

Random Matrix Theory with Power-Law Tails

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GERNOT AKEMANN

+ PIERPAOLO VIVO arXiv:0806.1861 (JSTAT)

Brunel
UNIVERSITY
WEST LONDON

EPSRC Engineering and Physical Sciences
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Outline:

1. **Motivation:** Applications
2. **What is Random Matrix Theory (RMT)?**
 - basic properties and results
3. **Why Power Laws:**
 - comparison to data
 - new universality classes
4. **Open Problems**

Motivation: RMT and Applications

- **general idea:**

describe complicated = many body, interacting system
by random variables

$$H\Psi_i = \lambda_i\Psi_i \longrightarrow \exp[-\sum H_{ij}H_{ji}^\dagger]$$

e.g.: Hamiltonian of Nucleus or Quantum Billiard,
Dirac Operator \mathcal{D} in **QCD**

- [Bohigas, Giannoni, Schmidt / Casati, Vals-Gris, Guarnieri 80's]:

Quantum Chaotic Systems display RMT Statistics!

- **RMT: eigenvalues correlated**

- based on global symmetries: universal

- analytically solvable for: spectrum $\rho(\lambda) \sim \langle \sum_i \delta(\lambda - \lambda_i) \rangle$,

individual eigenvalues, consecutive eigenvalue spacings

What is RMT

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where $i = 1, \dots, N$ and $j = 1, \dots, T$

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Wishard/Laguerre/chiral Ensemble

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- correlation functions: $\langle \mathbb{O} \rangle \equiv \mathcal{Z}^{-1} \int d\mathbf{X} \mathbb{O}(\mathbf{X}) P[\mathbf{X}]$
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- eigenvalues: $\mathbf{X}\mathbf{X}^\dagger \rightarrow U\Lambda U^\dagger$ with $\lambda_{i=1, \dots, N} \in \mathbb{R}_+$

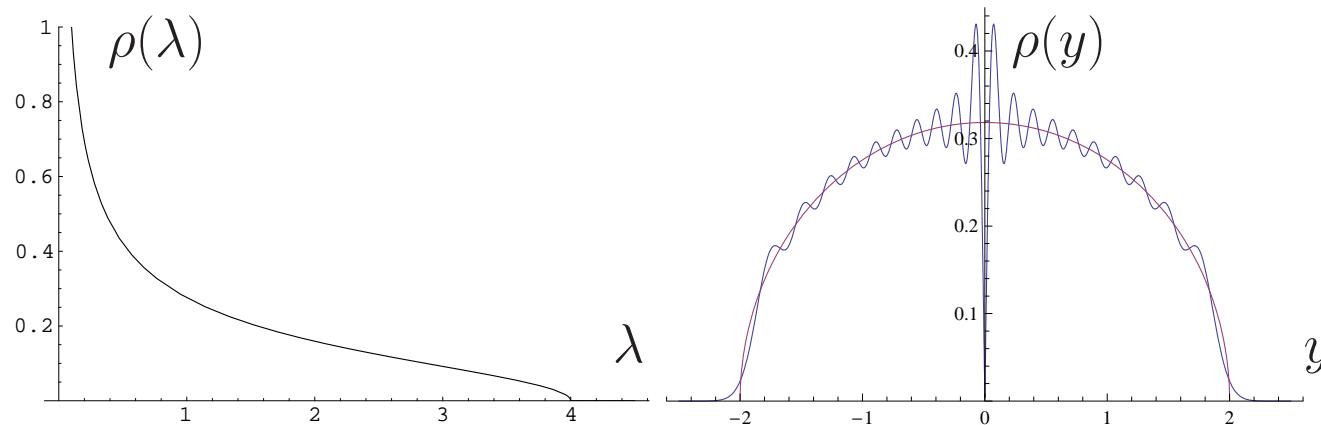
$$\mathcal{Z} = \int d\mathbf{X} P[\mathbf{X}] \rightarrow \int_0^\infty \prod_i d\lambda_i \lambda_i^{T-N} e^{-NV(\lambda_i)} \prod_{k>l} |\lambda_k - \lambda_l|^2$$

