

El modelo de Hopfield 2

Tenemos N neuronas binarias

$$S_i(t + \Delta t) = \text{signo}(h_i(t))$$

$$= \text{signo}\left(\sum_{j \neq i} w_{ij} S_j(t)\right)$$

$N(N-1)$ sinapsis recíprocas que

$$w_{ij} = w_{ji}$$

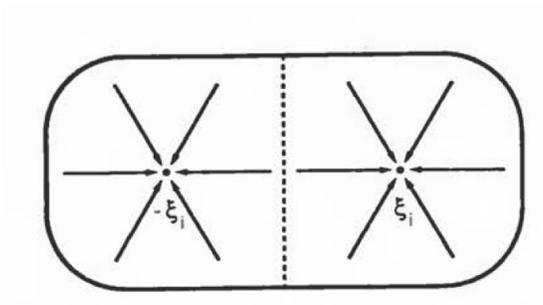
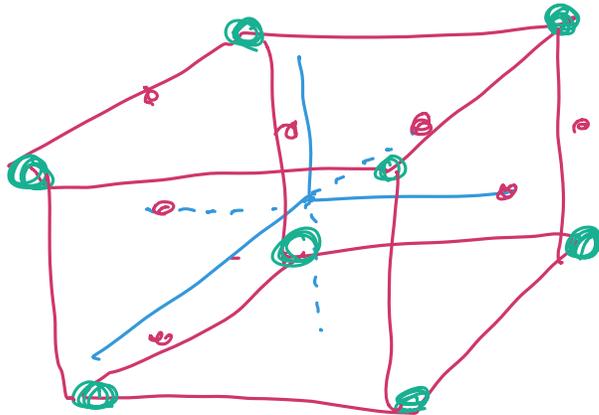
$\frac{N(N-1)}{2}$ sinapsis independientes

Para 1 memoria:

$$w_{ij} = w_{ji} = \frac{1}{N} \sum_i^1 \sum_j^1$$

$$\xi_i = \pm 1 \quad i=1, \dots, N$$

$$\sum_i \xi_i = \pm 1$$

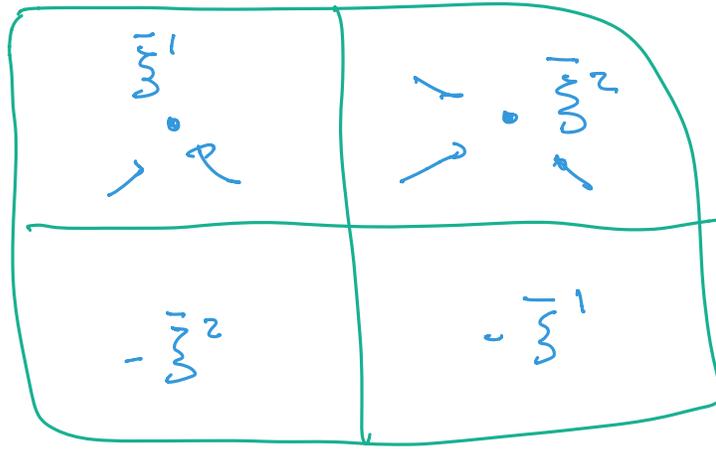


Mattis

c. ¿Cómo puedo introducir más patrones?

Si son $p=2$ patrones

$$W_{ij} = \frac{1}{2} \left(\sum_i^1 \sum_j^1 + \sum_i^2 \sum_j^2 \right)$$



van Hemmen

$S_i \quad p > 2$

$$W_{ij} = \frac{1}{N} \sum_{\mu=1}^P \sum_i^{\mu} \sum_j^{\mu}$$

$\dot{S}_i \quad S_i(0) = S_i^0 \quad \text{edwede estere' en } S_i(1) ?$

Análisis de estabilidad

$$\text{See } S_i(0) = \sum_i^v \quad \forall i$$

$$S_i(t) = \text{signo} (h_i(t))$$

$$= \text{signo} \left(\sum_{j \neq i} W_{ij} S_j(0) \right)$$

$$= \text{signo} \left(\sum_{j \neq i} \left(\frac{1}{N} \sum_{\mu=1}^P \sum_i^{\mu} \sum_j^{\mu} \right) \sum_j^v \right)$$

