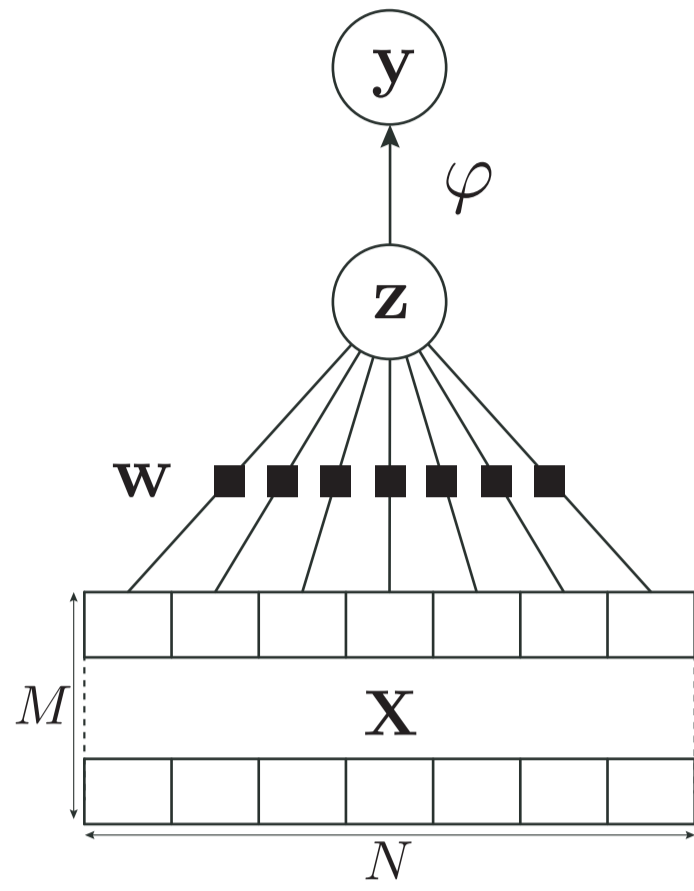


Summary

We reviewed the problem of computing the storage capacity of binary perceptron [1, 2] with random patterns. The perceptron problem is motivated by studies of simple artificial neural networks and we view it here as a random constraint satisfaction problem (CSP) where the vector of binary weights $\mathbf{w} \in \{\pm 1\}^N$ must satisfy M constraints. For the *step* constraint, the storage capacity was predicted in [1, 2] using replica method, but proving this result stays up to date an **open mathematical problem**. We consider instead **symmetric constraints** and **compute the capacity** α_c rigorously.

The simplest neural network

- $K \in \mathbb{R}$ a threshold / stability
- $X_{\mu i} \sim \mathcal{N}(0, \frac{1}{N})$ i.i.d
- $\mathbf{X} \in \mathbb{R}^{M \times N}$
- M patterns $\mathbf{X}_{\mu=1 \dots M}$
- Binary weights $\mathbf{w} \in \{\pm 1\}^N$
- Constraint density $\alpha \equiv M/N$



The perceptron *satisfies the pattern* μ if the vector \mathbf{w} satisfies the constraint:
 $\mathbf{X}_{\mu} \mathbf{w} \geq K$

Storage capacity problem

- Number of satisfying vectors:

$$\mathcal{Z}(\mathbf{X}) = \sum_{\mathbf{w} \in \{\pm 1\}^N} \mathbb{1}(\mathbf{w} \text{ satisfies all patterns } \mathbf{X}_{\mu})$$

- Free entropy:

$$\phi(\alpha) = \lim_{N \rightarrow \infty} \frac{1}{N} \mathbb{E}_{\mathbf{X}} [\log(\mathcal{Z}(\mathbf{X}))]$$

For a general constraint function φ , the **storage capacity** $\alpha_c(K)$ is defined as the largest value of α such that in the limit $N \rightarrow \infty$ and $\alpha = \mathcal{O}(1)$, all the constraints are simultaneously satisfiable with high probability (satisfiability threshold):

$$\alpha_c(K) = \inf \left\{ \alpha : \lim_{N \rightarrow \infty} \mathbb{P} [\exists \mathbf{w} / \forall \mu \in [1 : M], \varphi(\mathbf{X}_{\mu} \mathbf{w}) = 1] = 0 \right\} = \sup \left\{ \alpha : \lim_{N \rightarrow \infty} \mathbb{P} [\exists \mathbf{w} / \forall \mu, \varphi(\mathbf{X}_{\mu} \mathbf{w}) = 1] = 1 \right\}$$