→ **2D Coulomb gas on 1D:**

$$\mathcal{S} = (T - N) \sum_i \log[\lambda_i] - N \sum_i V(\lambda_i) + \sum_{k \neq l} \log |\lambda_k - \lambda_l|$$

Large- N Limits

1) semi-circle: $T - N = \mathcal{O}(1)$

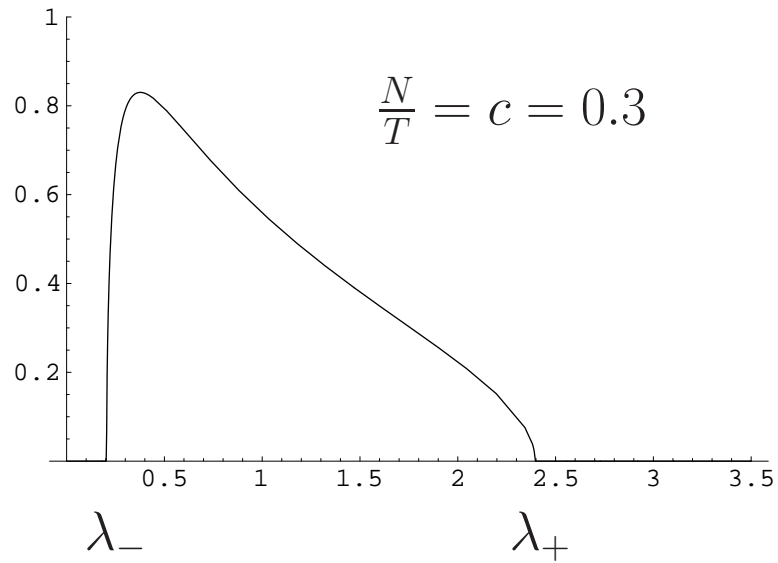


$$\rho(\lambda) \sim \sqrt{\frac{4}{\lambda} - 1} \text{ or in variable } \lambda \rightarrow y^2 : \hat{\rho}(y) \sim \sqrt{4 - \lambda^2}$$

