

30 Clase Hopfield

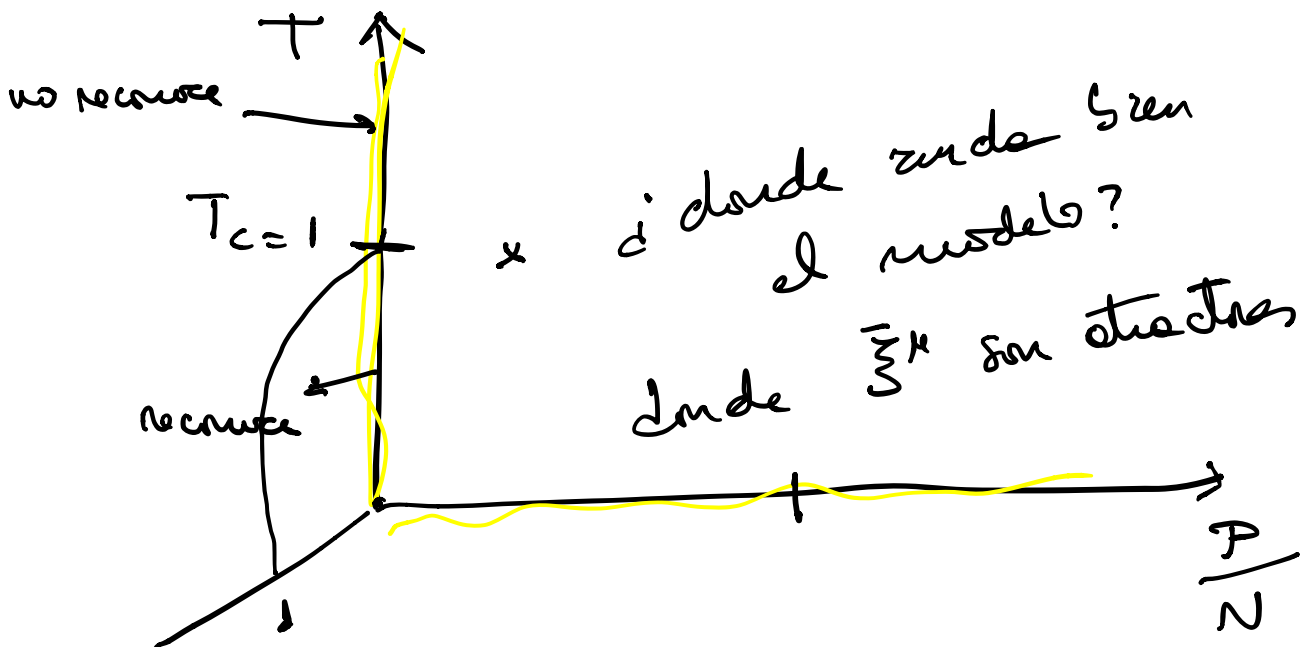
Ayer mostramos que la regla de Hopfield se puede deducir de la función energía

$$E_w(\{S\}) = -\frac{1}{2} \sum_i \sum_j w_{ij} S_i S_j$$

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

\sum_i^N : configuración almacenada ($\mu=1, \dots, P$)

Tenemos 2 parámetros haciendo un diagrama de fase



P : es el # de neuronas

Guardamos PN bits

calculo $N(N-1)$ acoplamiento w_{ij}

una medida del esfuerzo

$$\frac{\# \text{ bits}}{\# \text{ acoplamiento}} = \frac{PN}{N(N-1)} \approx \frac{P}{N}$$

si $N \gg 1$

T : pseudo temperatura (mido)

Dinámica estocástica $E = -\frac{1}{2} \sum_i E_i$

$$P^+ \equiv P(S_i(t + \frac{1}{N}) = +1) = \frac{1}{2} \left[1 + \tanh\left(\frac{h}{T}\right) \right] = P^+$$

$$P^- \equiv P(S_i(t + \frac{1}{N}) = -1) = \frac{1}{2} \left[1 - \tanh\left(\frac{h}{T}\right) \right] = P^-$$

$$\text{Si } T=0 \quad S_i(t + \frac{1}{N}) = \text{signo}(h)$$

$$T=\infty \quad P^+ = P^- = \frac{1}{2}$$

ΔC fund calculo ω

$$M^H = \frac{1}{N} \sum_i \sum_i^H S_i(t=\infty)$$

la red anda bien si en $t=\infty$

$$M^H \approx 1$$

Elapir las configuraciones aleatorias

$$P(S_i^H = 1) = \frac{1}{2}$$

$$P(S_i^H = -1) = \frac{1}{2}$$

Tengo que calcular $\langle M^H \rangle$ promedio



$$\langle M^H \rangle_+ = \left\langle \frac{1}{N} \sum_i \sum_i^H S_i \right\rangle_+ = \frac{1}{N} \sum_i \sum_i^H \langle S_i \rangle_+$$

$$\langle S_i \rangle_+ = +1 P(S_i=1) + (-1) P(-1)$$

$$\rightarrow = \frac{1}{2} + \frac{1}{2} \tanh\left(\frac{h}{T}\right) - \frac{1}{2} + \frac{1}{2} \tanh\left(\frac{h}{T}\right)$$

$$\rightarrow = \tanh\left(\frac{h}{T}\right)$$

$$= \tanh \left(\frac{\sum_j w_{ij} \langle S_j \rangle}{T} \right)$$

$$\langle S_1 \rangle = \tanh \left(\frac{w_{11} \langle S_1 \rangle + w_{12} \langle S_2 \rangle + \dots + w_{1N} \langle S_N \rangle}{T} \right)$$