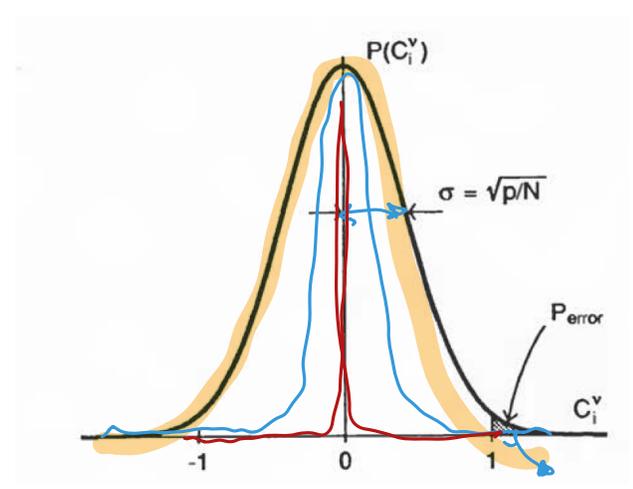
$$= \text{signo} \left(\underbrace{\sum_{i=1}^N v_i} + \sum_{i=1}^N \sum_{\mu \neq i}^{p-1} \underbrace{\sum_{k=1}^M x_{i\mu}^k \sum_{j=1}^M x_{j\mu}^k}_{> 0} \right)$$

$$= \text{signo} \left(\sum_{i=1}^N v_i \left(1 + \underbrace{\sum_{i=1}^N \sum_{\mu \neq i}^{p-1} \sum_{k=1}^M x_{i\mu}^k \sum_{j=1}^M x_{j\mu}^k}_{> 0} \right) \right) = \text{signo} \left(\sum_{i=1}^N v_i \right) \underbrace{c_i}_{> 1}$$

$$= \text{signo} \left(\sum_{i=1}^N v_i \right) c_i \left| \sum_{i=1}^N \sum_{\mu \neq i}^{p-1} \sum_{k=1}^M x_{i\mu}^k \sum_{j=1}^M x_{j\mu}^k \right| > 1$$

$$P(\{x_{i\mu}^k\}_{i=1, \mu=1, k=1}^{pN}) = \prod_{\mu=1}^p \prod_{i=1}^N \phi(x_{i\mu}^k)$$

$$\phi(x_{i\mu}^k) = \begin{cases} +1 & \text{if } < \frac{1}{2} \\ -1 & \text{if } > \frac{1}{2} \end{cases}$$

