

# From the Manhattan Project to Number Theory: How Nuclear Physics Helps Us Understand Primes

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## Introduction

## Fundamental Problem: Spacing Between Events

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

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- Spacings b/w Energy Levels of Nuclei.
- Spacings b/w Eigenvalues of Matrices.
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- Spacings b/w Energy Levels of Nuclei.
- Spacings b/w Eigenvalues of Matrices.
- Spacings b/w Primes.
- Bus routes in Cuernavaca, Mexico.
- Scandinavian trees?

## Background Material: Linear Algebra

### Eigenvalue, Eigenvector

Say  $\vec{v} \neq \vec{0}$  is an eigenvector of  $A$  with eigenvalue  $\lambda$  if  $A\vec{v} = \lambda\vec{v}$ .

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Example:

$$\begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = 3 \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$
$$\begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = -1 \begin{pmatrix} 1 \\ -1 \end{pmatrix}.$$



## Background Material: Probability

### Probability Density

A random variable  $X$  has a probability density  $p(x)$  if

- $p(x) \geq 0$ ;
- $\int_{-\infty}^{\infty} p(x) dx = 1$ ;
- $\text{Prob}(X \in [a, b]) = \int_a^b p(x) dx$ .

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### Examples:

- 1 Exponential:  $p(x) = e^{-x/\lambda}/\lambda$  for  $x \geq 0$ ;
- 2 Normal:  $p(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x-\mu)^2/2\sigma^2}$ ;
- 3 Uniform:  $p(x) = \frac{1}{b-a}$  for  $a \leq x \leq b$  and 0 otherwise.

## Background Material: Central Limit Theorem

$k^{\text{th}}$  moment:  $\int_{-\infty}^{\infty} x^k p(x) dx$ .

### Central Limit Theorem

Let  $X_1, X_2, \dots$  be independent, identically distributed random variables with mean  $\mu$ , standard deviation  $\sigma$  and finite higher moments. Then

$$Y_n = \frac{\frac{X_1 + \dots + X_N}{N} - \mu}{\sigma / \sqrt{N}}, \quad \lim_{N \rightarrow \infty} Y_N = N(0, 1).$$

- Universality.
- Rate of convergence depends on higher moments.

## Classical Random Matrix Theory

## Origins of Random Matrix Theory

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**Fundamental Equation:**

$$H\psi_n = E_n\psi_n$$

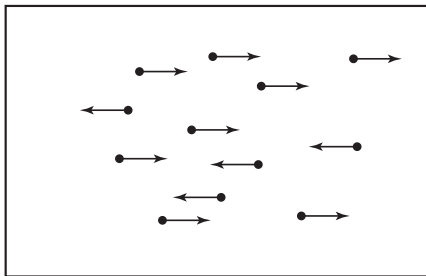
$H$  : matrix, entries depend on system

$E_n$  : energy levels

$\psi_n$  : energy eigenfunctions



## Origins (continued)



- Nuclear physics: choose matrix at random, calculate eigenvalues, average over matrices (real Symmetric  $A = A^T$ , complex Hermitian  $\bar{A}^T = A$ ).

## Random Matrix Ensembles

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

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

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This means

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

## Eigenvalue Trace Lemma

Want to understand the eigenvalues of  $A$ , but it is the matrix elements that are chosen randomly and independently.

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### Eigenvalue Trace Lemma

Let  $A$  be an  $N \times N$  matrix with eigenvalues  $\lambda_i(A)$ . Then

$$\text{Trace}(A^k) = \sum_{n=1}^N \lambda_i(A)^k,$$

where

$$\text{Trace}(A^k) = \sum_{i_1=1}^N \cdots \sum_{i_k=1}^N a_{i_1 i_2} a_{i_2 i_3} \cdots a_{i_N i_1}.$$

