

# Disorder-generated multifractals and Random Matrices: freezing phenomena and extremes<sup>1</sup>

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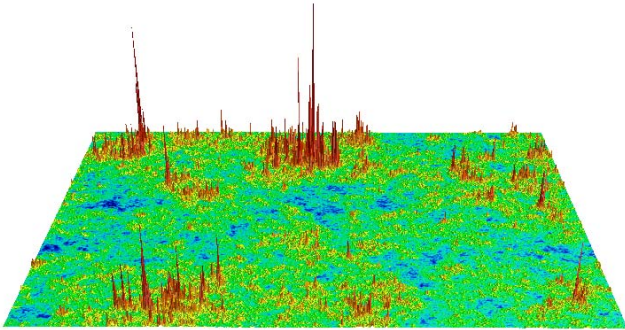
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<sup>1</sup>Based on: **YVF**, **P Le Doussal** and **A Rosso** J Stat Phys: **149** (2012), 898-920  
**YVF**, **G Hiary**, **J Keating** Phys. Rev. Lett. **108**, 170601 (2012) & [arXiv:1211.6063](https://arxiv.org/abs/1211.6063)  
**YVF**, **B. A. Khoruzhenko** and **N. Simm** in preparation

## Disorder-generated multifractals:

Intensity patterns in systems with disorder frequently display high variability over a wide range of space or time scales, associated with huge fluctuations which can be visually detected.



**Intensity of a multifractal wavefunction at the point of Integer Quantum Hall Effect.**

Courtesy of F. Evers, A. Mirlin and A. Mildenberger.

**Multifractality** characterizes such patterns in a lattice of  $M = (L/a)^d \gg 1$  sites by attributing different scaling of intensities  $h_i \sim M^{x_i}$  at different lattice sites  $i = 1, \dots, M$ , with exponents  $x_i$  forming a dense set such that

$$\rho_M(x) = \sum_{i=1}^M \delta \left( \frac{\ln h_i}{\ln M} - x \right) \approx c_M(x) \sqrt{\ln M} M^{f(x)}, \quad M \gg 1,$$

which is frequently referred to as the **multifractal Ansatz**. Whereas the **singularity spectrum**  $f(x)$  is typically self-averaging, there are essential sample-to-sample fluctuations of the **prefactor**  $c_M(x)$  in different realizations of disorder, as well as fluctuations in the number and height of **extreme peaks** of the pattern. Those fluctuations will be the subject of our interest.

## From disorder-generated multifractals to log-correlated fields:

Disorder-generated multifractal patterns of intensities  $h(\mathbf{r})$  are typically **self-similar**

$$\mathbb{E} \{h^q(\mathbf{r}_1)h^s(\mathbf{r}_2)\} \propto \left(\frac{L}{a}\right)^{y_{q,s}} \left(\frac{|\mathbf{r}_1 - \mathbf{r}_2|}{a}\right)^{-z_{q,s}}, \quad q, s \geq 0, \quad a \ll |\mathbf{r}_1 - \mathbf{r}_2| \ll L$$

and **spatially homogeneous**

$$\mathbb{E} \{h^q(\mathbf{r}_1)\} \sim \frac{1}{M} \sum_{\mathbf{r}} h^q(\mathbf{r}) \propto \left(\frac{L}{a}\right)^{d(\zeta_q - 1)}$$

where  $\zeta_q$  and  $f(x)$  are related by the **Legendre transform**:

$$f'(y_*) = -q \text{ and } \zeta_q = f(y_*) + q y_*.$$

**The consistency of the two conditions for  $|\mathbf{r}_1 - \mathbf{r}_2| \sim a$  and  $|\mathbf{r}_1 - \mathbf{r}_2| \sim L$  implies:**

$$y_{q,s} = d(\zeta_{q+s} - 1), \quad z_{q,s} = d(\zeta_{q+s} - \zeta_q - \zeta_s + 1)$$

**If we now introduce the field  $V(\mathbf{r}) = \ln h(\mathbf{r}) - \mathbb{E} \{\ln h(\mathbf{r})\}$  and exploit the identities**

**$\frac{d}{ds} h^s|_{s=0} = \ln h$  and  $\zeta_0 = 1$  we arrive at the relation:**

$$\mathbb{E} \{V(\mathbf{r}_1)V(\mathbf{r}_2)\} = -g^2 \ln \frac{|\mathbf{r}_1 - \mathbf{r}_2|}{L}, \quad g^2 = d \frac{\partial^2}{\partial s \partial q} \zeta_{q+s} |_{s=q=0}$$

**Conclusion:** logarithm of a multifractal intensity is a **log**-correlated random field.

To understand statistics of **high values** and **extremes** of general logarithmically correlated random fields we consider the simplest 1D case of such process: the **Gaussian**  $1/f$  noise.

