

MOTIVATION : FROM THEORETICAL ECOLOGY

- Model:**

The interactions between species in an ecological environment can be modeled using Lotka Volterra differential equations.

$$\begin{cases} \dot{\mathbf{x}}(t) = \mathbf{x}(t) \odot \left(\mathbf{r} - \mathbf{x}(t) + \frac{1}{\alpha\sqrt{n}} A \mathbf{x}(t) \right) \\ \mathbf{x}(t) \geq 0, \quad \forall t \end{cases}$$

→ $A_{i,j}$ is the effect of species i on species j .

- Goal:** We are interested in the equilibrium point \mathbf{x}^* and its statistical properties (distribution, number of positive components, etc).

$$\mathbf{x}^* \odot \left(\mathbf{r} - \left(I_n - \frac{A}{\alpha\sqrt{n}} \right) \mathbf{x}^* \right) = 0 \quad (1)$$

- Specs:** The interaction matrix A is a large random matrix such that the interactions $A_{i,j}$ and $A_{j,i}$ are correlated ($\rho \in [-1, 1]$). This model is very popular in theoretical ecology.

LCP

In addition to equation 1, the equilibrium \mathbf{x}^* is taken to be **Lyapunov stable** which adds another constraint for \mathbf{x}^* . Hence:

$$\begin{cases} \mathbf{x}^* \geq 0, \\ \mathbf{x}^* \odot \left(\mathbf{r} - \left(I_n - \frac{A}{\alpha\sqrt{n}} \right) \mathbf{x}^* \right) = 0, \\ \mathbf{r} - \left(I_n - \frac{A}{\alpha\sqrt{n}} \right) \mathbf{x}^* \leq 0 \end{cases}$$

Consequence: \mathbf{x}^* is a solution to LCP $\left(I_n - \frac{A}{\alpha\sqrt{n}}, -\mathbf{r} \right)$ which is a classical non-linear optimization problem.

OBJECTIVES

- Goal:** The analysis of the asymptotic behavior of the empirical distribution

$$\mu^{\mathbf{x}^*} = \frac{1}{n} \sum_{k=1}^n \delta_{x_k^*} \text{ when } n \text{ is large.}$$

- Tool:** Approximate Message Passing (AMP) can solve this problem by approximating of the measure $\mu^{\mathbf{x}^*}$ and by providing rich information of its prop-

erties.

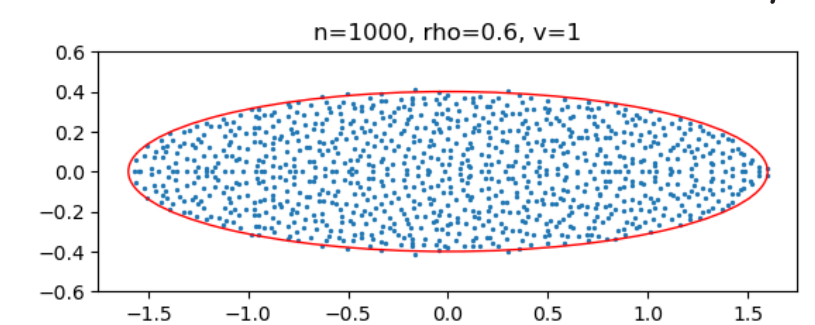
- Procedure:** For a well-chosen activation function, the iterates \mathbf{x}^k of AMP converge to a solution of LCP.

- Our contribution:** AMP is known for GOE matrices. We generalize AMP convergence results for a more general type of matrices: elliptic matrices.

ELLIPTIC MATRICES

$A \sim \text{Ellip}(n, \rho)$ if $A_{i,j} \sim \mathcal{N}(0, 1)$, $A_{i,i} \sim \mathcal{N}(0, 1 + \rho)$ and $\text{cov}(A_{i,j}, A_{j,i}) = \rho$.

The asymptotic spectrum of A is the uniform law over the ellipse of semi-major axis $= 1 + \rho$ and semi-minor axis $= 1 - \rho$.