## Density of States

## Wigner's Semi-Circle Law

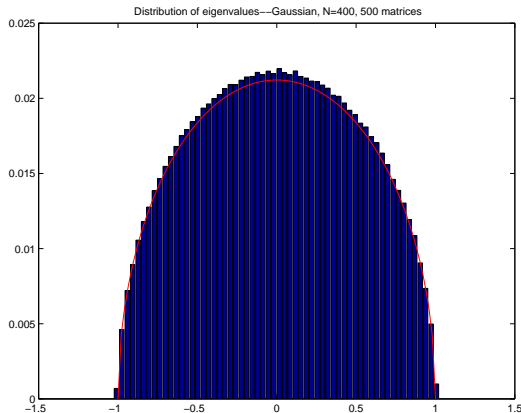
### Wigner's Semi-Circle Law

$N \times N$  real symmetric matrices, entries i.i.d.r.v. from a fixed  $p(x)$  with mean 0, variance 1, and other moments finite. Then for almost all  $A$ , as  $N \rightarrow \infty$

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



## Numerical example: Gaussian density

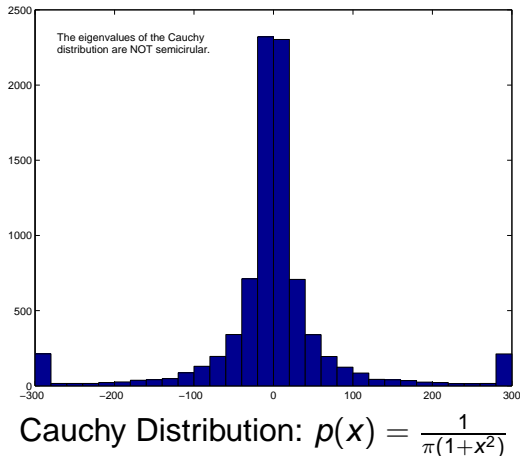


500 Matrices: Gaussian  $400 \times 400$

$$p(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$

## Numerical example: Cauchy density $p(x) = 1/\pi(1 + x^2)$

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## Spacings between events

## GOE Conjecture

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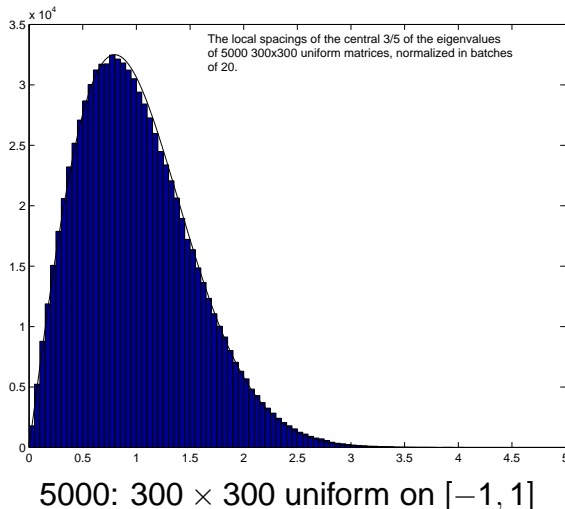
As  $N \rightarrow \infty$ , the probability density of the spacing b/w consecutive normalized eigenvalues approaches a limit independent of  $p$ .

Only known if  $p$  is a Gaussian.

$$\text{GOE}(x) \approx \frac{\pi}{2} x e^{-\pi x^2/4}.$$

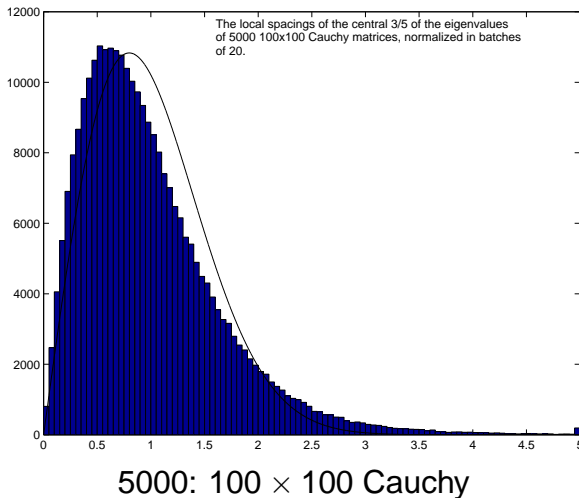
## Numerical Experiment: Uniform Distribution

Let  $p(x) = \frac{1}{2}$  for  $|x| \leq 1$ .