## Ideal Gaussian periodic 1/f noise:

We will mainly study an idealized periodic model for Gaussian **1/f** noise defined as

$$V(t) = \sum_{n=1}^{\infty} \frac{1}{\sqrt{n}} [v_n e^{int} + \bar{v}_n e^{-int}] , \quad t \in [0, 2\pi)$$

where  $v_n, \bar{v}_n$  are **complex standard Gaussian i.i.d.** with  $\mathbb{E}\{v_n \bar{v}_n\} = 1$ . It implies the formal covariance structure:

$$\mathbb{E}\{V(t_1)V(t_2)\} = -2 \ln |2 \sin \frac{t_1 - t_2}{2}|, \quad t_1 \neq t_2$$

The corresponding process is a **random distribution** and in applications should be regularized. There are several alternative regularizations. E.g. one can replace the function  $V(t)$ ,  $t \in [0, 2\pi)$  with a sequence of  $M \gg 1$  random mean-zero Gaussian variables  $V_k \equiv V(t = \frac{2\pi}{M}k)$  with the covariance matrix  $\mathbb{E}\{V_k V_m\}$  given by

$$\mathbb{E}\{V_k V_m\} = -2 \ln |2 \sin \frac{\pi}{M}(k - m)|, \quad C_{kk} = \mathbb{E}\{V_k^2\} > 2 \ln M, \quad \forall k = 1, \dots, M$$

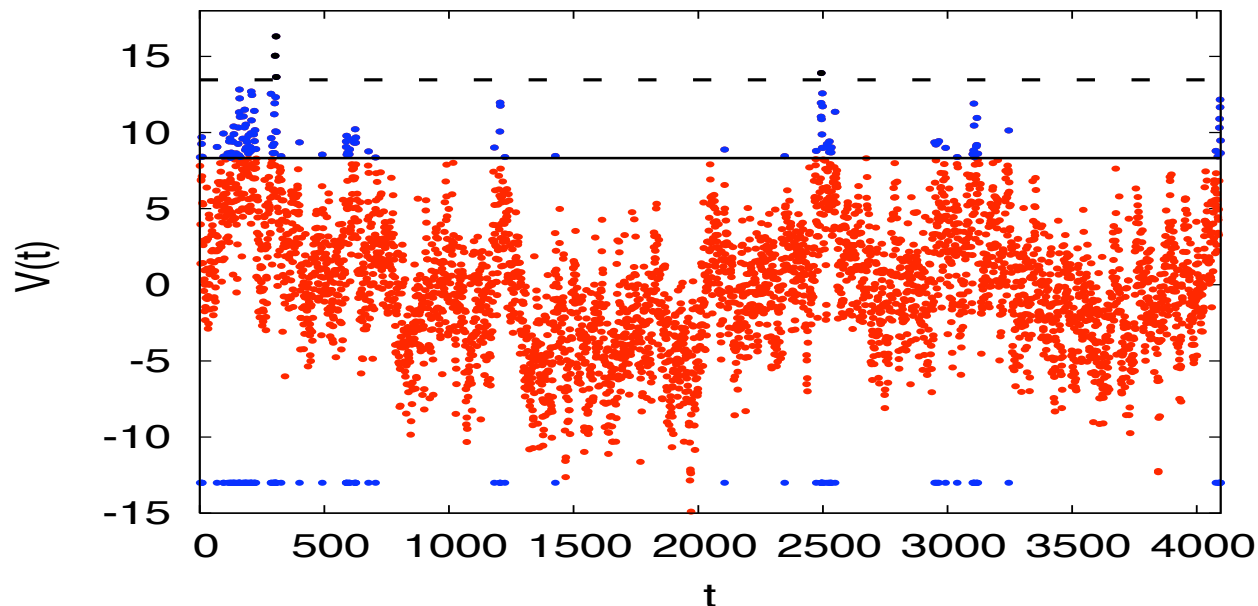
One also may give *bona fide* mathematical definition as e.g. 1D "projection" of the 2D **Gaussian Free Field**. We shall see that one may further consider aperiodic **stationary** logarithmically-correlated processes as well as similar processes with **stationary increments**:

$$\mathbb{E}\{V(t_1)V(t_2)\} \propto -\log |t_1 - t_2| \quad \text{or} \quad \mathbb{E}\{[V(t_1) - V(t_2)]^2\} \propto \log |t_1 - t_2|$$

The **multifractal intensity pattern** is then generated by setting  $h_i = e^{V_i}$  for each  $i = 1, \dots, M$ .

