  $N^2$  summands, answer is  $\frac{1}{4} = 2^{\text{nd}}$  moment of Semi-Circle.

# Random Matrix Theory: Semi-Circle Law



500 Matrices: Gaussian  $400 \times 400$   
 $p(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$

# Random Matrix Theory: Semi-Circle Law



$$\text{Cauchy Distribution: } p(x) = \frac{1}{\pi(1+x^2)}$$

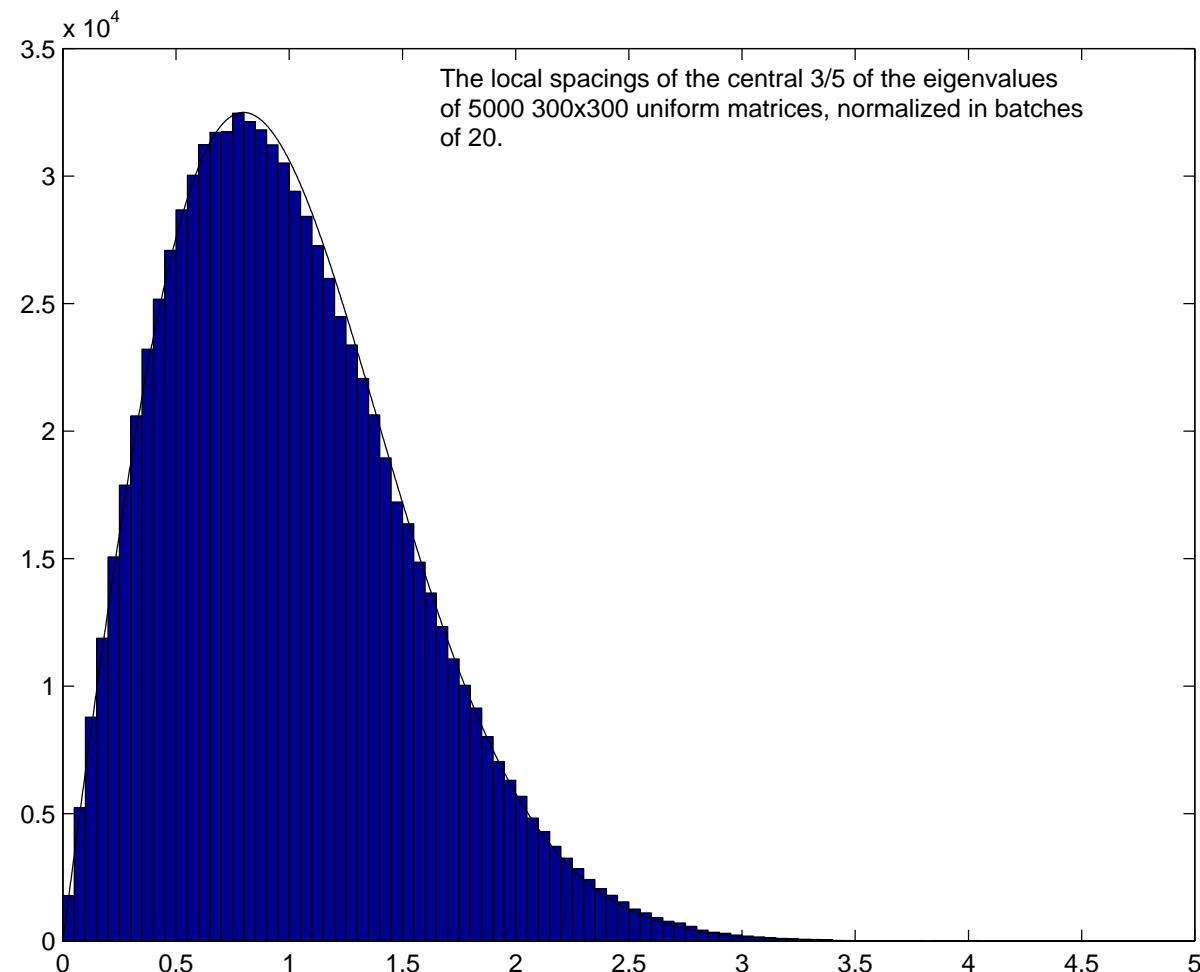
## GOE Conjecture

**GOE Conjecture:** As  $N \rightarrow \infty$ , the probability density of the spacing b/w consecutive normalized eigenvalues approaches a limit independent of  $p$ .

Only known if  $p$  is a Gaussian.