- **local behaviour universal:** zoom $x = (\lambda - \lambda_0)N^\delta$

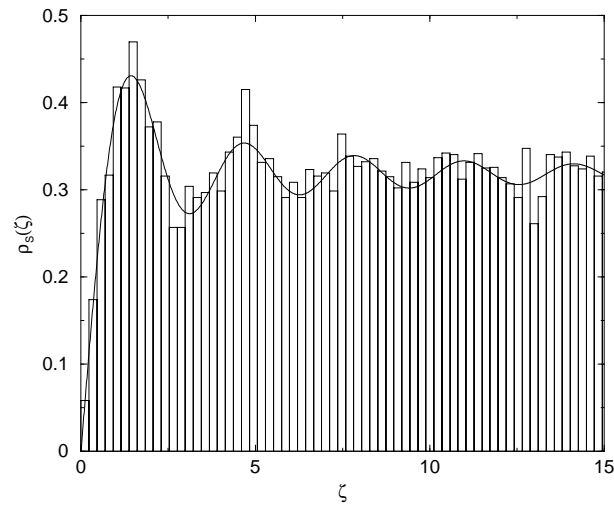
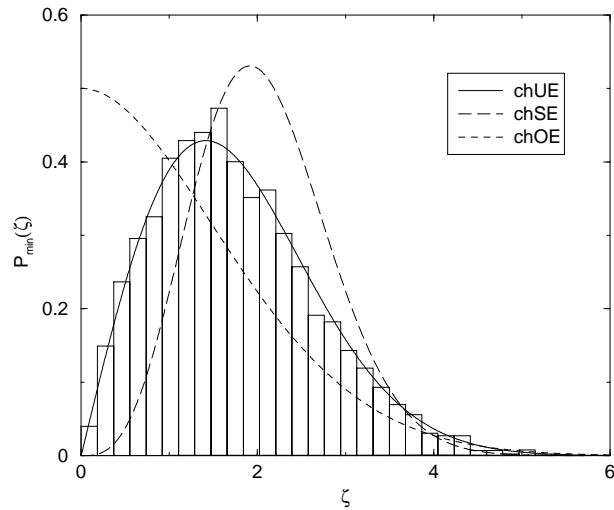
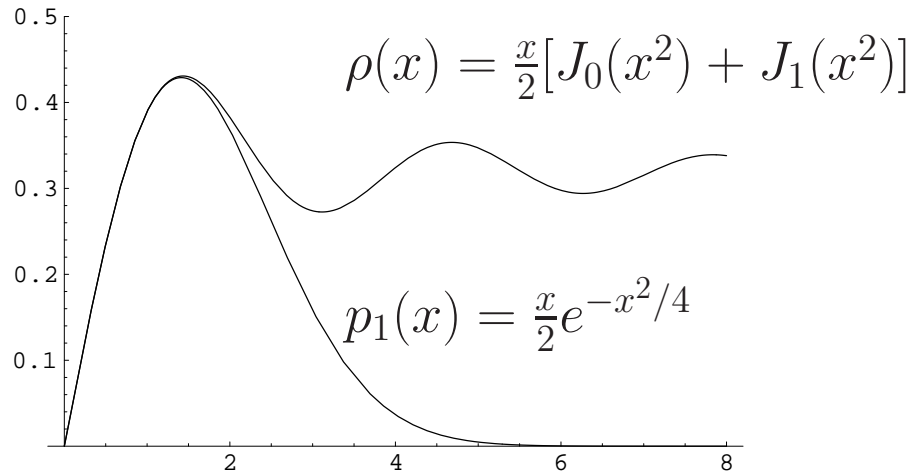
correlations: origin = Bessel; bulk = Sine; edge = Airy

2) Marčenko-Pastur: $T - N = \mathcal{O}(N)$



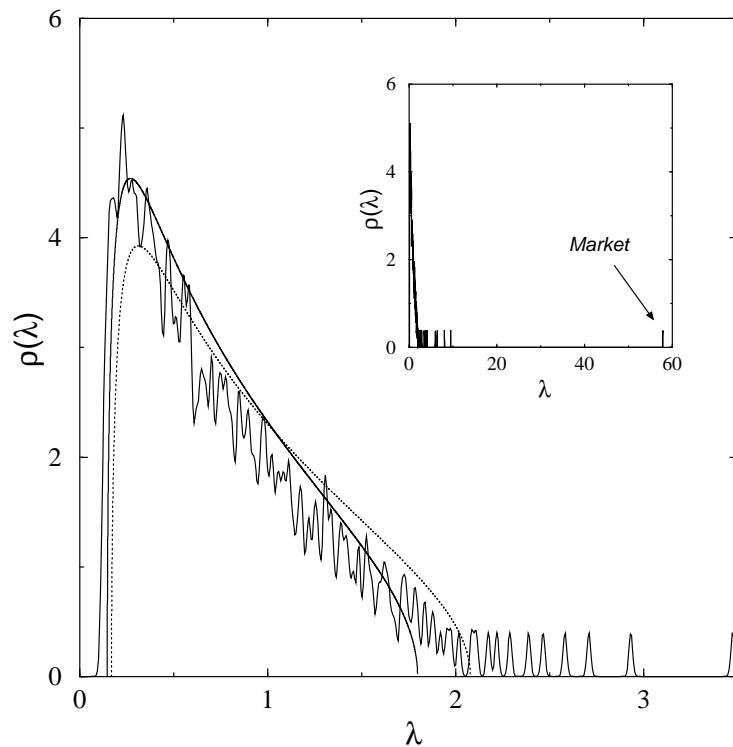
$$\rho(\lambda) \sim \frac{1}{\lambda} \sqrt{(\lambda - \lambda_+(c))(\lambda_-(c) - \lambda)}$$