- For the binary perceptron, the capacity is computed when the RS free entropy vanishes $\phi(\alpha_c) = 0$ [2]
 $\rightarrow \alpha_c^s(K=0) \simeq 0.833$, but proving this result remains a challenging mathematical problem
- We analyze **symmetric variants** of the classical step **binary perceptron**: **rectangle** and **symmetric step**

Definitions

Let $z, z' \sim \mathcal{N}(0, 1)$ i.i.d. Let $\beta \in [0, 1]$, $z_1 = \sqrt{\beta}z + \sqrt{1-\beta}z'$ and $z_2 = \sqrt{\beta}z - \sqrt{1-\beta}z'$

- $p_K^{\mathbf{s}, \mathbf{r}, \mathbf{u}} \equiv \mathbb{P}[\varphi^{\mathbf{s}, \mathbf{r}, \mathbf{u}}(z) = 1]$
- $q_K^{\mathbf{r}, \mathbf{u}}(\beta) \equiv \mathbb{P}[\varphi^{\mathbf{r}, \mathbf{u}}(z_1) = 1 \wedge \varphi^{\mathbf{r}, \mathbf{u}}(z_2) = 1]$
- $H(\beta) \equiv -\beta \log(\beta) - (1-\beta) \log(1-\beta)$
- $F_{K, \alpha}^{\mathbf{r}, \mathbf{u}}(\beta) \equiv H(\beta) + \alpha \log(q_K^{\mathbf{r}, \mathbf{u}}(\beta))$

Theorem

Theorem 1. Under the following assumption: for all choices of $K > 0$ and $\alpha > 0$ so that $\partial_{\beta}^2 F_{K, \alpha}^{\mathbf{r}, \mathbf{u}}(1/2) < 0$, there is exactly one $\beta \in (1/2, 1)$ so that $\partial_{\beta} F_{K, \alpha}^{\mathbf{r}, \mathbf{u}}(\beta) = 0$:

1. For all $K > 0$, we have $\alpha_c^{\mathbf{r}}(K) = -\log(2) / \log(p_K^{\mathbf{r}})$.
2. For all $K \in (0, K^* \simeq 0.817)$, we have $\alpha_c^{\mathbf{u}}(K) = -\log(2) / \log(p_K^{\mathbf{u}})$.

1st moment

Proposition 2. If $\alpha > \alpha_a^{\mathbf{s}, \mathbf{r}, \mathbf{u}}(K) = \frac{-\log(2)}{\log(p_K^{\mathbf{s}, \mathbf{r}, \mathbf{u}})}$, with high probability (whp) there is no satisfying assignment to the binary perceptron with the $\{\mathbf{s}, \mathbf{r}, \mathbf{u}\}$ activation functions.

Proof. $\mathbb{P}[\mathcal{Z}_{\mathbf{s}, \mathbf{r}, \mathbf{u}}(\mathbf{X}) > 0] \leq \mathbb{E}[\mathcal{Z}_{\mathbf{s}, \mathbf{r}, \mathbf{u}}(\mathbf{X})] = 2^N \mathbb{E} \left[\prod_{\mu=1}^M \varphi_{\mathbf{s}, \mathbf{r}, \mathbf{u}}(z_{\mu}(1)) \right] = 2^N p_K^{\mathbf{s}, \mathbf{r}, \mathbf{u}} = \exp(N(\log(2) + \alpha \log(p_K^{\mathbf{s}, \mathbf{r}, \mathbf{u}}))) \rightarrow 0$ as $N \rightarrow \infty$ □