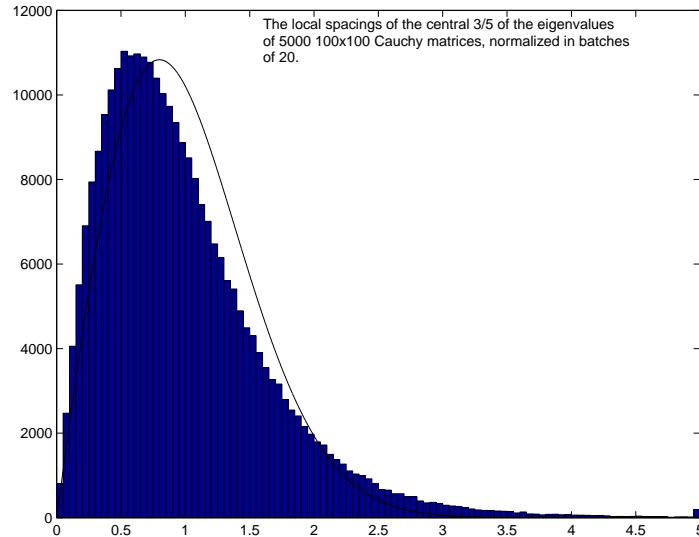
$$\text{GOE}(x) \approx Axe^{-Bx^2}.$$

**Uniform Distribution:**  $p(x) = \frac{1}{2}$  for  $|x| \leq 1$

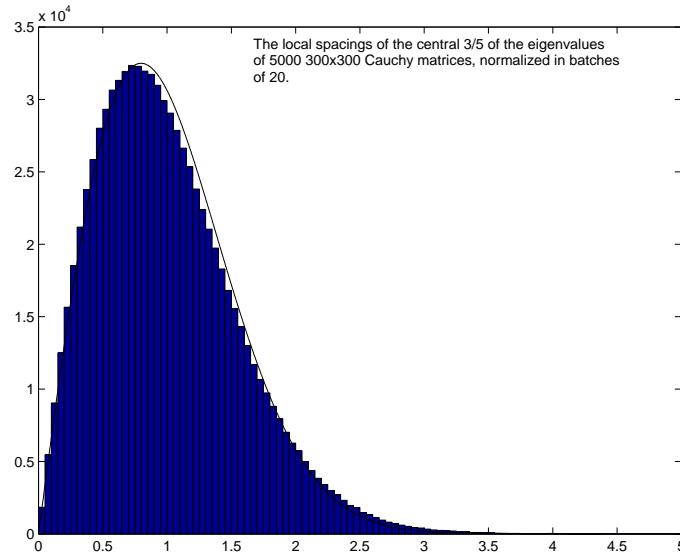


5000:  $300 \times 300$  uniform on  $[-1, 1]$

**Cauchy Distribution:**  $p(x) = \frac{1}{\pi(1+x^2)}$



5000:  $100 \times 100$  Cauchy



5000:  $300 \times 300$  Cauchy

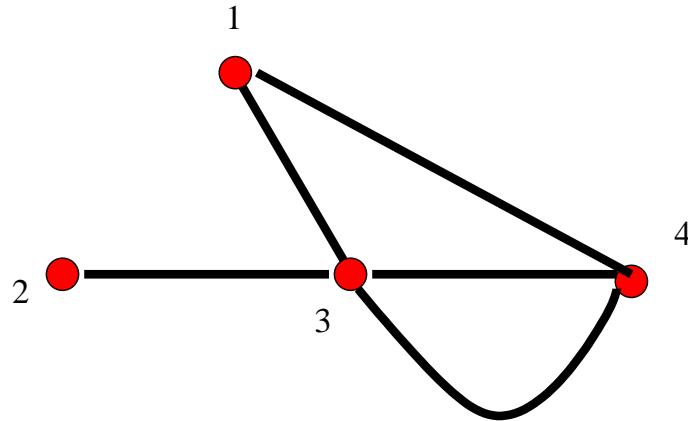
## Fat Thin Families

Need a family **FAT** enough to do averaging.

Need a family **THIN** enough so that everything isn't averaged out.

Real Symmetric Matrices have  $\frac{N(N+1)}{2}$  independent entries.

## Random Graphs



Degree of a vertex = number of edges leaving the vertex.

Adjacency matrix:  $a_{ij}$  = number edges b/w Vertex  $i$  and Vertex  $j$ .

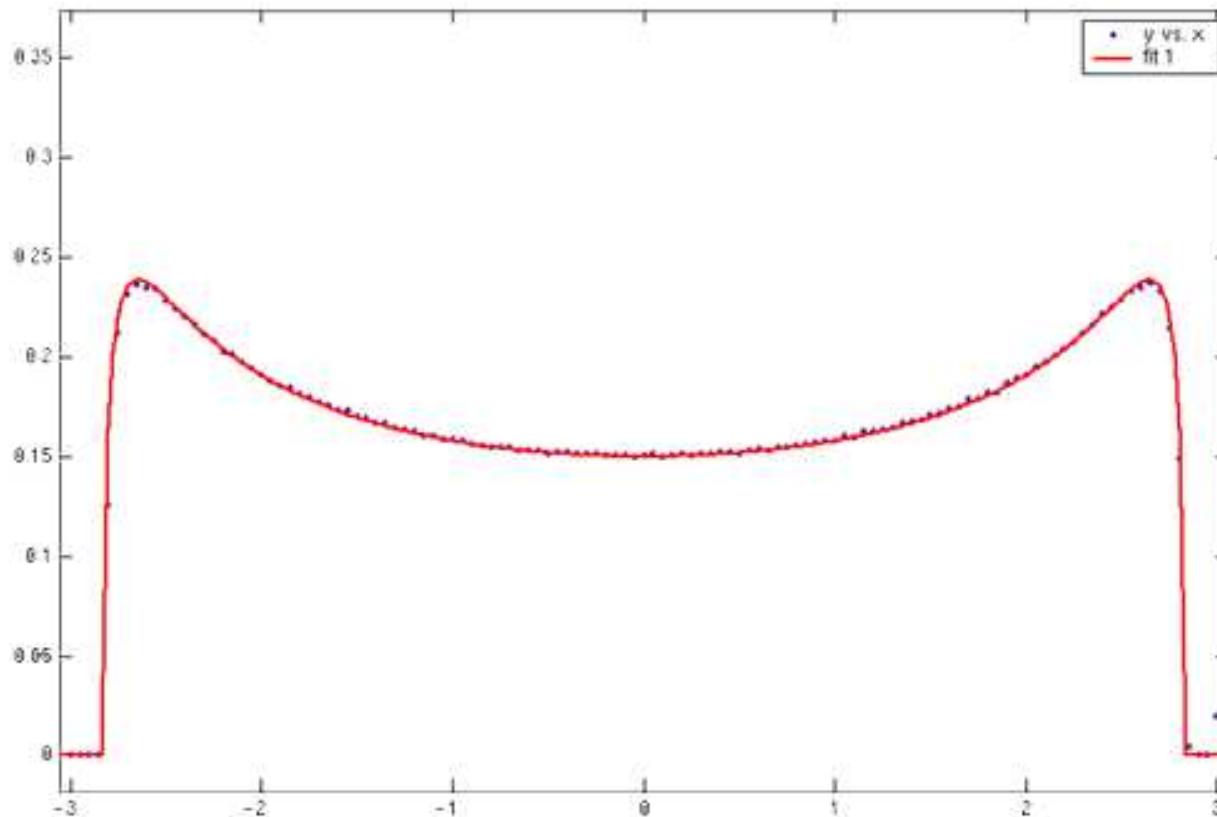
$$A = \begin{pmatrix} 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 1 & 1 & 0 & 2 \\ 1 & 0 & 2 & 0 \end{pmatrix}$$

These are Real Symmetric Matrices.