## Ideal Gaussian periodic 1/f noise:

An example of the  $1/f$  signal sequence generated for  $M = 4096$  for the discretized version of periodic  $1/f$  noise is given in the figure.



The upper line marks the typical value of the **extreme value threshold**  $V_m = 2 \ln M - \frac{3}{2} \ln \ln M$ .

The lower line is the level  $\frac{1}{\sqrt{2}} V_m$  and blue dots mark points supporting  $V_i > \frac{1}{\sqrt{2}} V_m$ .

**Questions we would like to answer:** How many points are typically above a given level of the noise? How strongly does this number fluctuate for  $M \gg 1$  from one realization to the other? How to understand the typical position  $V_m$  and statistics of the **extreme values** (maxima or minima), etc. And, after all, what parts of the answers are **universal** and what is the universality class?

## Characteristic polynomial of random CUE matrix and periodic 1/f noise:

Let  $U_N$  be a  $N \times N$  **unitary matrix**, chosen at random from the unitary group  $\mathcal{U}(N)$ . Introduce its **characteristic polynomial**  $p_N(\theta) = \det(1 - U_N e^{-i\theta})$  and further consider  $V_N(\theta) = -2 \log |p_N(\theta)|$ . Following **Hughes, Keating & O'Connell** 2001 one can employ the following representation

$$V_N^{(U)}(\theta) = -2 \log |p_N(\theta)| = \sum_{n=1}^{\infty} \frac{1}{\sqrt{n}} \left[ e^{-in\theta} v_n^{(N)} + \text{comp. conj.} \right]$$

where  $v_n^{(N)} = \frac{1}{\sqrt{n}} \text{Tr}(U_N^{-n})$ .

According to **Diaconis & Shahshahani** 1994 the coefficients  $v_{n_1}^{(N)}, \dots, v_{n_k}^{(N)}$  for any fixed finite set  $n_1, \dots, n_k$  tend in the limit  $N \rightarrow \infty$  to i.i.d. complex gaussian variables with zero mean and variance  $\mathbb{E}\{|\zeta_n|^2\} = 1$ . This suggests that for finite  $N$  **Log-Mod** of the characteristic polynomial of CUE matrices is just a **certain regularization** of the stationary random **Gaussian Fourier series** of the form

$$V(t) = \sum_{n=1}^{\infty} \frac{1}{\sqrt{n}} \left[ v_n e^{int} + \bar{v}_n e^{-int} \right], \quad t \in [0, 2\pi)$$

where  $v_n, \bar{v}_n$  are **complex standard Gaussian i.i.d.** with  $\mathbb{E}\{v_n \bar{v}_n\} = 1$ . It implies

$$\mathbb{E}\{V(t_1)V(t_2)\} = 2 \sum_{n=1}^{\infty} \frac{1}{n} \cos n(t_1 - t_2) \equiv -2 \ln \left| 2 \sin \frac{t_1 - t_2}{2} \right|, \quad t_1 \neq t_2.$$

## Characteristic polynomial of random GUE matrix and aperiodic 1/f noise:

Let  $\mathbf{H}$  be a  $N \times N$  **Hermitian** GUE random matrix:  $\mathcal{P}(\mathbf{H}) \propto \exp \left\{ -\frac{N}{2} \text{Tr} \mathbf{H}^2 \right\}$  such that the mean eigenvalue density  $\rho_N(u) = \mathbb{E} \left\{ \frac{1}{N} \text{Tr} \delta(u - \mathbf{H}) \right\}$  tends to the Wigner semicircular law in  $x \in (-2, 2)$ . Introduce its **characteristic polynomial**  $\pi_N(x) = \det(x - \mathbf{H})$ . Let  $T_n(u) = \cos(n \arccos u)$  be **Tschebyshev** polynomials orthogonal in  $u \in (-1, 1)$  with the weight  $1/\sqrt{1-u^2}$ . It turns out that the following representation is valid:

$$f_N(x) = \mathbb{E} \left\{ \log |\pi_N(x)| \right\} - \log |\pi_N(x)| = \sum_{n=1}^{\infty} \frac{2}{\sqrt{n}} \left[ w_n^{(N)} + R_n^{(N)} \right] T_n \left( \frac{x}{2} \right)$$

where  $w_n^{(N)} = \sqrt{\frac{2}{n}} \left[ \text{Tr} \left\{ T_n \left( \frac{\mathbf{H}}{2} \right) \right\} - N \int_{-\infty}^{\infty} T_n \left( \frac{u}{2} \right) \rho_N(u) du \right]$ .