# Cauchy Distribution

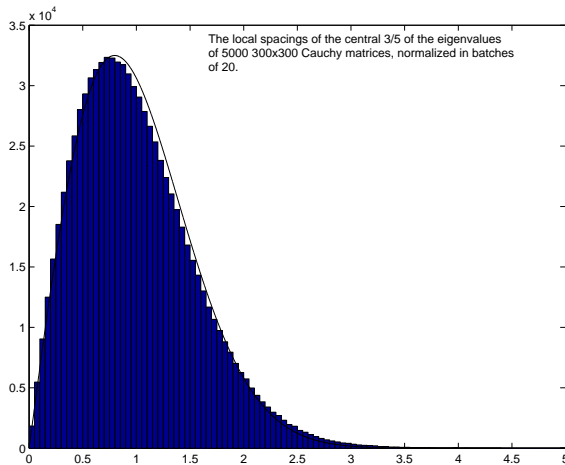
$$\text{Let } p(x) = \frac{1}{\pi(1+x^2)}.$$





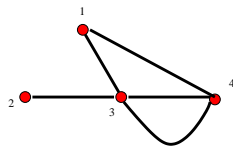
# Cauchy Distribution

$$\text{Let } p(x) = \frac{1}{\pi(1+x^2)}.$$



5000:  $300 \times 300$  Cauchy

## Random Graphs



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

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

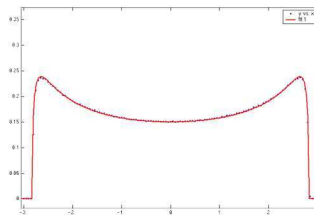
$$A = \begin{pmatrix} 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 1 & 1 & 0 & 2 \\ 1 & 0 & 2 & 0 \end{pmatrix}$$

These are Real Symmetric Matrices.

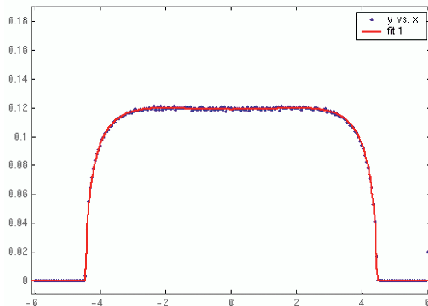
# McKay's Law (Kesten Measure) with $d = 3$

Density of Eigenvalues for  $d$ -regular graphs

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



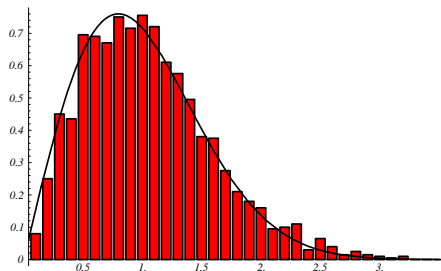
## McKay's Law (Kesten Measure) with $d = 6$



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

### 3-Regular, 2000 Vertices and GOE

Spacings between eigenvalues of 3-regular graphs and the GOE:



## Introduction to $L$ -Functions

## Riemann Zeta Function

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

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



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**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)

$$\zeta(s) = \sum_n \frac{1}{n^s} = \prod_p \left(1 - \frac{1}{p^s}\right)^{-1}, \quad \text{Re}(s) > 1$$

$$\pi(x) = \#\{p : p \text{ is prime}, p \leq x\}$$

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

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

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

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$$\zeta(s) = \sum_n \frac{1}{n^s} = \prod_p \left(1 - \frac{1}{p^s}\right)^{-1}, \quad \text{Re}(s) > 1$$

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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

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

### Functional Equation:

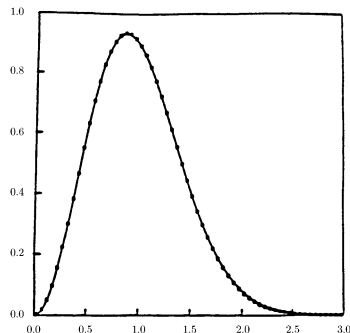
$$\xi(s) = \Gamma\left(\frac{s}{2}\right) \pi^{-\frac{s}{2}} \zeta(s) = \xi(1-s).$$

### Riemann Hypothesis (RH):

All non-trivial zeros have  $\text{Re}(s) = \frac{1}{2}$ ; 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  $10^{20\text{th}}$  zero (from Odlyzko)

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