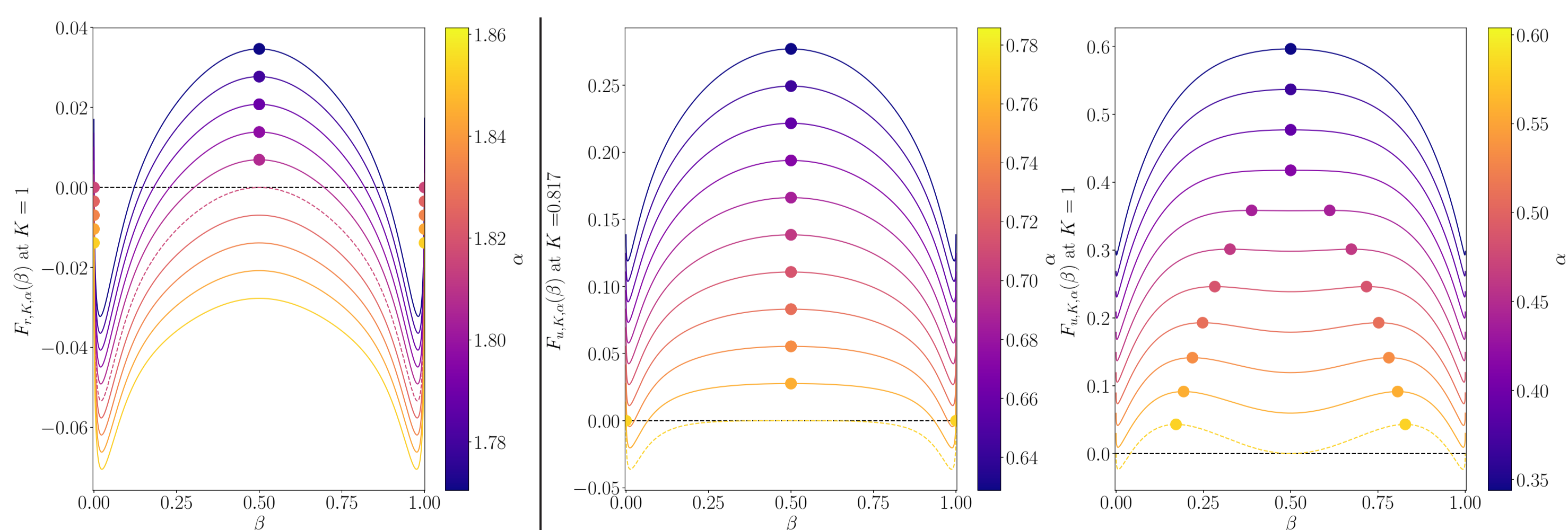
2nd moment

- Proposition 3.**
1. $\forall K > 0, \forall \alpha < \alpha_a^{\mathbf{r}}(K), \exists c_2 > 0$ s.t. $\mathbb{P}[\mathcal{Z}_{\mathbf{r}}(\mathbf{X}) > 0] \geq \frac{1}{c_2} > 0$
 2. $\forall K \in (0, K^*), \forall \alpha < \alpha_a^{\mathbf{u}}(K), \exists c_2 > 0$ s.t. $\mathbb{P}[\mathcal{Z}_{\mathbf{u}}(\mathbf{X}) > 0] \geq \frac{1}{c_2} > 0$

Proof. As $\mathcal{Z}_{\mathbf{r}, \mathbf{u}}(\mathbf{X}) \in \mathbb{N}$, $\mathbb{P}[\mathcal{Z}_{\mathbf{r}, \mathbf{u}}(\mathbf{X}) > 0] \geq \frac{\mathbb{E}[\mathcal{Z}_{\mathbf{r}, \mathbf{u}}(\mathbf{X})]^2}{\mathbb{E}[\mathcal{Z}_{\mathbf{r}, \mathbf{u}}(\mathbf{X})^2]}$

