# How many stable equilibria will a large complex system have?

Boris Khoruzhenko

Queen Mary University of London

after joint work with Yan Fyodorov (PNAS 2016) and unpublished work with Gerard Ben Arous and Yan Fyodorov

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$$\dot{\mathbf{y}} = -\mu \mathbf{y} + J\mathbf{y}, \qquad J \in \mathbf{R}^{N \times N}.$$

Here  $\mu > 0$  sets the relaxation time scale and  $J_{jk}$  characterises interaction. In an ecological context,  $y_j(t)$  is the variation about equilibrium value in popn dens of species j, and  $J_{jk}$  measures per capita effect of species k on species j, hence **community matrix** J **is asymmetric** (think of sea lions and sardines).

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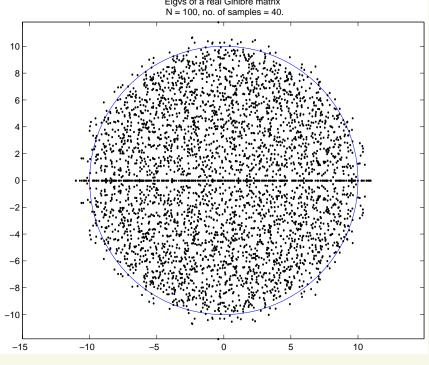
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# **May-Wigner instability transition**

Have Girko's circular law: As  $N \to \infty$ , EV distribution of  $J/\sqrt{N}$  converges to unif distrib on the unit disk, Girko1984, Bai1997, Götze &.Tikhomirov, Tau & Vu 2010.

Also, the rescaled spectral radius of J is  $\leq 1$  in the limit of large N, Geman 1984.

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the linearised system is stable if  $\frac{\mu}{\alpha\sqrt{N}}>1$  and unstable if  $\frac{\mu}{\alpha\sqrt{N}}<1$ .

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Is there a signature of May-Wigner instability transition on the global scale?

#### A simple model for generic large complex systems: consider

$$\frac{dx_i}{dt} = -\mu x_i + f_i(x_1, \dots, x_N), \quad i = 1, \dots, N.$$

 $\mu > 0$  as before, and now  $\mathbf{f}(\mathbf{x})$  is a smooth random field with N components.

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This system may have multiple equilibria depending on the realisation of f(x).

Near equilibrium  $\mathbf{x}_e$  it 'reduces' to May's model with  $\mathbf{y} = \mathbf{x} - \mathbf{x}_e$ ,  $J_{jk} = \frac{\partial f_j}{\partial x_k}(\mathbf{x}_e)$ .

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Gradient-descent flow,  $\mathbf{f} = -\nabla V$ , is special (but typical) case. Have

$$\frac{d\mathbf{x}}{dt} = -\nabla L, \quad L(\mathbf{x}) = \frac{\mu |\mathbf{x}|^2}{2} + V(\mathbf{x})$$
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Helpful for building geometric intuition:  $\mathbf{x}(t)$  moves in the direction of the steepest descent, perpendicular to level lines  $L(\mathbf{x}) = h$  towards ever smaller values of h.

The term  $\mu |\mathbf{x}|^2/2$  represents the globally confining parabolic potential, a deep well on the surface of  $L(\mathbf{x})$ . The random potential  $V(\mathbf{x})$  generates many local minima of  $L(\mathbf{x})$  (shallow wells). Have two competing terms...

Consider **non-linear** systems  $\dot{\mathbf{x}} = -\mu \mathbf{x} + \mathbf{f}(\mathbf{x})$ ,  $x \in \mathbf{R}^N$  with

$$f_i(\mathbf{x}) = -\frac{\partial V}{\partial x_i} + \frac{1}{\sqrt{N}} \sum_{j=1}^N \frac{\partial A_{ij}}{\partial x_j}, \quad A_{ij}(\mathbf{x}) = -A_{ji}(\mathbf{x}) \ \forall i, j.$$

This is a fairly general class. If N=3 can find V and  $\mathbf{A}$  s.t.  $\mathbf{f} = \nabla V + \nabla \times \mathbf{A}$  (Helmholtz decomposition). In these terms, our  $\mathbf{f}(\mathbf{x})$  is the sum of 'gradient' (irrotational) and 'solenoidal' parts.