## McKay's Law (Kesten Measure)

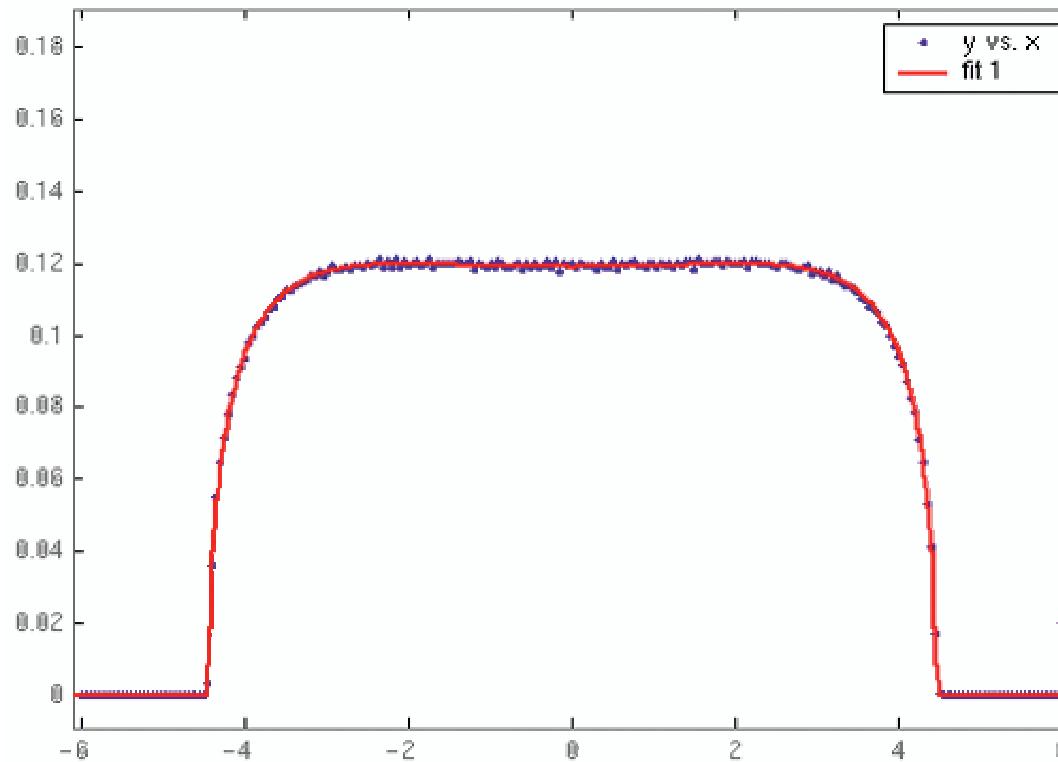
Density of Eigenvalues for  $d$ -regular graphs

$$f(x) = \begin{cases} \frac{d}{2\pi(d^2-x^2)} \sqrt{4(d-1) - x^2} & |x| \leq 2\sqrt{d-1} \\ 0 & \text{otherwise.} \end{cases}$$



$$d = 3.$$

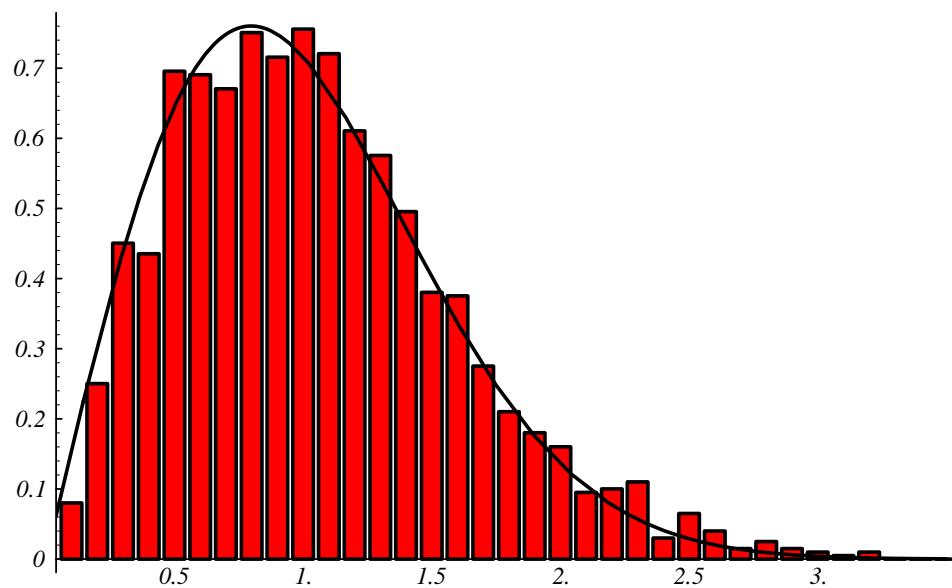
## McKay's Law (Kesten Measure)



$$d = 6.$$

Fat Thin: fat enough to average, thin enough to get something different than Semi-circle.

## 3-Regular, 2000 Vertices and GOE



# **PART III**

# **NUMBER THEORY**

## Riemann Zeta Function

$$\zeta(s) = \sum_{n=1}^{\infty} \frac{1}{n^s} = \prod_{p \text{ prime}} \left(1 - \frac{1}{p^s}\right)^{-1}, \quad \operatorname{Re}(s) > 1.$$

Unique Factorization:  $n = p_1^{r_1} \cdots p_m^{r_m}$ .

$$\begin{aligned} \prod_p \left(1 - \frac{1}{p^s}\right)^{-1} &= \left[1 + \frac{1}{2^s} + \left(\frac{1}{2^s}\right)^2 + \cdots\right] \left[1 + \frac{1}{3^s} + \left(\frac{1}{3^s}\right)^2 + \cdots\right] \cdots \\ &= \sum_n \frac{1}{n^s}. \end{aligned}$$

## Riemann Zeta Function (cont):

$$\zeta(s) = \sum_n \frac{1}{n^s} = \prod_p \left(1 - \frac{1}{p^s}\right)^{-1}, \quad \operatorname{Re}(s) > 1$$
$$\pi(x) = \#\{p : p \text{ is prime}, p \leq x\}$$

Properties of  $\zeta(s)$  and Primes:

- $\lim_{s \rightarrow 1^+} \zeta(s) = \infty, \pi(x) \rightarrow \infty.$
- $\zeta(2) = \frac{\pi^2}{6}, \pi(x) \rightarrow \infty.$

## Riemann Zeta Function (cont):

$$\zeta(s) = \sum_{n=1}^{\infty} \frac{1}{n^s} = \prod_{p \text{ prime}} \left(1 - \frac{1}{p^s}\right)^{-1}, \quad \operatorname{Re}(s) > 1.$$

## Functional Equation:

$$\xi(s) = \Gamma\left(\frac{s}{2}\right)\pi^{-\frac{s}{2}}\zeta(s) = \xi(1-s).$$