- $\mathbb{E}[\mathcal{Z}_{\mathbf{r}, \mathbf{u}}(\mathbf{X})]^2 = (2^N p_K^{\mathbf{r}, \mathbf{u}})^2$
- If the supremum is achieved for $\beta = 1/2$, using Laplace's method, $\exists c_2 > 0$ s.t:

$$\mathbb{E}[\mathcal{Z}_{\mathbf{r}, \mathbf{u}}^2(\mathbf{X})] \leq 2^N e^{N \sup_{\beta \in [0, 1]} \{F_{K, \alpha}^{\mathbf{r}, \mathbf{u}}(\beta)\}} \leq c_2 2^N e^{N(\{H(1/2) + \alpha \log(q_K^{\mathbf{r}, \mathbf{u}}(1/2))\})} = c_2 4^N (p_K^{\mathbf{r}, \mathbf{u}})^{2\alpha N}$$
- **u:** $\forall K, \forall \alpha < \alpha_a^{\mathbf{u}}(K), \beta^* = 1/2$ || **r:** $\forall K \leq K^*, \forall \alpha < \alpha_a^{\mathbf{r}}(K), \beta^* = 1/2$. Wrong for $K \geq K^*$: 2nd method fails - onset of the RSB region.



Conclusion and summary of the results

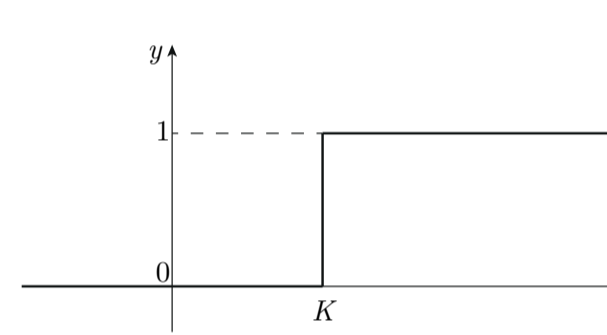
While the storage capacity for the step binary perceptron has not been proven yet, we provided a rigorous proof of the capacity for similar symmetric constraints (rectangle and symmetric step).

	Constraint	Range of K	Storage capacity
Step function	$z \geq K, \varphi^s$	$\forall K \in \mathbb{R}$	RS
Rectangle	$ z \leq K, \varphi^r$	$\forall K \in \mathbb{R}^+$	Annealed
Symmetric step	$ z \geq K, \varphi^u$	$0 < K < K^* = 0.817$	Annealed
Symmetric step	$ z \geq K, \varphi^u$	$\forall K > K^* = 0.817$	FRSB(?)