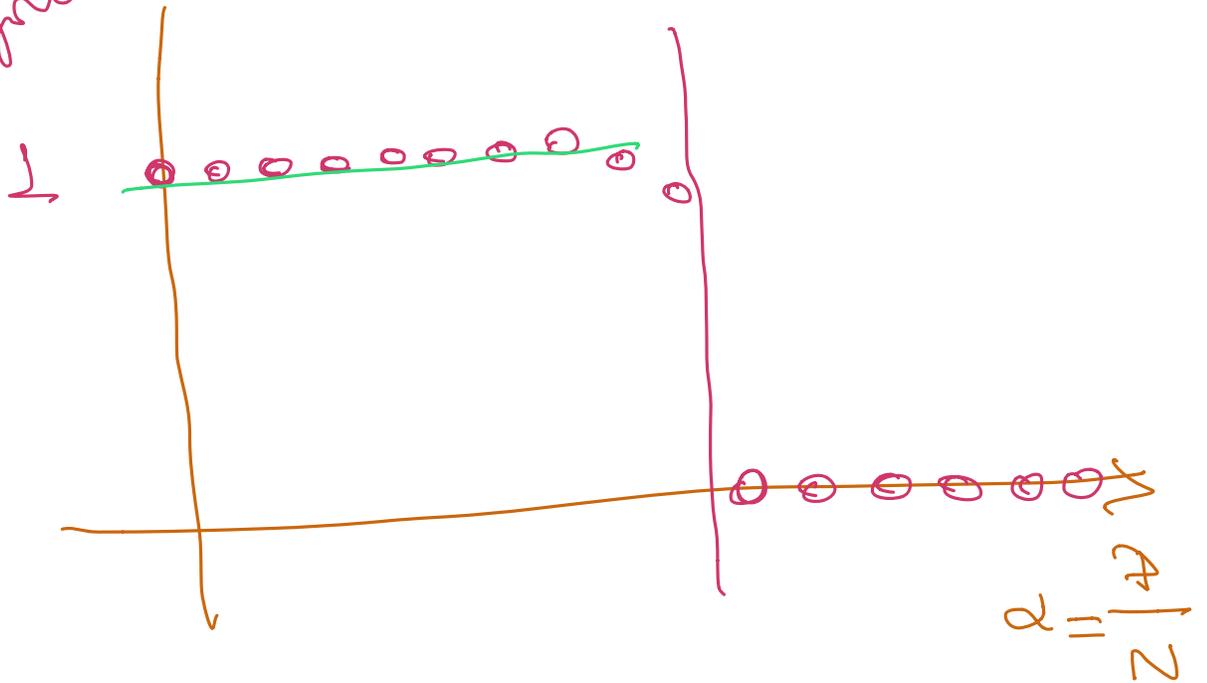


$$c_i^v = 1 + \sum_{i=1}^N \sum_{\mu=1}^{p-1} \sum_{k=1}^M x_{i\mu}^k \sum_{j=1}^M x_{j\mu}^k$$

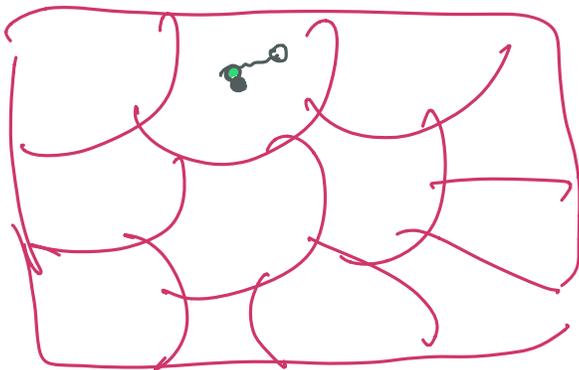
$$P_{\text{err}} = \frac{1}{\sqrt{2\pi\sigma^2}} \int_1^{\infty} e^{-\frac{x^2}{2\sigma^2}} dx$$