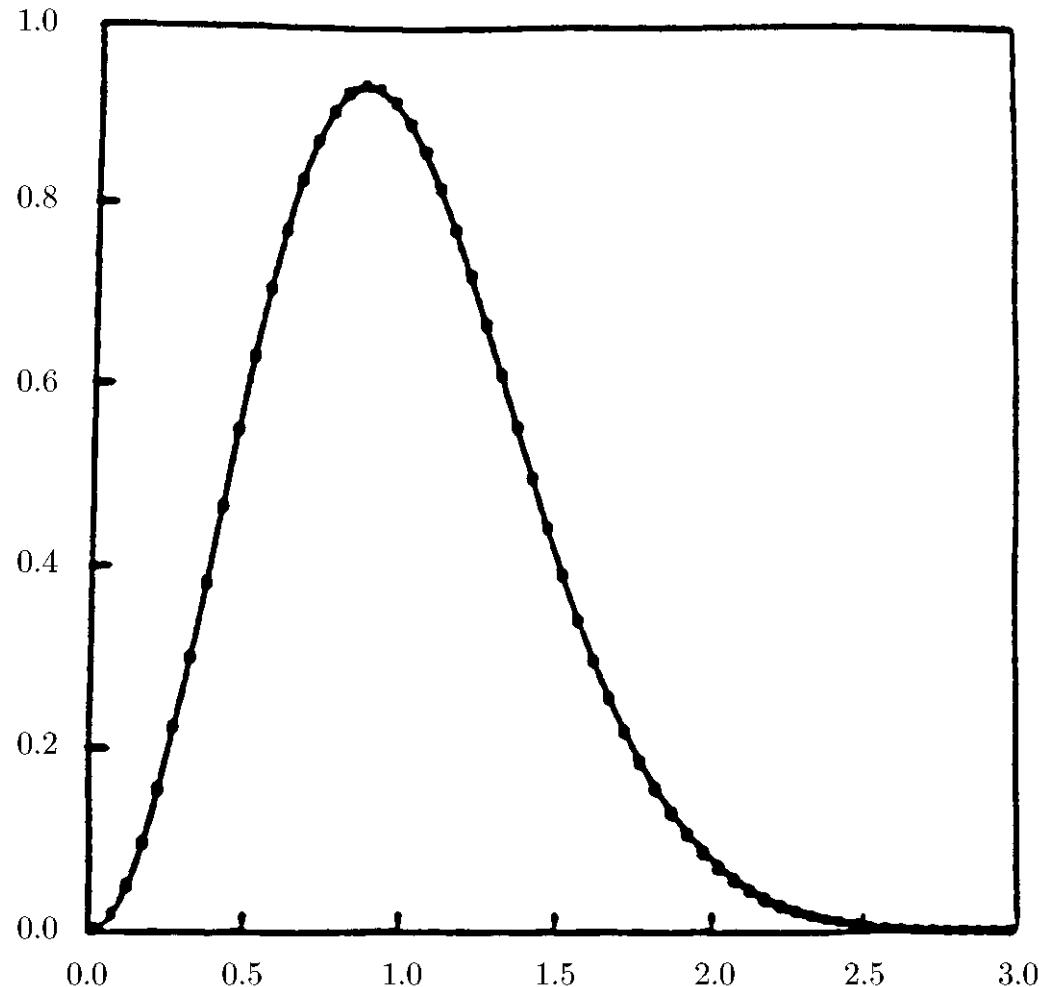
## Riemann Hypothesis (RH):

All non-trivial zeros have  $\operatorname{Re}(s) = \frac{1}{2}$ ; can write zeros as  $\frac{1}{2} + i\gamma$ .

## Observation:

Spacings b/w zeros appear same as b/w eigenvalues of Complex Hermitian matrices ( $\overline{A}^T = A$ )

## Zeros of $\zeta(s)$ vs GUE



70 million spacings b/w adjacent zeros of  $\zeta(s)$ , starting at the  $10^{20}$ th zero  
(from Odlyzko)

## Explicit Formula: (Contour Integration)

$$\begin{aligned}-\frac{\zeta'(s)}{\zeta(s)} &= -\frac{d}{ds} \log \zeta(s) \\&= \frac{d}{ds} \sum_p \log(1 - p^{-s}) \\&= \sum_p \frac{\log p \cdot p^{-s}}{1 - p^{-s}} \\&= \sum_p \frac{\log p}{p^s} + \text{Good}(s).\end{aligned}$$

Contour Integration:

$$\int -\frac{\zeta'(s)}{\zeta(s)} \frac{x^s}{s} ds \quad \text{vs} \quad \sum_p \log p \int \left(\frac{x}{p}\right)^s \frac{ds}{s}.$$

**Knowledge of zeros gives info on coefficients.**

# **PART IV**

# **ELLIPTIC CURVES**

## Elliptic Curves:

$$E_t : y^2 = x^3 + A(T)x + B(T), \quad A(T), B(T) \in \mathbb{Z}(T).$$

$$a_t(p) = - \sum_{x \bmod p} \left( \frac{x^3 + A(t)x + B(t)}{p} \right) = a_{t+mp}(p)$$

$$L(E, s) = \sum_{n=1}^{\infty} \frac{a_E(n)}{n^s} = \prod_p L_p(E, s).$$

By GRH: All zeros on the critical line.

Rational solutions:  $E(\mathbb{Q}) = \mathbb{Z}^r \oplus T$ .

**Birch and Swinnerton-Dyer Conjecture:**

Geometric rank  $r$  = analytic rank (order of vanishing at central point).

## Limiting Behavior

$$\lim_{N \rightarrow \infty} \frac{1}{|\mathcal{F}_N|} \sum_{f \in \mathcal{F}_N} \sum_j \phi\left(\frac{\gamma_f^{(j)} \log N_f}{2\pi}\right) = \int \phi(x) W_{1,\mathcal{G}(\mathcal{F})}(x) dx.$$

**Density Conjecture:** Distribution of low zeros of  $L$ -functions agree with the distribution of eigenvalues near 1 of a classical compact group.

## Tools to Study Low Zeros

- explicit formula relating zeros and Fourier coeffs;
  - ◊ Analogue of Eigenvalue Trace Lemma
- averaging formulas for the family;
  - ◊ Analogue of integration formulas for  $\text{Trace}(A^k)$ .

## Explicit Formula

$$\begin{aligned}
\frac{1}{|\mathcal{F}|} \sum_{E \in \mathcal{F}} \sum_j \phi \left( \frac{\log N_E}{2\pi} \gamma_E^{(j)} \right) &= \frac{1}{|\mathcal{F}|} \sum_{E \in \mathcal{F}} \widehat{\phi}(0) + \phi_i(0) \\
&\quad - \frac{2}{|\mathcal{F}|} \sum_{E \in \mathcal{F}} \sum_p \frac{\log p}{\log N_E p} \frac{1}{p} \widehat{\phi} \left( \frac{\log p}{\log N_E} \right) a_E(p) \\
&\quad - \frac{2}{|\mathcal{F}|} \sum_{E \in \mathcal{F}} \sum_p \frac{\log p}{\log N_E p^2} \frac{1}{p^2} \widehat{\phi} \left( 2 \frac{\log p}{\log N_E} \right) a_E^2(p) \\
&\quad + O \left( \frac{\log \log N_E}{\log N_E} \right)
\end{aligned}$$

Want to move  $\frac{1}{|\mathcal{F}|} \sum_{E \in \mathcal{F}}$ , Leads us to study

$$A_{r,\mathcal{F}}(p) = \sum_{t \bmod p} a_t^r(p), \quad r = 1 \text{ or } 2.$$

## One-Level Result

For small support, one-param family of rank  $r$  over  $\mathbb{Q}(T)$ :

$$\lim_{N \rightarrow \infty} \frac{1}{|\mathcal{F}_N|} \sum_{E \in \mathcal{F}_N} \sum_j \phi\left(\frac{\log N_E}{2\pi} \gamma_E^{(j)}\right) = \int \phi(x) W_{\mathcal{G}}(x) dx + r\phi(0),$$

where

$$\mathcal{G} = \begin{cases} \text{O} & \text{if half odd} \\ \text{SO(even)} & \text{if all even} \\ \text{SO(odd)} & \text{if all odd} \end{cases}$$

**Confirm Katz-Sarnak, B-SD predictions for small support.**

**Forced zeros seem independent.**

## Interesting Families

Let  $\mathcal{E} : y^2 = x^3 + A(T)x + B(T)$  be a one-parameter family of elliptic curves of rank  $r$  over  $\mathbb{Q}$ .