According to **Johansson** 1998 in the limit  $N \rightarrow \infty$  the coefficients  $w_{n_1}^{(N)}, \dots, w_{n_k}^{(N)}$  tend to a set of i.i.d. real gaussian variables with zero mean and unit variance. We also can show that  $R_n^{(N)} \rightarrow 0$  in probability. This suggests that for finite  $N$  **Log-Mod** of the characteristic polynomial of GUE matrices is just a **certain regularization** of a random **Gaussian series** of the form

$$f(x) = \sum_{n=1}^{\infty} \sqrt{\frac{2}{n}} w_n T_n \left( \frac{x}{2} \right), \quad x \in (-2, 2)$$

where  $w_n$  are **real standard** Gaussian i.i.d variables. It implies

$$\mathbb{E} \left\{ f(x_1) f(x_2) \right\} = \sum_{n=1}^{\infty} \frac{2}{n} T_n \left( \frac{x_1}{2} \right) T_n \left( \frac{x_2}{2} \right) \equiv -\ln |x_1 - x_2|, \quad x_1 \neq x_2.$$

## **Mesoscopic regime of GUE and fractional Brownian motion with $H = 0$ :**

Define parameter  $d_N$  such that  $1 \ll d_N \ll N$  for  $N \gg 1$ , and consider for  $\eta > 0$

$$W_{N,x}^{(\eta)}(t) = -\log \left| \det \left( \mathcal{H} - x - \frac{t+i\eta}{d_N} \right) \right| + \log \left| \det \left( \mathcal{H} - x - \frac{i\eta}{d_N} \right) \right|,$$

One can show that

$$W_{N,x}^{(\eta)}(t) = \frac{1}{\sqrt{\pi}} \int_0^\infty \frac{d\omega}{\sqrt{\omega}} e^{-\eta\omega} \left[ v_{N,x}(\omega) \frac{e^{i\omega t} - 1}{2} + \bar{v}_{N,x}(\omega) \frac{e^{-i\omega t} - 1}{2} \right]$$

where  $v_{N,x}(\omega) = \frac{1}{\sqrt{\omega}} e^{ixd_N \omega} \text{Tr} [e^{-i\omega d_N \mathbf{H}}]$ .

It can be further verified that for  $N \rightarrow \infty$  and  $x \in (-2, 2)$  the Fourier coefficients  $v_{N,x}(\omega)$  tend to the standard complex **Gaussian "white noise"**:

$$\mathbb{E}\{v_{N,x}(\omega)\} \rightarrow 0, \quad \mathbb{E}\{v_{N,x}(\omega_1) \overline{v_{N,x}(\omega_2)}\} \rightarrow \delta(\omega_1 - \omega_2)$$

which implies that  $W_{N,x}^{(\eta)}(t) \rightarrow B_0^{(\eta)}(t)$  such that

$$\mathbb{E} \left\{ \left[ B_0^{(\eta)}(t_1) - B_0^{(\eta)}(t_2) \right]^2 \right\} = \frac{1}{\pi} \ln \frac{(t_1 - t_2)^2 + 4\eta^2}{4\eta^2}.$$

This is a **properly regularized** limit  $H \rightarrow 0$  of the **Fractional Brownian Motion**  $B_H(t)$  such that  $\mathbb{E}\{B_H(t_1)B_H(t_2)\} = \frac{\sigma_H^2}{2} (|t_1|^{2H} + |t_2|^{2H} - |t_1 - t_2|^{2H})$ .



## 1/f Noises, Disordered Energy Landscapes, and Burgers Turbulence:

In the area of **Statistical Mechanics** of **Disordered Systems** logarithmically correlated fields and **1/f** noises have been recently identified as **potential energy landscapes** underlying an intriguing phenomenon of the **freezing transition** which takes place at some finite temperature  $T = T_c$  ( **Carpentier & Le Doussal** 2001; **YVF & Bouchaud** 2008; **YVF, Le Doussal & Rosso** 2009). In a related development, it was shown that a freezing transition shows up also in the problem of **decaying Burgers turbulence**, i.e. analysis of the Burgers equation  $\partial_t v + (v \nabla) v = \nu \nabla^2 v$ ,  $\nu > 0$  with random initial conditions given by the gradient of the **1/f** noise (**YVF, Le Doussal & Rosso** 2010 & unpublished).