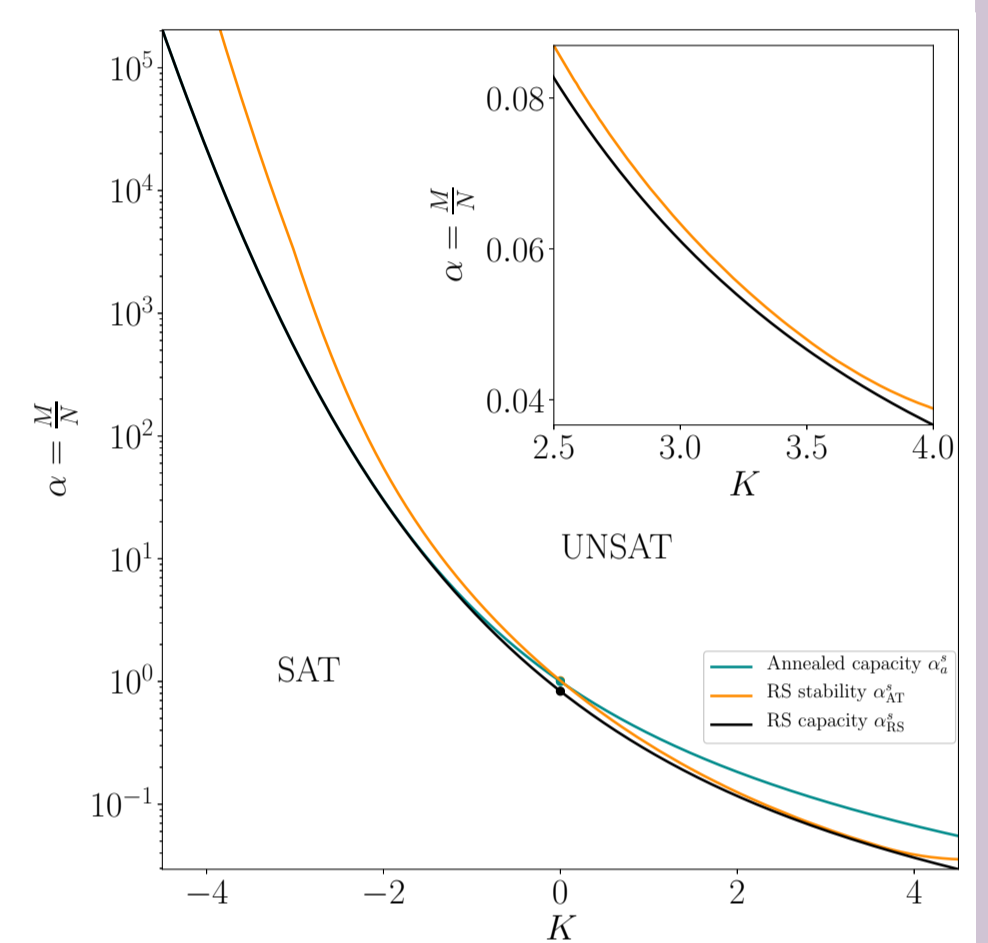
Main results and plots

◊ Step - s

$$\varphi^s(z) \equiv \mathbb{1}(z \geq K)$$

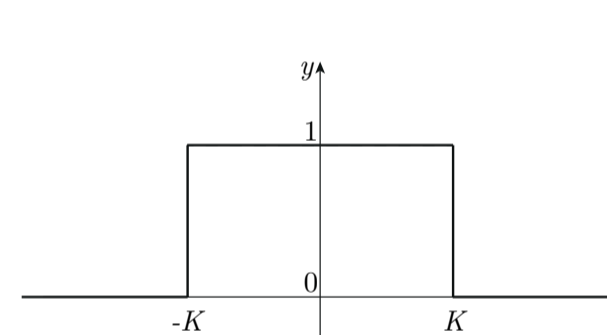


- ★ RS solution stable $\forall K \in \mathbb{R}$
- ★ $\alpha_c^s = \alpha_{RS}^s < \alpha_a^s$ for $K \in \mathbb{R}$ but **not proven rigorously**, second moment method fails...!
- ★ $\alpha_c^s(0) \simeq 0.833, \alpha_{AT}^s(0) = 1.015$
- ★ Configuration space: *frozen* 1RSB