Natural sub-families

- Curves of rank  $r$ .
- Curves of rank  $r + 2$ .

**Question:** Does the sub-family of rank  $r + 2$  curves in a rank  $r$  family behave like the sub-family of rank  $r + 2$  curves in a rank  $r + 2$  family?

Equivalently, does it matter how one conditions on a curve being rank  $r+2$ ?

# Orthogonal Random Matrix Models

RMT:  $2N$  eigenvalues, in pairs  $e^{\pm i\theta_j}$ , probability measure on  $[0, \pi]^N$ :

$$d\epsilon_0(\theta) \propto \prod_{j < k} (\cos \theta_k - \cos \theta_j)^2 \prod_j d\theta_j.$$

## Independent Model:

$$\mathcal{A}_{2N,2r} = \left\{ \begin{pmatrix} I_{2r \times 2r} & \\ & g \end{pmatrix} : g \in SO(2N - 2r) \right\}.$$

## Interaction Model:

Sub-ensemble of  $SO(2N)$  with the last  $2r$  of the  $2N$  eigenvalues equal  $+1$ :

$$d\epsilon_{2r}(\theta) \propto \prod_{j < k} (\cos \theta_k - \cos \theta_j)^2 \prod_j (1 - \cos \theta_j)^{2r} \prod_j d\theta_j,$$

with  $1 \leq j, k \leq N - r$ .

## Comparing the RMT Models

Small support, one-parameter families agree with  $\rho_{r,\text{Indep}}$  and not  $\rho_{r,\text{Inter}}$ .

Curve  $E$ , conductor  $N_E$ , expect first zero  $\frac{1}{2} + i\gamma_E^{(1)}$  with  $\gamma_E^{(1)} \approx \frac{1}{\log N_E}$ .

$r$  zeros at central point, if repulsion of zeros is of size  $\frac{c_r}{\log N_E}$ , can detect in

$$\frac{1}{|\mathcal{F}_N|} \sum_{E \in \mathcal{F}_N} \sum_j \phi\left(\frac{\gamma_E^{(j)} \log N_E}{2\pi}\right).$$

Corrections of size

$$\phi(x_0 + c_r) - \phi(x_0) \approx \phi'(x(x_0, c_r)) \cdot c_r.$$

# Testing Random Matrix Theory Predictions

**First (Normalized) Zero above Central Point:** Do extra zeros at the central point affect the distribution of zeros near the central point?

## RMT: Theoretical Results ( $N \rightarrow \infty$ , Mean $\rightarrow 0.321$ )

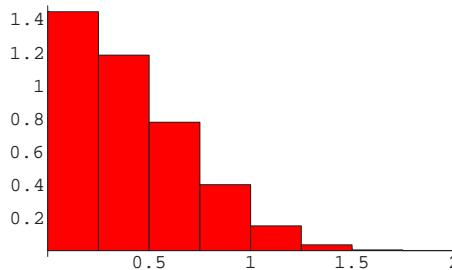


Figure 1a: First normalized eigenangle above 0: 23,040 SO(4) matrices  
Mean = .357, Standard Deviation about the Mean = .302, Median = .357

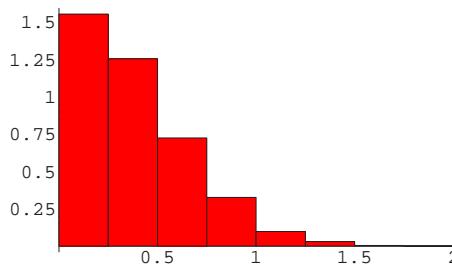


Figure 1b: First normalized eigenangle above 0: 23,040 SO(6) matrices  
Mean = .325, Standard Deviation about the Mean = .284, Median = .325

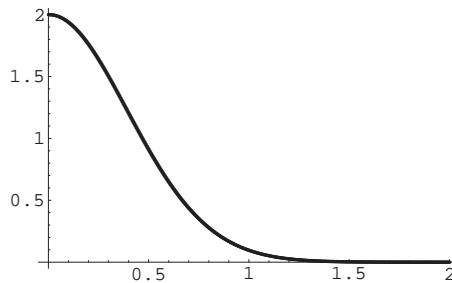


Figure 1c: First normalized eigenangle above 0:  
 $N \rightarrow \infty$  scaling limit of  $\text{SO}(2N)$ : Mean = .321.

## Rank 0 Curves: 1st Normalized Zero above Central Point

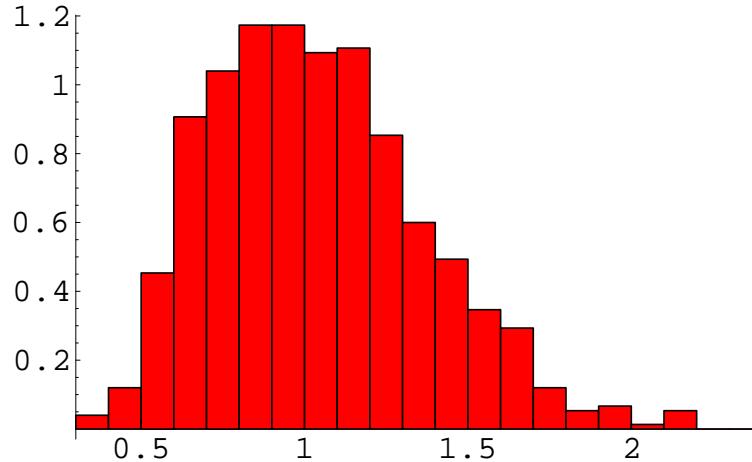


Figure 2a: 750 rank 0 curves from  $y^2 + a_1xy + a_3y = x^3 + a_2x^2 + a_4x + a_6$ .  
 $\log(\text{cond}) \in [3.2, 12.6]$ , median = 1.00 mean = 1.04,  $\sigma_\mu = .32$