Example 1): Origin Limit vs. QCD



[Damgaard et al., 98, 01] $x = \lambda_{\mathcal{D}} V \langle \bar{q}q \rangle$ rescaled ev. $\mathcal{D} = \begin{pmatrix} 0 & i\mathbf{X} \\ i\mathbf{X}^\dagger & 0 \end{pmatrix}$

Example 2): M-P vs. Finance



[Plerou et al., 01]

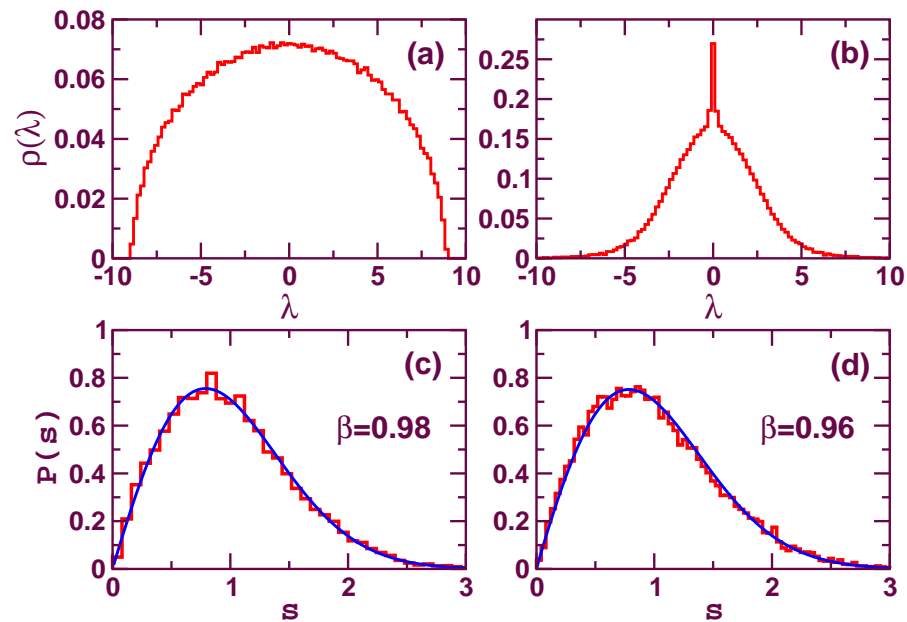
Correlation matrices of times series of stock prices $S_i(t)$