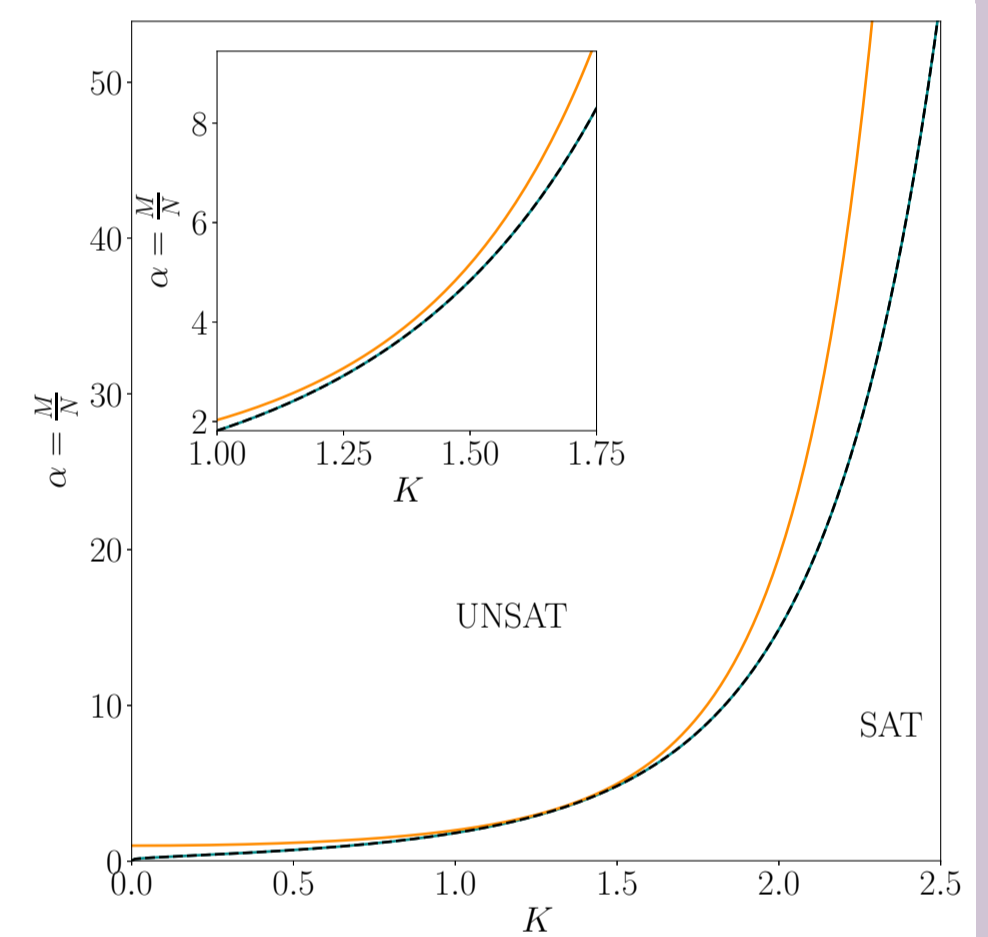


◊ Rectangle - r

$$\varphi^r(z) \equiv \mathbb{1}(|z| \leq K)$$

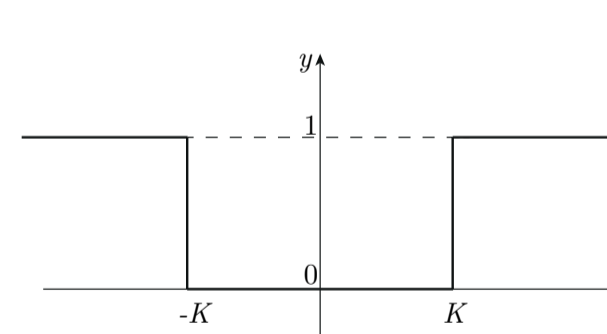


- ★ RS solution stable $\forall K \in \mathbb{R}^+$
- ★ Annealed entropy matches the RS entropy: $\phi_a = \phi_{RS}$
- ★ $\forall K > 0, \alpha_c^r(K) = \alpha_a^r(K) = \alpha_{RS}^r(K)$: by 2nd moment method
- ★ Configuration space: *frozen* 1RSB