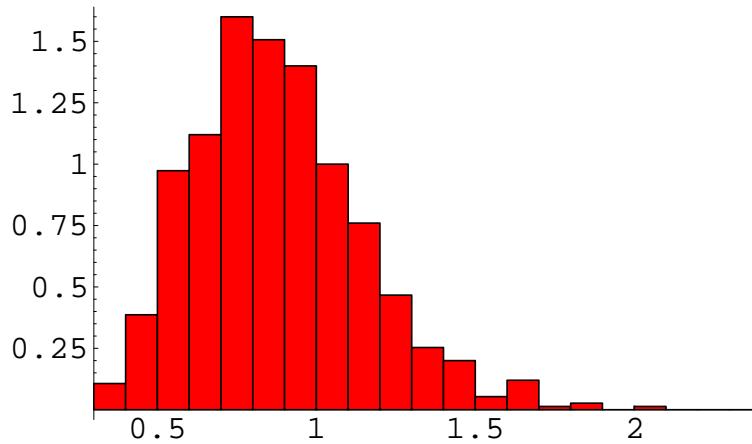


Figure 2b: 750 rank 0 curves from  $y^2 + a_1xy + a_3y = x^3 + a_2x^2 + a_4x + a_6$ .  
 $\log(\text{cond}) \in [12.6, 14.9]$ , median = .85, mean = .88,  $\sigma_\mu = .27$

## Rank 2 Curves: 1st Norm. Zero above the Central Point

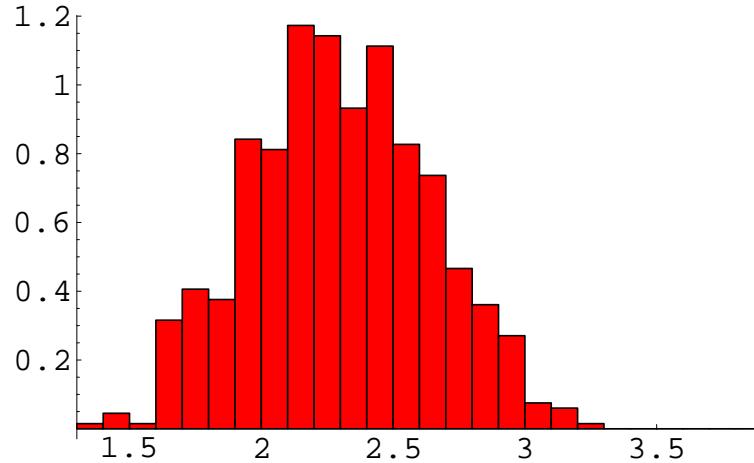


Figure 3a: 665 rank 2 curves from  $y^2 + a_1xy + a_3y = x^3 + a_2x^2 + a_4x + a_6$ .  
 $\log(\text{cond}) \in [10, 10.3125]$ , median = 2.29, mean = 2.30

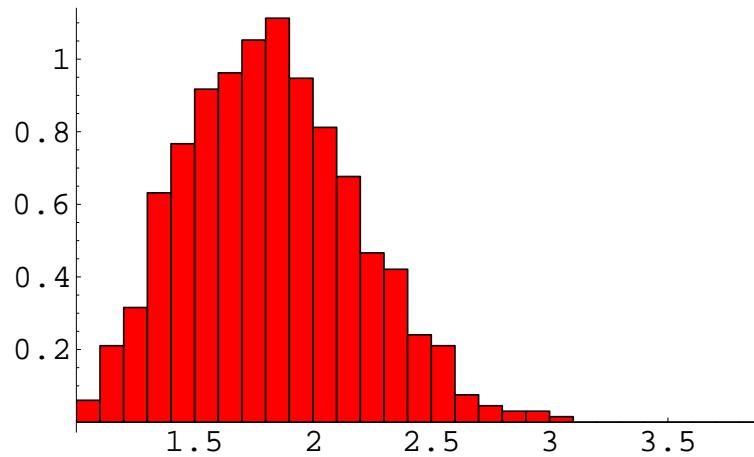


Figure 3b: 665 rank 2 curves from  $y^2 + a_1xy + a_3y = x^3 + a_2x^2 + a_4x + a_6$ .  
 $\log(\text{cond}) \in [16, 16.5]$ , median = 1.81, mean = 1.82

## Rank 0 Curves: 1st Norm Zero: 14 One-Param of Rank 0

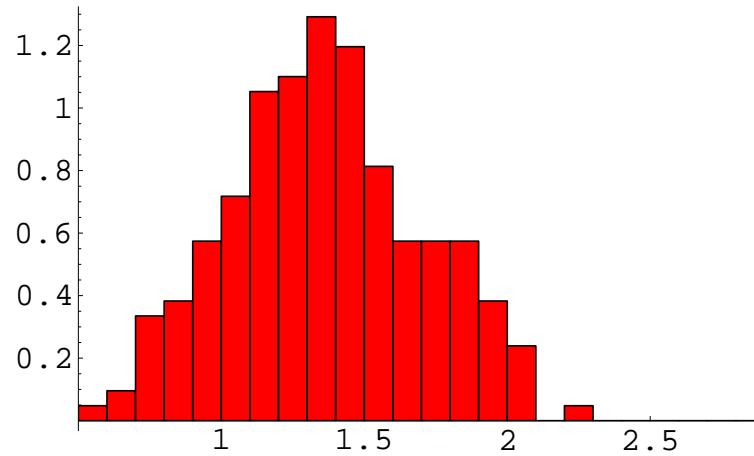


Figure 4a: 209 rank 0 curves from 14 rank 0 families,  
 $\log(\text{cond}) \in [3.26, 9.98]$ , median = 1.35, mean = 1.36

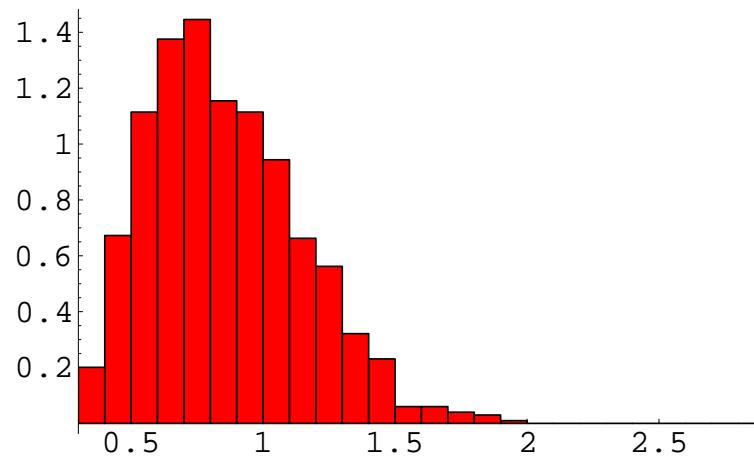


Figure 4b: 996 rank 0 curves from 14 rank 0 families,  
 $\log(\text{cond}) \in [15.00, 16.00]$ , median = .81, mean = .86.