$$X_{it} = \sigma_i^{-1}[G_i(t) - \langle G_i(t) \rangle]$$

with normalised return $G_i(t) = \log[S_i(t+\Delta t)] - \log[S_i(t)]$

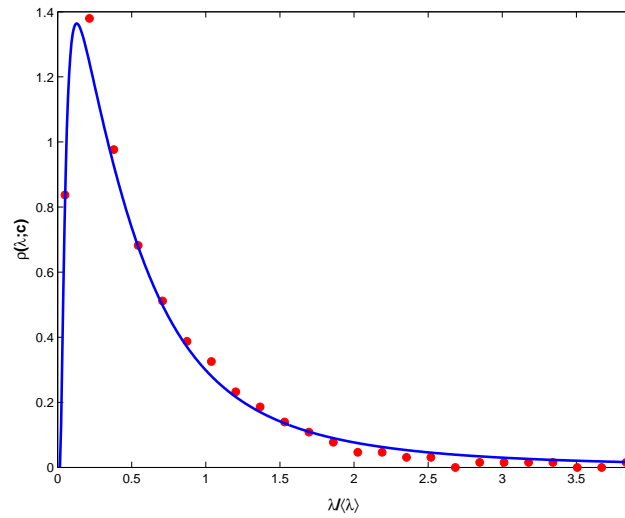
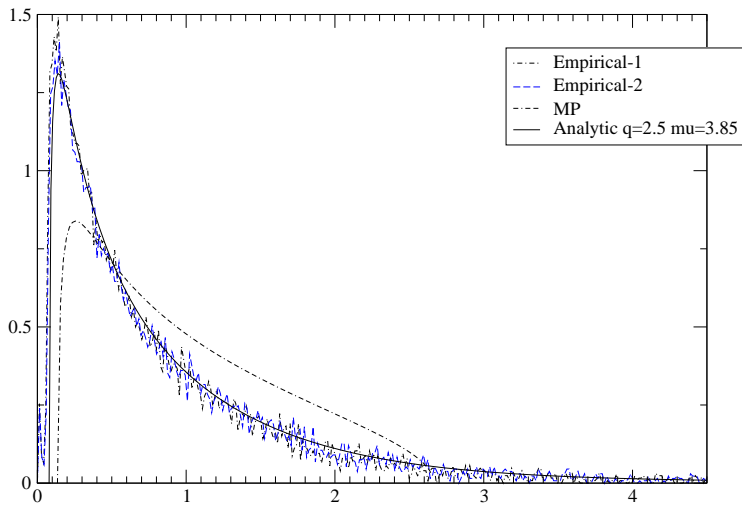
Where are the Power-Laws?

- general feature in complex networks: [Barabási, Albert 99; Bandyopadhyay, Jalan 07]



spectrum of **adjacency matrix** of random network (Erdős-Reyni) **left** and scale-free network **right**

- stock prices have power law tails [Biroli, Bouchaud, Potters 07; G.A., Fischmann, Vivo 08]



RMTs with Heavy Tails

- correlation matrix $\langle \mathbf{X}_{it} \mathbf{X}_{jt'} \rangle \sim \mathbf{C}_{ij} \mathbf{A}_{tt'}$
- empirical correlation matrix $c_{ij} = \frac{1}{T} \sum_{t=1}^T \mathbf{X}_{it} \mathbf{X}_{jt}$
 - compare to Gauß $\frac{1}{T} \text{Tr} \mathbf{X} \mathbf{X}^\dagger$: Marčenko-Pastur
- **Power-Law RMT: non-invariant**
 - incl. correlation $\text{Tr} \mathbf{X} \mathbf{X}^\dagger \rightarrow \text{Tr} \mathbf{X} \mathbf{C}^{-1} \mathbf{X}^\dagger \mathbf{A}^{-1}$ [Burda et al. 07]
 - variance random variable: $\mathbf{X}_{it} \rightarrow \sigma^t \mathbf{X}_{it}$ [Biroli et al. 07]
- **Power-Law RMT: invariant** [Bohigas et al. 04; G.A., Vivo 08]

$$P[\mathbf{X} \mathbf{X}^\dagger] \rightarrow \frac{1}{(1 + \gamma^{-1} \text{Tr} \mathbf{X} \mathbf{X}^\dagger)^\gamma}$$