Though most of the above considerations are not yet rigorous, as the result we by now have a qualitative, and sometimes, quite precise quantitative understanding of statistics of **high** and **extreme** values of such random processes: the statistics of the **number of points** in a pattern above a **given level**, and the distribution of the **highest intensity**  $V_m$  in the pattern. In particular, for the periodic Gaussian **1/f** noise we have a prediction  $V_m = 2 \ln M - \frac{3}{2} \ln \ln M - x$ , where  $x$  is a random variable with the density  $p(x) = 2e^x K_0(2e^{x/2})$ . This is manifestly different from the ubiquitous double-exponential **Gumbel distribution**  $p_{Gumb}(x) = -\frac{d}{dx} \exp\{-e^x\}$  universally valid for maxima of **short-range correlated** random variables.

## Statistics of the counting function $\mathcal{N}_M(x)$ and threshold of extreme values:

By relating moments of the counting function  $\mathcal{N}_M(x) = \int_x^\infty \rho_M(y) dy$  for log-correlated **1/f noise** to **Selberg integrals** we conjecture that the probability density for the (scaled) counting function  $n = \mathcal{N}_M(x)/\mathcal{N}_t(x)$  is given by:

$$\mathcal{P}_x(n) = \frac{4}{x^2} e^{-n^{-\frac{4}{x^2}}} n^{-\left(1+\frac{4}{x^2}\right)}, \quad n \ll n_c(x), \quad 0 < x < 2.$$

with  $n_c \rightarrow \infty$  for  $M \rightarrow \infty$  and the **characteristic scale**  $\mathcal{N}_t(x)$  given by

$$\mathcal{N}_t(x) = \frac{M^{f(x)}}{x\sqrt{\pi \ln M}} \frac{1}{\Gamma(1-x^2/4)} = \mathbb{E} \{ \mathcal{N}_M(x) \} \frac{1}{\Gamma(1-x^2/4)}, \quad f(x) = 1 - x^2/4$$

**Note:** For  $x \rightarrow 2$  the **typical** value of the counting function  $e^{\mathbb{E}\{\ln \mathcal{N}_M(x)\}} \sim \mathcal{N}_t(x)$  and hence is **parametrically smaller** than the mean value  $\mathbb{E} \{ \mathcal{N}_M(x) \}$ . In particular, the position  $x_m$  of the typical threshold of **extreme values** determined from the condition  $\mathcal{N}_t(x) \sim 1$  is given by

$$x_m = 2 - c \frac{\ln \ln M}{\ln M} + O(1/\ln M) \text{ with } c = 3/2.$$

In contrast, for **short-ranged** random sequences **mean=typical**. Had we instead decided to use the condition  $\mathbb{E} \{ \mathcal{N}_M(x) \} \sim 1$  that would give the above but with  $c = 3/2 \rightarrow c = 1/2$ . The value  $c = \frac{3}{2}$  is conjectured to be a universal feature of systems with **logarithmic** correlations (cf. **Bramson & Zeitouni**).

## Threshold of extreme values for self-similar multifractal fields:

Apart from **1/f noise** and its incarnations (like modulus of characteristic polynomials of random matrices) the new universality class is believed to include: the  $2D$  **Gaussian free field**, branching random walks & polymers on disordered trees, modulus of zeta-function along the critical line, fluctuations of shapes of random Young tableaux sampled with Plancherel measure, some models in turbulence and financial mathematics and, with due modifications the **disorder-generated multifractals**.

Namely, consider a multifractal random **probability measure**  $p_i \sim M^{-\alpha_i}$ ,  $i = 1, \dots, M$  such that  $\sum_{i=1}^M p_i = 1$  characterized by a general non-parabolic **singularity spectrum**  $f(\alpha)$  with the left endpoint at  $\alpha = \alpha_- > 0$ . Then very similar consideration based on insights from **Mirlin & Evers** 2000 suggests that the **extreme value threshold** should be given by  $p_m = M^{-\alpha_m}$ , where  $\alpha_m$

$$\alpha_m \approx \alpha_- + \frac{3}{2} \frac{1}{f'(\alpha_-)} \frac{\ln \ln M}{\ln M} \quad \Rightarrow \quad -\ln p_m \approx \alpha_- \ln M + \frac{3}{2} \frac{1}{f'(\alpha_-)} \ln \ln M$$

For branching random walks this is indeed rigorously proved: **L. Addario-Berry & B. Reed** 2009; **E. Aidekon** 2012

Work in progress: testing such a prediction for multifractal eigenvectors of a certain  $N \times N$  random matrix ensemble introduced by E. Bogomolny & O. Giraud, *Phys. Rev. Lett.* **106** 044101 (2011). Preliminary numerics is supportive of the theory.