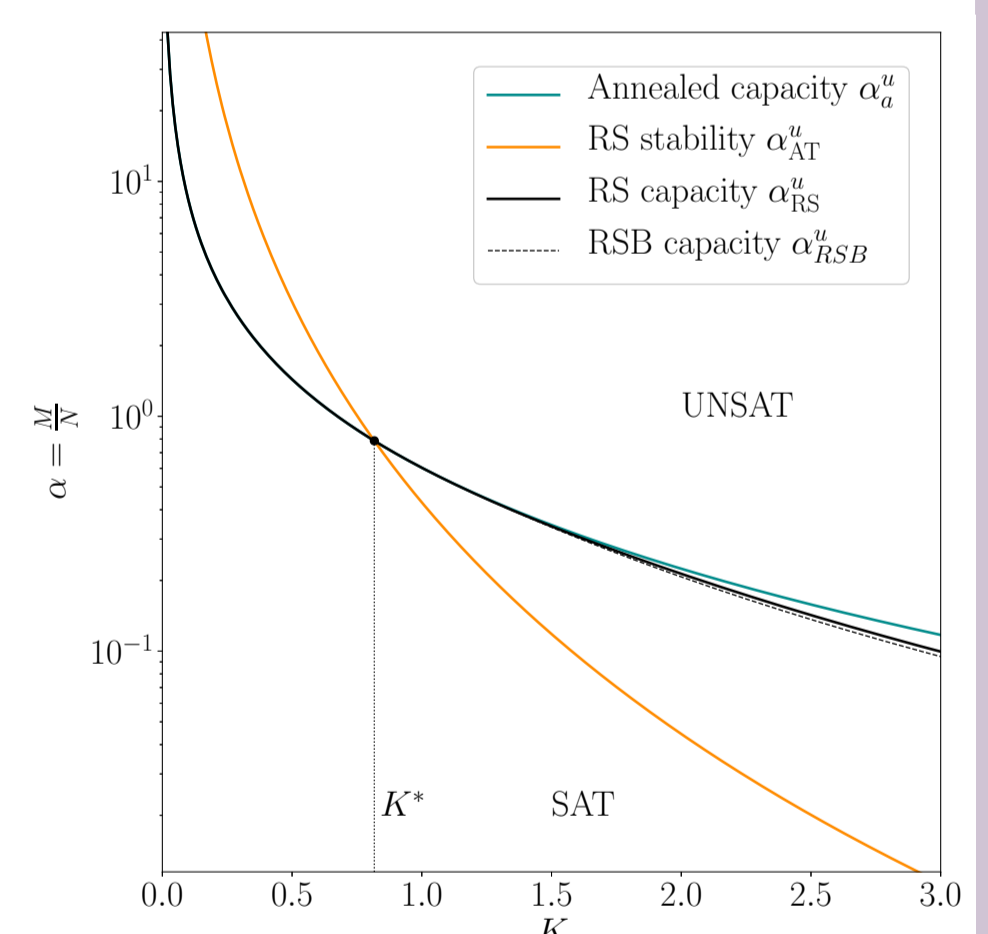


◊ Symmetric step - u

$$\varphi^u(z) \equiv \mathbb{1}(|z| \geq K)$$

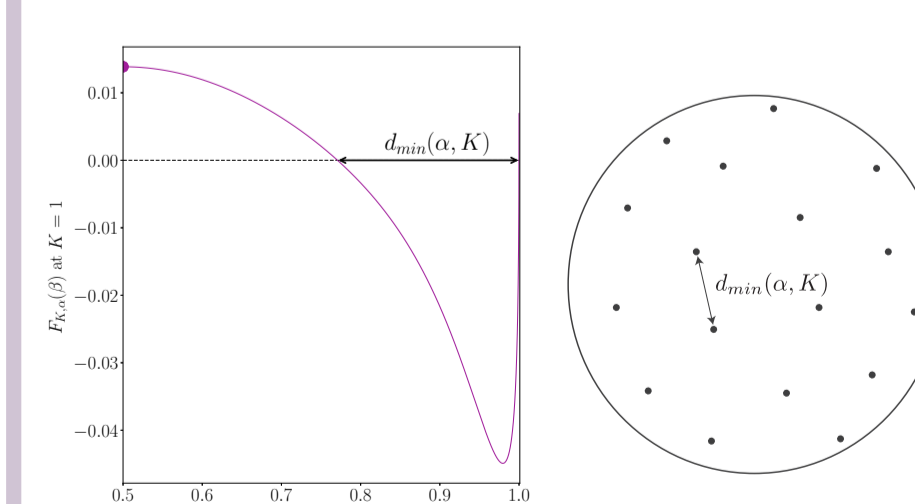


- For $K < K^* \simeq 0.817$
 - ★ RS solution stable
 - ★ Annealed entropy matches the RS entropy: $\phi_a = \phi_{RS}$
 - ★ $\alpha_c^u(K) = \alpha_a^u(K) = \alpha_{RS}^u(K)$: by 2nd moment method
 - ★ Configuration space: *frozen* 1RSB
- For $K \geq K^*$
 - ★ RS solution becomes unstable at $K = K^*$
 - ★ 2nd moment method fails exactly at the same value K^* (AT line / RS capacity)
 - ★ 1RSB solution unstable \rightarrow Full RSB ?



Frozen 1RSB

- $\partial_{\beta} F_{K, \alpha}^{\mathbf{r}, \mathbf{u}}|_{\beta=1} = +\infty$
- $\forall K, \alpha, \exists d_{min}(\alpha, K) : \forall d \in]0, d_{min}[, \forall \mathbf{w}_1, \mathbf{w}_2 \in \{\pm 1\}^N$
SAT, $\mathbb{P}[\frac{1}{N} \mathbf{w}_1^T \mathbf{w}_2 = 1 - d] = 0$