AMP FOR GOE MATRICES

Approximate Message Passing algorithms for GOE matrices:

$$\mathbf{x}^{k+1} = \frac{1}{\sqrt{n}} A f_k(\mathbf{x}^k) - \overline{f'_k(\mathbf{x}^k)}^n f_{k-1}(\mathbf{x}^{k-1})$$

Result: Under some mild initial conditions, we have

$$\mu^{(\mathbf{x}^1, \dots, \mathbf{x}^p)} \xrightarrow[n \rightarrow \infty]{W_2} \mathcal{L}(Z^1, \dots, Z^p) \text{ a.s.}$$

Where (Z^1, \dots, Z^p) is a centered Gaussian vector whose co-variance matrix is given by Density Evolution equations.

ELEMENTS OF PROOF

Bolthausen's conditioning technique

Let $\mathbf{q}^k = f_k(\mathbf{x}^k)$, P_k is the projection matrix over $\text{span}(\mathbf{q}^0, \dots, \mathbf{q}^{k-1})$ and $\mathcal{F}_k = \sigma(\mathbf{x}^0, \dots, \mathbf{x}^k)$. Then

$$\mathbf{x}^{k+1} = \frac{1}{\sqrt{n}} A f_k(\mathbf{x}^k) - \rho b_k \mathbf{q}^{k-1}$$

$$\stackrel{\mathcal{L}}{=} |_{\mathcal{F}_k} \frac{1}{\sqrt{n}} \sum_{l=1}^k \alpha_l \mathbf{x}^l + o(1) + (\tilde{A} - \rho P_k \tilde{A}^\top) P_k^\perp$$

Where **green** is \mathcal{F}_k measurable and **red** is independant from \mathcal{F}_k .

FIXED POINT EQUATIONS

Let $\rho \in [-1, 1]$ and $Z \sim \mathcal{N}(0, 1)$, the system of unknown $(\delta, \sigma^2, \gamma)$ has a unique solution if $\alpha > 2$.

$$\begin{cases} \alpha &= \delta + \rho \frac{\gamma}{\delta} \\ \sigma^2 &= \frac{1}{\delta^2} \mathbb{E}(\sigma Z + r)_+^2 \\ \gamma &= \mathbb{P}[\sigma Z + r > 0] \end{cases}$$

Key ideas:

- Using the AMP scheme with specific activation functions $f_k(x) = \frac{(x + r_k)_+}{\delta}$, the iterates \mathbf{x}^k and \mathbf{x}^{k+1} become more correlated when k grows.
- Density evolution equations explicitly describe the variances which lead to a fixed point equation for σ^2 .

Result:

$$\mu^{\mathbf{x}^*} \xrightarrow[n \rightarrow \infty]{W_2} \mathcal{L}\left(\left(1 + \rho \frac{\gamma}{\delta^2}\right) (\sigma Z + r)_+\right) \text{ a.s.}$$

Remark. For k large enough, $\mathbf{x}^k \approx \mathbf{x}^*$.

AMP FOR ELLIPTIC MATRICES

For the elliptic matrix model: AMP iterations will be defined as:

$$\mathbf{x}^{k+1} = \frac{1}{\sqrt{n}} A f_k(\mathbf{x}^k) - \rho \overline{f'_k(\mathbf{x}^k)}^n f_{k-1}(\mathbf{x}^{k-1})$$

The "Onsager term" $\rho \overline{f'_k(\mathbf{x}^k)}^n f_{k-1}(\mathbf{x}^{k-1})$ is an interpolation between the GOE case : $\rho = 1$ and the i.i.d. case : $\rho = 0$ (where the Onsager term disappear).

PERSPECTIVES

- Prove a universality result of the convergence of AMP (elliptic model), i.e. when the entries are not necessarily gaussian.
- Generalize the AMP scheme to matrices with variance profile, i.e. models of the form $M = V \odot A$, where $A \sim \text{Ellip}(n, \rho)$ and V is a deterministic matrix.

CONCLUSION

In this work,

- we analysed the statistical properties of the equilibrium \mathbf{x}^* .
- we developed of a new AMP algorithm.

REFERENCES

- [1] Oliver Y. Feng and Ramji Venkataramanan and Cynthia Rush and Richard J. Samworth. A unifying tutorial on Approximate Message Passing, 2021.
- [2] M. Bayati and A. Montanari. The dynamics of message passing on dense graphs, with applications to compressed sensing, 2011.
- [3] Akjouj, I. and Hachem, W. and Maïda, M. and Najim, J. Equilibria of large random Lotka-Volterra systems with vanishing species : a mathematical approach, 2023.