## Distribution of the absolute maximum: partition function approach:

Given the sequence  $\{V_i, i = 1, \dots, M\}$  we are interested in finding the **distribution** of  $V_{(m)} = \max(V_1, \dots, V_M)$  that is

$$P(v) = \text{Prob}(V_{(m)} < v) = \text{Prob}(V_i < v, \forall i) = \mathbb{E} \left\{ \prod_{i=1}^M \theta(v - V_i) \right\}$$

$$\text{Next we use: } \lim_{q \rightarrow \infty} \exp \left[ -e^{-q(v-V_i)} \right] = \begin{cases} 1 & v > V_i \\ 0 & v < V_i \end{cases} \equiv \theta(v - V_i)$$

which immediately shows that:

$$P(v) = \text{Prob}(V_{(m)} < v) = \lim_{q \rightarrow \infty} \mathbb{E} \left\{ \exp \left[ -e^{-qv} Z_q \right] \right\}, \text{ where } Z_q = \sum_{i=1}^M e^{qV_i}$$

**In the limit  $\ln M \gg 1$  moments of  $Z_q$  for  $|q| < 1$  can be again related to **Selberg** integrals, which allows to extract the function  $G_q(v) = \mathbb{E} \left\{ \exp \left[ -e^{-qv} Z_q \right] \right\}$  for  $q < 1$ :**

$$G_q(v) = g_q(v - c_q \ln M) \text{ where } c_q = \left( q + \frac{1}{q} \right) \text{ and } g_q(v) = \int_0^\infty dt \exp \left\{ -t - e^{-qv} t^{-q^2} \right\}$$

**One may further notice that not only  $c_q = c_{q^{-1}}$  but the whole function satisfies a quite remarkable **duality relation****

$$g_q(v) = 1 + \sum_{n=1}^{\infty} \frac{(-1)^n}{n!} \left[ e^{-nqv} \Gamma(1 - nq^2) + e^{-n\frac{v}{q}} \Gamma\left(1 - \frac{n}{q^2}\right) \right] = g_{\frac{1}{q}}(x)$$

**THIS HOWEVER STILL DOES NOT ALLOW TO CONTINUE TO  $q > 1$ !**

## Freezing conjecture and the distribution of extremes:

Using certain analogy with the **Derrida-Spohn** model of polymers on disordered trees we conjecture the following **freezing scenario**: for the  $\log$ -circular model the same sort of **freezing transition** takes place at  $q = 1$ . Namely, the function

$$g_{q<1}(v) = \int_0^\infty dt \exp \left\{ -t - e^{-qv} t^{-q^2} \right\}$$

**freezes** to the **q-independent** profile  $g_{q=1}(v) = 2e^{-v/2} K_1(2e^{-v/2})$  in the whole "glassy" phase  $q > 1$ .

### **Consequences:**

(i) The latter profile then is precisely the distribution  $P(v)$  of the (shifted) absolute maximum:  $V_m = 2 \ln M - \frac{3}{2} \ln \ln M + v$ . This distribution is manifestly **non-Gumbel**, and shows the tail behaviour:  $P(v \rightarrow -\infty) \approx 1 - |v|e^v$

(ii) The probability density of the partition function  $Z_q$  in the whole regime  $q > 1$  must display a power-law forward tail of the form:

$$\mathcal{P}_{q>1}(Z) \propto Z^{-(1+\frac{1}{q})} \ln Z$$

We conjecture that such a tail, including the meaningful **log-factor**, is to be **universal** for the whole class of logarithmically correlated processes.