gjs con $\frac{P}{N} \ll 1$

Performance



$$\frac{P}{N} \approx 0.138$$



$$M^v = \frac{1}{N} \sum_{i=1}^N \sum_i^v \underline{S_i(\infty)}$$

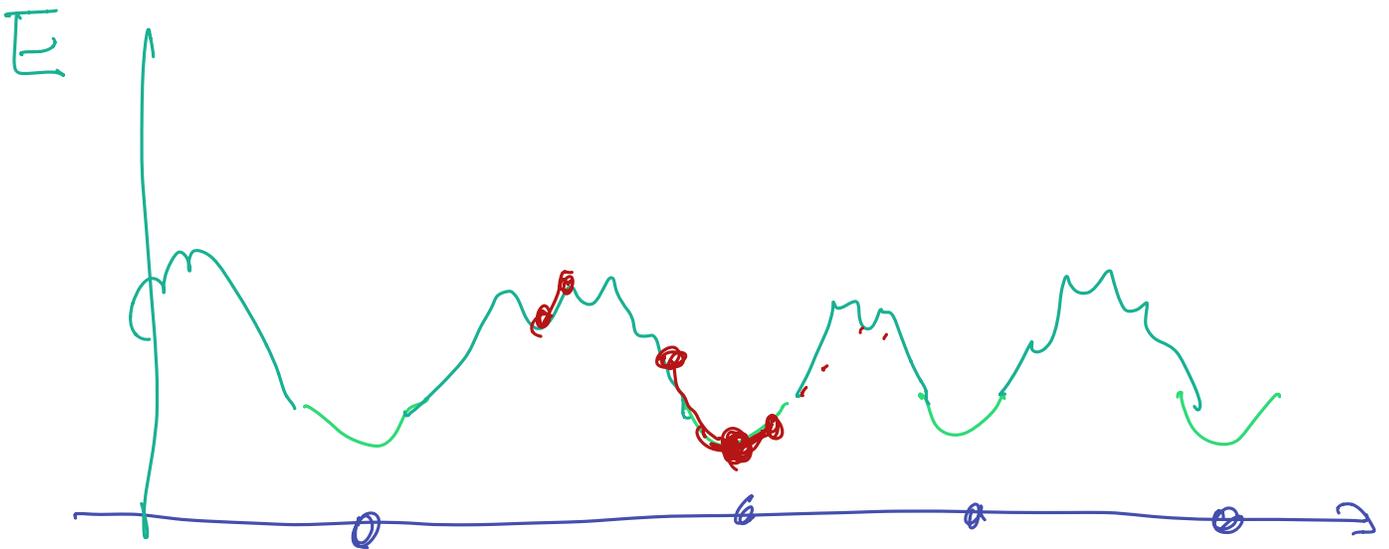
Si $M^v = 1$ depende del patrón

$M^v \approx 1$ estoy cerca del patrón

$$E(t) = -\frac{1}{2} \sum_i^N \sum_j^N w_{ij} S_i(t) S_j(t)$$

$$\rightarrow \underline{E(t+1) \leq E(t)}$$

$$E \geq E_{\min}$$

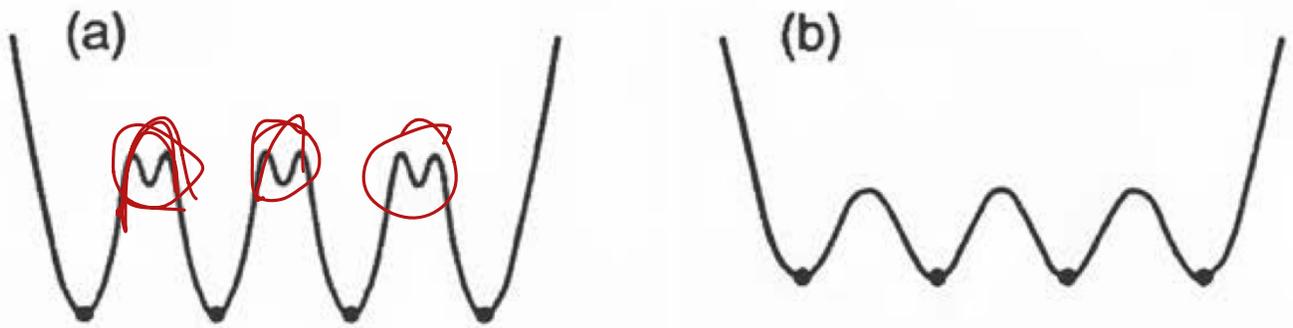


Amit, Gutfreund, Sompolinsky (1985)

$$\frac{P}{N} \rightarrow 0$$

$$\frac{P}{N} > 0$$

1987



Neurons probabilities

$$\text{Prob}(S_i(t+1) = 1) = \frac{1}{1 + e^{-\frac{2h}{T}}}$$

$$\text{Prob}(S_i(t+1) = -1) = 1 - \frac{1}{1 + e^{-\frac{2h_i}{T}}}$$