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$$\langle V(\mathbf{x})V(\mathbf{y})\rangle = \mathbf{v^2} F_V(|\mathbf{x} - \mathbf{y}|^2), \langle A_{ij}(\mathbf{x})A_{nm}(\mathbf{y})\rangle = \mathbf{a^2} F_A(|\mathbf{x} - \mathbf{y}|^2)(\delta_{in}\delta_{jm} - \delta_{im}\delta_{jn})$$

with suitable  $F_V$  and  $F_A$ :

$$F_{V,A}(s) = \int_0^\infty e^{-st} d\sigma_{V,A}(t) \quad \forall s \ge 0,$$
 (Schoenberg,1938)

where  $d\sigma_V$  and  $d\sigma_A$  have finite mass. Also assume finite 3rd moments, and normalise  $d^2F_{V,A}(s)/ds^2\big|_{s=0}=1$ .

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Let  $\tau = \frac{v^2}{v^2 + a^2}$ . This is a measure of relative strength of the two components: if  $\tau = 1$  then  $\mathbf{f}(\mathbf{x})$  is purelly gradient, and if  $\tau = 0$  then  $\mathbf{f}(\mathbf{x})$  is purelly solenoidal.

Let  $\mathcal{N}_{tot}$  be the total number of equilibrium pnts of  $\dot{\mathbf{x}} = -\mu \mathbf{x} + \mathbf{f}(\mathbf{x})$ .

These are solutions of the system of N equations  $-\mu \mathbf{x} + \mathbf{f}(\mathbf{x}) = \mathbf{0}$ .

Introduce dimensionless  $m=\frac{\mu}{2\alpha\sqrt{N}}$ , where  $\alpha=\sqrt{v^2+a^2}$  interaction strength.

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**Theorem.** [YF and BK 2016] Assume  $0 \le \tau < 1$ . To leading order for N large,

$$\langle \mathcal{N}_{tot} \rangle = \begin{cases} 1 & \text{if } m > 1\\ \sqrt{\frac{2(1+\tau)}{1-\tau}} e^{N \Sigma_{tot}(m)} & \text{if } 0 < m < 1 \end{cases}$$

where  $\Sigma_{tot}(m) = \frac{m^2-1}{2} - \ln m$ . Moreover, the relative width of the crossover region is  $N^{-1/2}$  and the crossover profile of  $\langle \mathcal{N}_{tot} \rangle$  can be found in closed form.

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Thus, as the complexity exceeds a critical level, large complex non-linear systems exhibit a sharp transition from having on average one point of equilibrium to having exponentially many equilibria.

Want to count zeros of  $-\mu x + f(x)$ . By Kac-Rice,

$$\mathcal{N}_{tot} = \int_{\mathbb{R}^N} \delta\left(-\mu \mathbf{x} + \mathbf{f}(\mathbf{x})\right) \left| \det\left(-\mu \delta_{ij} + \frac{\partial f_i}{\partial x_j}(\mathbf{x})\right) \right| d\mathbf{x}.$$

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Matrix J is Gaussian, with zero mean and matrix entry correlators

$$\langle J_{ij}J_{nm}\rangle = \alpha^2(\delta_{in}\delta_{jm} + \tau\delta_{jn}\delta_{im} + \tau\delta_{ij}\delta_{nm}) + O(1/N).$$

Thus,  $J = \alpha(X + \sqrt{\tau}\xi I)$  where  $X \sim \text{RealGin}(\tau)$  and  $\xi \sim N(0,1)$ , independent.

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where 
$$x=(m+t\sqrt{\tau})\sqrt{N}$$
 and  $P(X)\propto \exp\left[-\frac{1}{2(1-\tau^2)}\left(\operatorname{Tr} XX^T-\tau\operatorname{Tr} X^2\right)\right]$ .

Analytic problem: find the average of the abs value of the characteristic polynomial in the real elliptic Ginibre ensemble.

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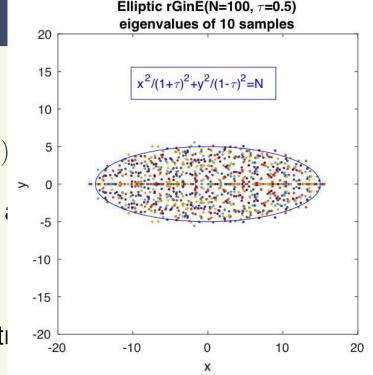
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#### **Edelman-Kostlan-Shub trick**

Start with the real elliptic ensemble  $X_{N+1}$  of  $(N+1)\times (N+1)$  matrices.

Decompose  $X_{N+1} = Q \begin{pmatrix} x & \mathbf{w} \\ 0 & X_N \end{pmatrix} Q^T$ , where x is a real eigenvalue of  $X_{N+1}$ , and Q is an orthogonal matrix that exchanges the corresponding eigenvector and  $(1,0,\ldots,0)$  (Householder reflection).