# Rank 0 Curves: 1st Norm Zero: 14 One-Param of Rank 0 over $\mathbb{Q}(T)$

Family	Median $\tilde{\mu}$	Mean $\mu$	StDev $\sigma_\mu$	log(conductor)	Number
1: [0,1,1,1,T]	1.28	1.33	0.26	[3.93, 9.66]	7
2: [1,0,0,1,T]	1.39	1.40	0.29	[4.66, 9.94]	11
3: [1,0,0,2,T]	1.40	1.41	0.33	[5.37, 9.97]	11
4: [1,0,0,-1,T]	1.50	1.42	0.37	[4.70, 9.98]	20
5: [1,0,0,-2,T]	1.40	1.48	0.32	[4.95, 9.85]	11
6: [1,0,0,T,0]	1.35	1.37	0.30	[4.74, 9.97]	44
7: [1,0,1,-2,T]	1.25	1.34	0.42	[4.04, 9.46]	10
8: [1,0,2,1,T]	1.40	1.41	0.33	[5.37, 9.97]	11
9: [1,0,-1,1,T]	1.39	1.32	0.25	[7.45, 9.96]	9
10: [1,0,-2,1,T]	1.34	1.34	0.42	[3.26, 9.56]	9
11: [1,1,-2,1,T]	1.21	1.19	0.41	[5.73, 9.92]	6
12: [1,1,-3,1,T]	1.32	1.32	0.32	[5.04, 9.98]	11
13: [1,-2,0,T,0]	1.31	1.29	0.37	[4.73, 9.91]	39
14: [-1,1,-3,1,T]	1.45	1.45	0.31	[5.76, 9.92]	10
<b>All Curves</b>	1.35	1.36	0.33	[3.26, 9.98]	209
<b>Distinct Curves</b>	1.35	1.36	0.33	[3.26, 9.98]	196

# Rank 0 Curves: 1st Norm Zero: 14 One-Param of Rank 0 over $\mathbb{Q}(T)$

<b>Family</b>	<b>Median <math>\tilde{\mu}</math></b>	<b>Mean <math>\mu</math></b>	<b>StDev <math>\sigma_\mu</math></b>	<b>log(conductor)</b>	<b>Number</b>
1: [0,1,1,1,T]	0.80	0.86	0.23	[15.02, 15.97]	49
2: [1,0,0,1,T]	0.91	0.93	0.29	[15.00, 15.99]	58
3: [1,0,0,2,T]	0.90	0.94	0.30	[15.00, 16.00]	55
4: [1,0,0,-1,T]	0.80	0.90	0.29	[15.02, 16.00]	59
5: [1,0,0,-2,T]	0.75	0.77	0.25	[15.04, 15.98]	53
6: [1,0,0,T,0]	0.75	0.82	0.27	[15.00, 16.00]	130
7: [1,0,1,-2,T]	0.84	0.84	0.25	[15.04, 15.99]	63
8: [1,0,2,1,T]	0.90	0.94	0.30	[15.00, 16.00]	55
9: [1,0,-1,1,T]	0.86	0.89	0.27	[15.02, 15.98]	57
10: [1,0,-2,1,T]	0.86	0.91	0.30	[15.03, 15.97]	59
11: [1,1,-2,1,T]	0.73	0.79	0.27	[15.00, 16.00]	124
12: [1,1,-3,1,T]	0.98	0.99	0.36	[15.01, 16.00]	66
13: [1,-2,0,T,0]	0.72	0.76	0.27	[15.00, 16.00]	120
14: [-1,1,-3,1,T]	0.90	0.91	0.24	[15.00, 15.99]	48
<b>All Curves</b>	0.81	0.86	0.29	[15.00, 16.00]	996
<b>Distinct Curves</b>	0.81	0.86	0.28	[15.00, 16.00]	863

# Rank 2 Curves from $y^2 = x^3 - T^2x + T^2$

## (Rank 2 over $\mathbb{Q}(T)$ )

### 1st Normalized Zero above Central Point

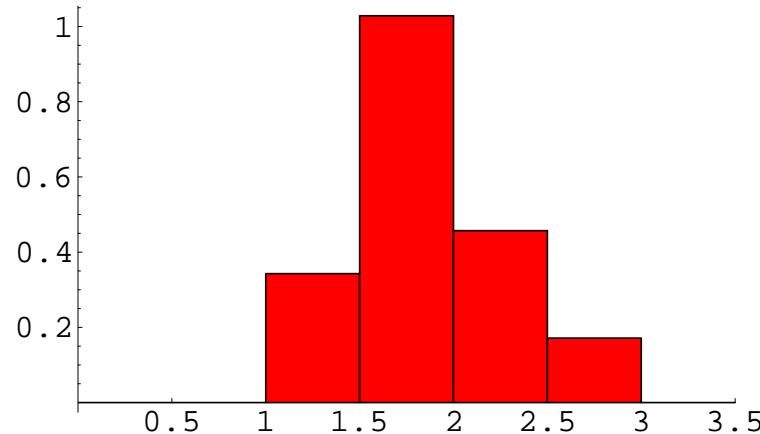


Figure 5a: 35 curves,  $\log(\text{cond}) \in [7.8, 16.1]$ ,  $\tilde{\mu} = 1.85$ ,  $\mu = 1.92$ ,  $\sigma_\mu = .41$

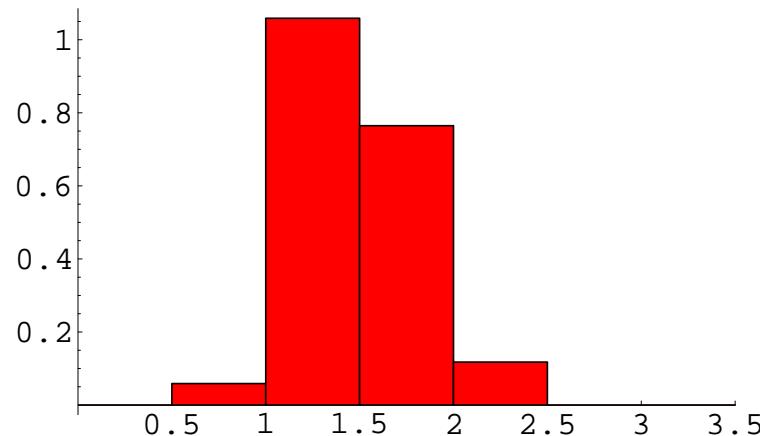


Figure 5b: 34 curves,  $\log(\text{cond}) \in [16.2, 23.3]$ ,  $\tilde{\mu} = 1.37$ ,  $\mu = 1.47$ ,  $\sigma_\mu = .34$

## Repulsion or Attraction?

Conductors in [15, 16]; first set is rank 0 curves from 14 one-parameter families of rank 0 over  $\mathbb{Q}$ ; second set rank 2 curves from 21 one-parameter families of rank 0 over  $\mathbb{Q}$ . The  $t$ -statistics exceed 6.