## Numerics for the maxima of CUE characteristic polynomials:

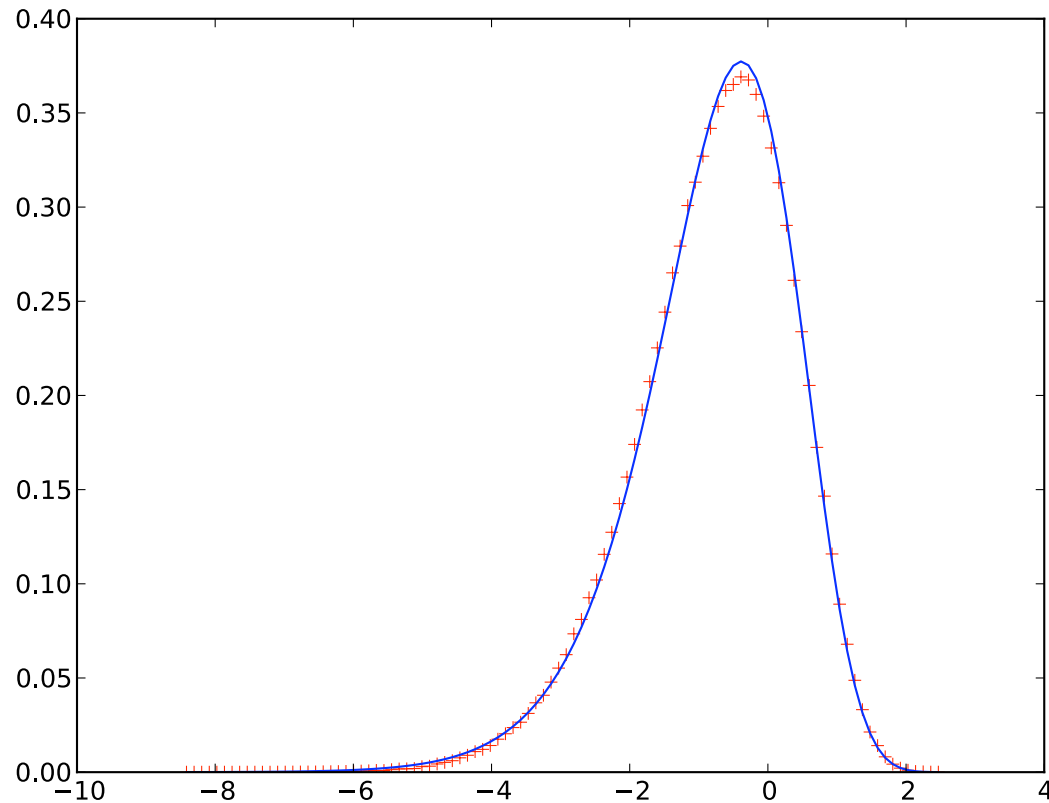


Figure 1: Density of maxima for CUE polynomials (  $N = 50, 10^6$  samples ) compared to periodic  $1/f$  noise prediction  $p(v) = 2e^v K_0(2e^{v/2})$ .

## From $1/f$ noise to Riemann $\zeta(1/2 + it)$ :

As is well-known, zeroes of the Riemann zeta-function  $\zeta(s)$  along the "critical line"  $s = 1/2 + it$ ,  $t \in \mathbb{R}$  behave statistically as a sequence of eigenvalues of random Hermitian GUE matrices (**Montgomery** 1983). Following the ideas of **Keating & Snaith** 2000 one can expect that **log-mod** of the Riemann zeta-function  $\zeta(1/2 + it)$  **locally** resembles **log-mod** of CUE characteristic polynomial, and hence a (non-periodic) version of the **1/f noise**, see also **P. Bourgade** 2010. One can exploit this fact to predict statistics of **moments** and **high values** of the Riemann zeta along the critical line using the previously exposed ideas (**YVF, Keating** 2012).

## Our approach to statistics of $\zeta(1/2 + it)$ :

We expect a **single** unitary matrix of size  $N_T = \log(T/2\pi) \gg 1$  to model the Riemann zeta  $\zeta(1/2 + it)$ , statistically, over a range of  $T \leq t \leq T + 2\pi$ . We thus suggest splitting the **critical line** into ranges of **length  $2\pi$** , and making the statistics of  $\zeta(1/2 + it)$  over the many ranges.

There are roughly  $N_T$  zeros in each range of length  $2\pi$ . At heights  $T \sim 10^{22} - 10^{28}$  we used samples that spans  $\approx 10^7$  zeros.

## Our predictions for $\zeta(1/2 + it)$ and CUE characteristic polynomials:

We expect

$$\log \zeta_{max}(T) \approx \log N_T - \frac{3}{4} \log \log N_T - \frac{1}{2} x, \quad N_T = \log(T/2\pi)$$

where  $x$  is distributed with a probability density behaving in the tail as  $\rho(x \rightarrow -\infty) \approx |x| e^x$ .