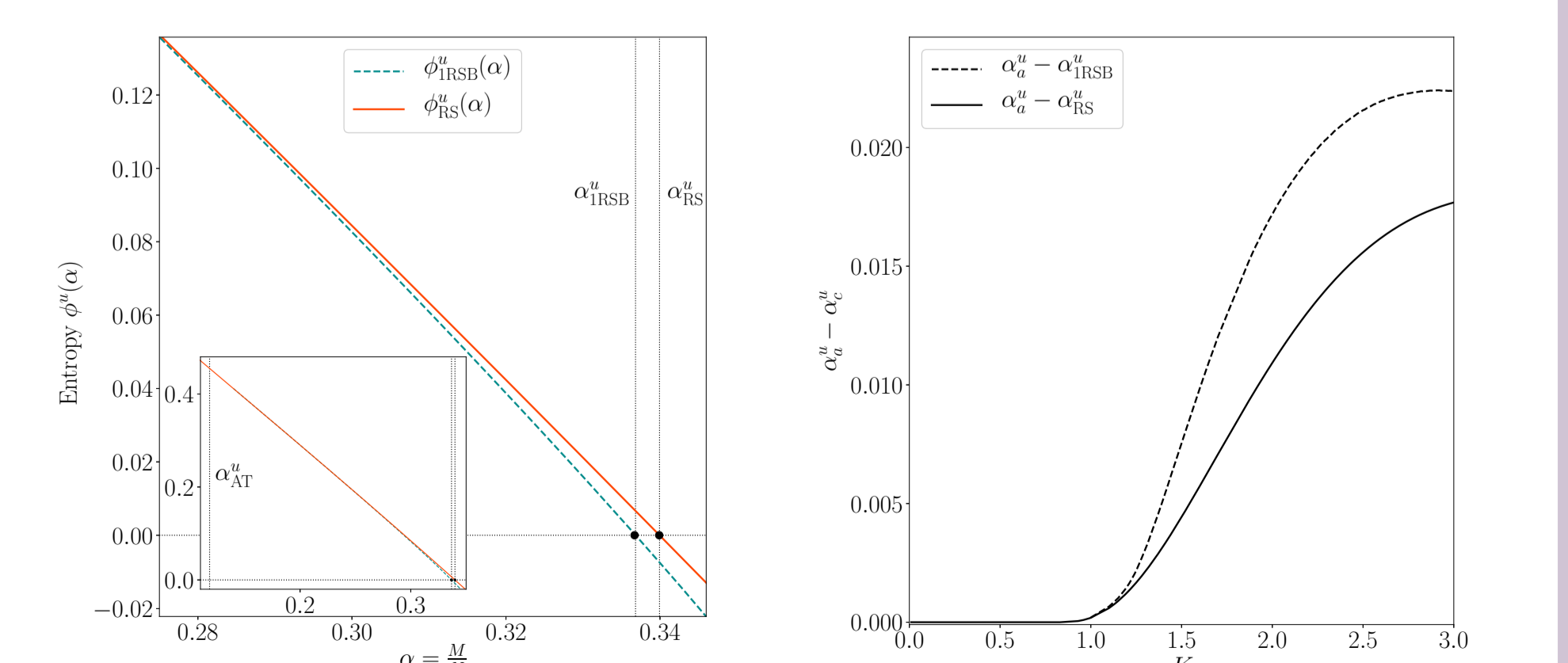


- The configuration space is *frozen*: clusters of single solutions at a distance $d \geq d_{min}$
- True for **s, r** and **u**

Replicas

$$\phi(\alpha) = \lim_{N \rightarrow +\infty} \lim_{n \rightarrow 0} \frac{1}{Nn} \frac{\partial \log(\mathbb{E}_{\mathbf{X}}[\mathcal{Z}(\mathbf{X})^n])}{\partial n}$$

- RS solution stable for $K < K^*$, where K^* is the intersection between AT-line and annealed capacity
- 1RSB capacity provides a small correction
- 1RSB solution unstable ($q_0 \neq 0$ while symmetry should impose $q_0 = 0$)
- Probably Full RSB (analysis to be done)



References

- [1] E. Gardner & B. Derrida. Optimal storage properties of neural network models. *J. Phys. A: Math. and Gen.*, 1988.
- [2] W. Krauth & M. Mézard. Storage capacity of memory networks with binary couplings. *J. Phys. France*, 1989.