- can be derived from generalised entropy principle

Analytic Results

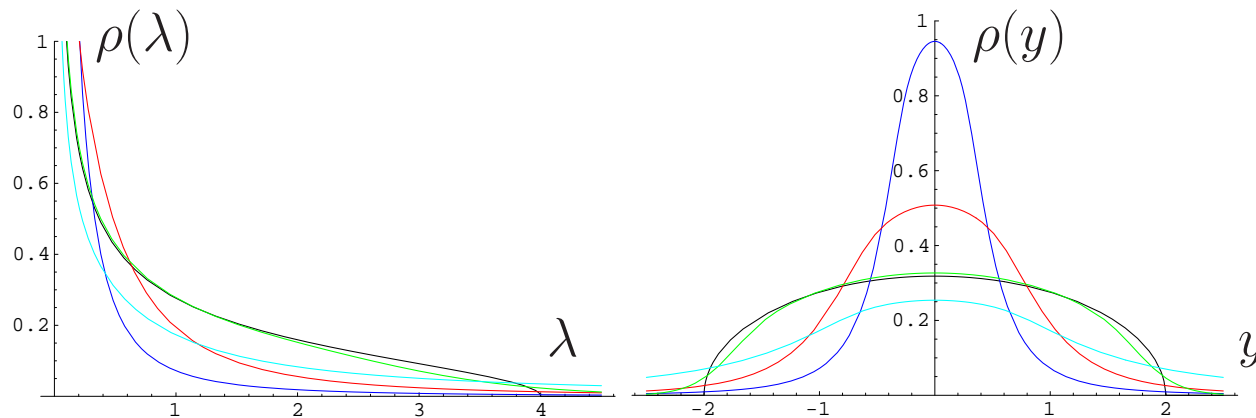
- **fold** know formulae for **Gauß**

$$P[\mathbf{X}\mathbf{X}^\dagger] \sim \int d\xi e^{-\xi} \xi^{\gamma-1} \exp \left[-\xi \frac{1}{\gamma} \text{Tr} \mathbf{X}\mathbf{X}^\dagger \right]$$

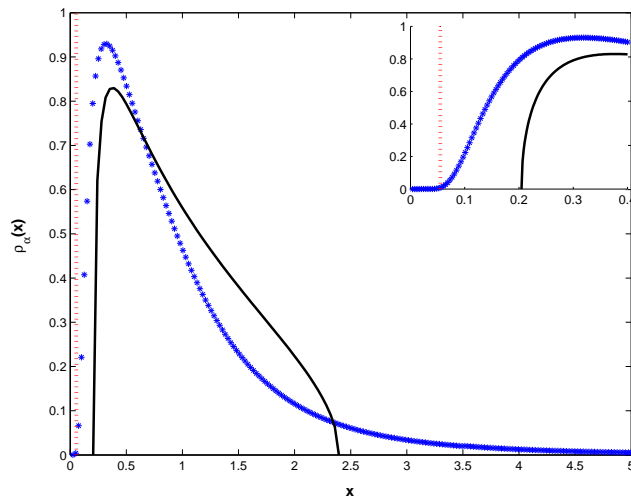
- Result: **one parameter deformation** (fixed $c = \frac{N}{T}$)

$$\alpha \sim \gamma - NT \quad \boxed{\rho(\lambda) \sim \lambda^{-\alpha}, \quad 1 < \alpha \neq 2}$$

- example \int semi-circle:

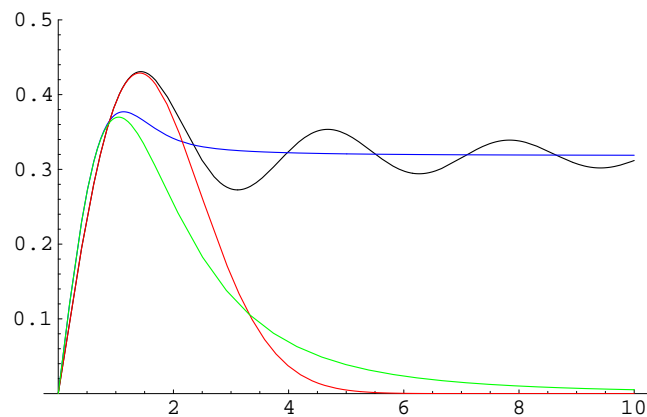


- example \int Marčenko-Pastur:



$$\frac{N}{T} = c = 0.3, \quad \alpha = 5$$

- example \int origin=Bessel: **universal**



$$N = T, \quad \alpha = 2.1$$

– even smallest eigenvalue has a tail!

Open Problems

- in ordinary RMT: largest ev λ_{max} has many applications, what happens with power-law?
- distribution of smallest eigenvalue for $c < 1$?
- (some) universality:
similar curves from multivariate student distribution
- comparison to other data, e.g. from complex networks
- how to subtract noise = RMT from data?