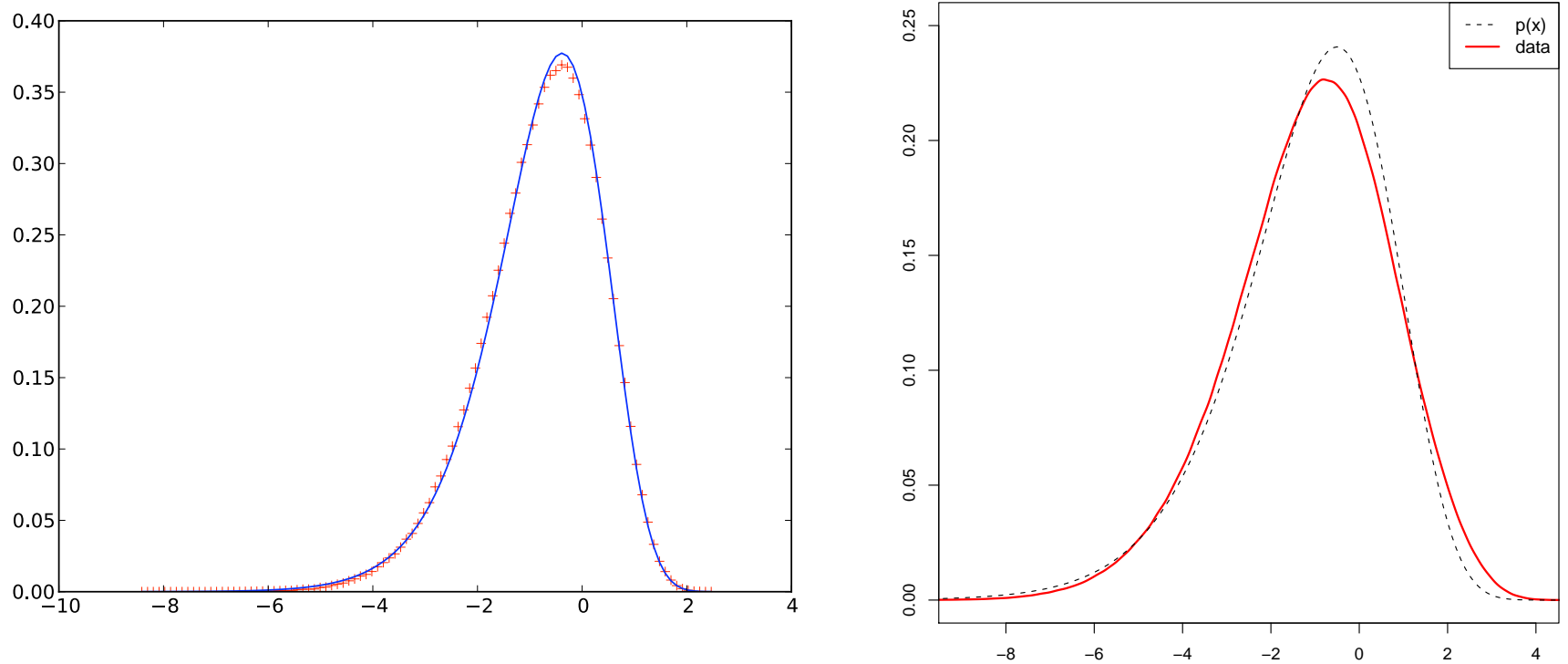


Figure 2: Statistics of maxima for CUE polynomials (left:  $N = 50$ ,  $10^6$  samples ) and  $|\zeta(1/2 + it)|$  (right:  $N_T = 65$ ,  $10^5$  samples ) compared to periodic  $1/f$  noise prediction  $p(x) = 2e^x K_0(2e^{x/2})$ .



## Summary:

- I. Disorder-generated multifractal patterns are intimately connected to log-correlated random fields.
- II. log-mod of characteristic polynomials of random matrices on the *global* scale are examples of log-correlated Gaussian **1/f noises**. The same objects on *mesoscopic* scale are examples of Fractional Brownian Motion with  $H = 0$ .
- III. Exploiting the methods of statistical mechanics of disordered systems we attempted to understand the statistics of **minima/maxima** of the CUE characteristic polynomial over various intervals, as well as the related **moments** and **high values**.
- IV. The above picture can be translated into making non-trivial conjectures about statistics of **moments** and **high values** of (i) the Riemann zeta along the critical line (ii) a more general class of disorder-generated multifractal fields.