<b>Family</b>	<b>2nd vs 1st Zero</b>	<b>3rd vs 2nd Zero</b>	<b>Number</b>
Rank 0 Curves	2.16	3.41	863
Rank 2 Curves	1.93	3.27	701

The additional repulsion from extra zeros at the central point cannot be entirely explained by *only* collapsing the first zero to the central point while leaving the other zeros alone.

## Comparison b/w One-Param Families of Different Rank

**First normalized zero above the central point.**

- The first family is the 701 rank 2 curves from the 21 one-parameter families of rank 0 over  $\mathbb{Q}(T)$  with  $\log(\text{cond}) \in [15, 16]$ ;
- the second family is the 64 rank 2 curves from the 21 one-parameter families of rank 2 over  $\mathbb{Q}(T)$  with  $\log(\text{cond}) \in [15, 16]$ .

<b>Family</b>	<b>Median</b>	<b>Mean</b>	<b>Std. Dev.</b>	<b>Number</b>
Rank 2 Curves from Rank 0 Families	1.926	1.936	0.388	701
Rank 2 Curves from Rank 2 Families	1.642	1.610	0.247	64

- $t$ -statistic is 6.60, indicating the means differ.
- The mean of the first normalized zero of rank 2 curves in a family above the central point (for conductors in this range) depends on *how* we choose the curves.

## Spacings b/w Norm Zeros: Rank 0 One-Param Families over $\mathbb{Q}(T)$

- All curves have  $\log(\text{cond}) \in [15, 16]$ ;
- $z_j$  = imaginary part of  $j^{\text{th}}$  normalized zero above the central point;
- 863 rank 0 curves from the 14 one-param families of rank 0 over  $\mathbb{Q}(T)$ ;
- 701 rank 2 curves from the 21 one-param families of rank 0 over  $\mathbb{Q}(T)$ .

	<b>863 Rank 0 Curves</b>	<b>701 Rank 2 Curves</b>	<b>t-Statistic</b>
<b>Median</b> $z_2 - z_1$	1.28	1.30	
<b>Mean</b> $z_2 - z_1$	1.30	1.34	-1.60
<b>StDev</b> $z_2 - z_1$	0.49	0.51	
<b>Median</b> $z_3 - z_2$	1.22	1.19	
<b>Mean</b> $z_3 - z_2$	1.24	1.22	0.80
<b>StDev</b> $z_3 - z_2$	0.52	0.47	
<b>Median</b> $z_3 - z_1$	2.54	2.56	
<b>Mean</b> $z_3 - z_1$	2.55	2.56	-0.38
<b>StDev</b> $z_3 - z_1$	0.52	0.52	

## Spacings b/w Norm Zeros: Rank 0 One-Param Families over $\mathbb{Q}(T)$

- While the normalized zeros are repelled from the central point (and by different amounts for the two sets), the *differences* between the normalized zeros are statistically independent of this repulsion ( $t$ -statistics  $< 2$ ).
- While for a given range of log-conductors the average second normalized zero of a rank 0 curve is close to the average first normalized zero of a rank 2 curve, they are not the same and the additional repulsion from extra zeros at the central point cannot be entirely explained by *only* collapsing the first zero to the central point while leaving the other zeros alone.

## Rank 2 Curves from Rank 0 & Rank 2 Families over $\mathbb{Q}(T)$

- All curves have  $\log(\text{cond}) \in [15, 16]$ ;
- $z_j$  = imaginary part of the  $j^{\text{th}}$  norm zero above the central point;
- 701 rank 2 curves from the 21 one-param families of rank 0 over  $\mathbb{Q}(T)$ ;
- 64 rank 2 curves from the 21 one-param families of rank 2 over  $\mathbb{Q}(T)$ .

	<b>701 Rank 2 Curves</b>	<b>64 Rank 2 Curves</b>	<b>t-Statistic</b>
<b>Median</b> $z_2 - z_1$	1.30	1.26	
<b>Mean</b> $z_2 - z_1$	1.34	1.36	0.69
<b>StDev</b> $z_2 - z_1$	0.51	0.50	
<b>Median</b> $z_3 - z_2$	1.19	1.22	
<b>Mean</b> $z_3 - z_2$	1.22	1.29	1.39
<b>StDev</b> $z_3 - z_2$	0.47	0.49	
<b>Median</b> $z_3 - z_1$	2.56	2.66	
<b>Mean</b> $z_3 - z_1$	2.56	2.65	1.93
<b>StDev</b> $z_3 - z_1$	0.52	0.44	

# **PART V**

# **CONCLUSIONS**

# Correspondences

Similarities b/w Nuclei and Primes:

Energy Levels  $\longleftrightarrow$  Zeros of  $L$ -Functions

Neutron Energy  $\longleftrightarrow$  Summation Lemmas (Test Fn Support)

Different Elements: U, Pu, ...  $\longleftrightarrow$  Different  $L$ -Functions or Families

## Summary

- Similar behavior in different systems.
- Find correct scale.
- Average over similar elements.
- Need a Trace Lemma.
- Thin subsets can exhibit very different behavior.

# Open Problems

## Real Symmetric Band Matrices

$$\begin{pmatrix} a_{1,1} & a_{1,2} & 0 & \cdots & 0 & 0 & 0 \\ a_{1,2} & a_{2,2} & a_{2,3} & \cdots & 0 & 0 & 0 \\ 0 & a_{2,3} & a_{3,3} & \cdots & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & a_{N-2,N-2} & a_{N-2,N} & 0 \\ 0 & 0 & 0 & \cdots & a_{N-2,N} & a_{N-1,N-1} & a_{N-1,N} \\ 0 & 0 & 0 & \cdots & 0 & a_{N-1,N} & a_{N,N} \end{pmatrix}$$

## Real Symmetric Toeplitz Matrices

$$\begin{pmatrix} b_0 & b_1 & b_2 & \cdots & b_{N-1} \\ b_1 & b_0 & b_1 & \cdots & b_{N-2} \\ b_2 & b_1 & b_0 & \cdots & b_{N-3} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ b_{N-1} & b_{N-2} & b_{N-3} & \cdots & b_0 \end{pmatrix}$$

## Rates of Convergence