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The Jacobian of changing from  $X_{N+1}$  to  $X_N, Q, x, \mathbf{w}$  is  $|\det(xI_N - X_N)|$ .

Note:  $\operatorname{Tr} X_{N+1} X_{N+1}^T = x^2 + \mathbf{w}^T \mathbf{w} + \operatorname{Tr} X_N X_N^T$  and  $\operatorname{Tr} X_{N+1}^2 = x^2 + \operatorname{Tr} X_N^2$ .

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Therefore, have the relation

$$\rho_{N+1}^{(r)}(x) = \frac{(N-2)!!}{(N-1)!} \frac{e^{-\frac{x^2}{2(1+\tau)}}}{2\sqrt{1+\tau}} \left\langle |\det(xI-X)| \right\rangle_{X_N}$$

where  $ho_{N+1}^{(r)}(x)$  is the mean density of real eigenvalues in the real elliptic ensemble  $X_{N+1}$  of (N+1) imes (N+1) matrices , and the average on the right is over the real elliptic ensemble  $X_N$  of N imes N.

This relation comes in handy. Forrester and Nagao 2008 found  $\rho_{N+1}^{(r)}(x)$  in closed form in terms of Hermite polynomials, the rest is asymptotic analysis.

Averaged number of stable equilibria  $\langle \mathcal{N}_{st} \rangle$  via Rice-Kac:

$$\langle \mathcal{N}_{st} \rangle = \frac{1}{m^N N^{N/2}} \int_{-\infty}^{\infty} \langle \det(xI - X) \chi_{x}(X) \rangle_{X} \frac{e^{-\frac{Nt^2}{2}} dt}{\sqrt{2\pi/N}},$$

where  $x = (m + t\sqrt{\tau})\sqrt{N}$  and  $\chi_x(X) = 1$  if all N EVs X have real parts less than x, and  $\chi_x(X) = 0$  otherwise. No need for absolute value because of  $\chi$ 

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This yields  $\langle \mathcal{N}_{st} \rangle \to 1$  if m > 1, and if 0 < m < 1 then, to leading order in N,  $\langle \mathcal{N}_{st} \rangle \propto e^{N\Sigma_{st}}$ , with  $0 < \Sigma_{st} < \Sigma_{tot}$ , Fyodorov & Nadal 2012.

Thus, for purely gradient dynamics, as the complexity increases, there is an abrupt change from a simple set of equilibria, typically a single stable equilibrium, to a phase portrait dominated by an exponential number of unstable equilibria with an admixture of a smaller, but still exp in N, number of stable equilibria.

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Bouchaud's conjecture: in the general case of non-gradient dynamics, there exists a further phase transition in the plane  $(m,\tau)$  such that below a certain number  $\tau_c(m)$  stable equilibria are no longer exponentially abundant in the limit  $N \to \infty$  (i.e.  $\Sigma_{st}(m,\tau) \to 0$ ).

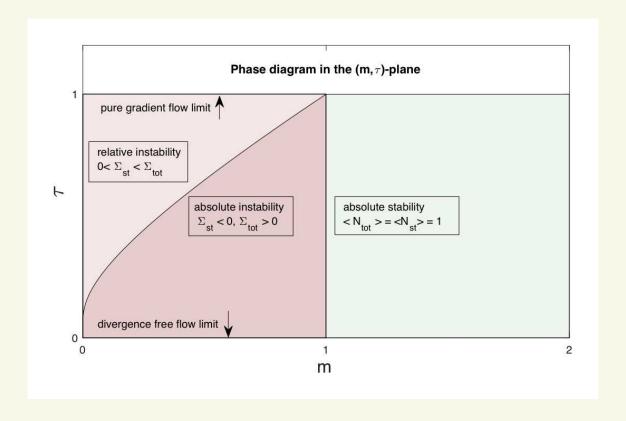
# Bouchaud's conjecture verified

**Claim** (Ben Arous, Fyodorov, Kh, unpublished, work in progress): For 0 < m < 1,

$$\lim_{N\to\infty} \frac{1}{N} \ln \langle \mathcal{N}_{st} \rangle = m - 1 - \ln m - \frac{(1-m)^2}{2\tau} = \Sigma_{st}(m,\tau).$$

Have three regions in the  $(m,\tau)$ -plane separated by critical lines  $m_c=1$  and

$$\tau_c(m) = \frac{1}{2} \frac{(1-m)^2}{m-1-\ln m}, \quad 0 < m < m_c$$



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- Our analysis is applicable to complex multi-species communities in which each kind of species on its own becomes extinct and thus interaction is key to persistence of the community.

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# THANK YOU