⋮

$$\langle S_N \rangle = \tanh \left(\frac{w_{N1} \langle S_1 \rangle + w_{N2} \langle S_2 \rangle + \dots + w_{NN} \langle S_N \rangle}{T} \right)$$

$$\langle m^H \rangle = \frac{1}{N} \sum_i \xi_i^H \langle S_i \rangle$$

$$\langle m^V \rangle = \frac{1}{N} \sum_i \xi_i^V \langle S_i \rangle$$

$$= \frac{1}{N} \sum_i \xi_i^V \tanh \left(\frac{1}{T} \sum_{j=1}^N w_{ij} \langle S_j \rangle \right)$$

$$= \frac{1}{N} \sum_i \xi_i^V \tanh \left(\frac{1}{T} \sum_j \sum_{\mu=1}^M \frac{\xi_i^{\mu} \xi_j^{\mu}}{2} \langle S_j \rangle \right)$$

$$= \frac{1}{N} \sum_i \xi_i^V \tanh \left(\frac{1}{T} \sum_{\mu=1}^M \xi_i^{\mu} \left[\sum_j \xi_j^{\mu} \langle S_j \rangle \right] \right)$$

$$\langle M^v \rangle = \frac{1}{N} \sum_i \sum_i^v \tanh \left(\frac{1}{T} \sum_{\mu} \sum_i^{\mu} \langle M^{\mu} \rangle \right)$$

$v=1 \dots P$

Una solución buena

$$\langle M \rangle = (\langle M^1 \rangle, \langle M^2 \rangle, \dots, \langle M^P \rangle)$$

$$\cong (0, 0, \dots, \underset{\substack{\uparrow \\ v}}{+1}, \dots, 0)$$

Hay 2^P buenas soluciones

Supongamos $P=3$

$$(1, 0, 0)$$

$$(-1, 0, 0)$$

$$(0, 1, 0)$$

$$(0, -1, 0)$$

$$(0, 0, 1)$$

$$(0, 0, -1)$$

$$\begin{pmatrix} +1 & 0 & 0 \\ 0 & +1 & 0 \\ 0 & 0 & +1 \end{pmatrix}$$

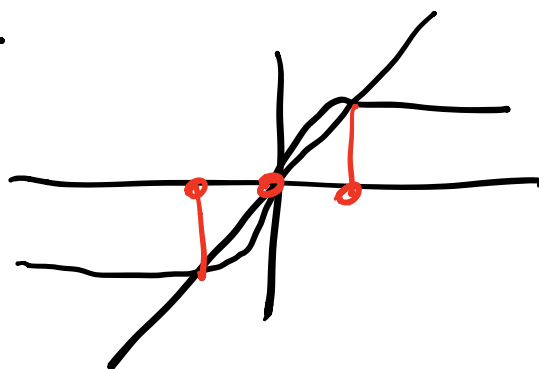
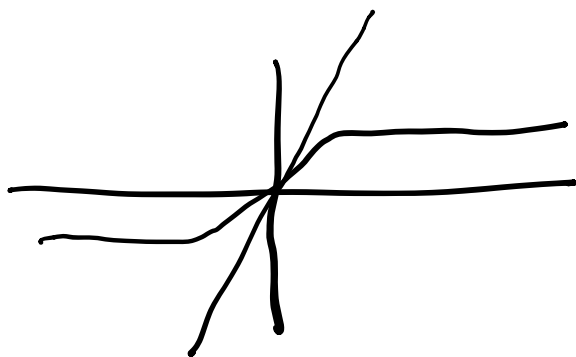
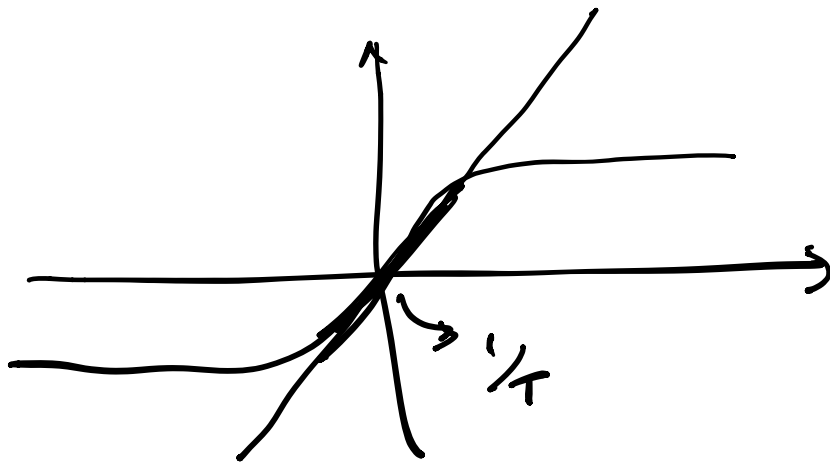
Superparam $\bar{m} = (0, 0, m^v, 0 \dots 0)$

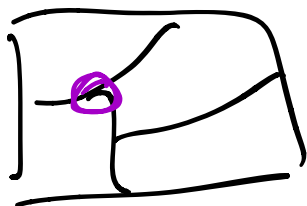
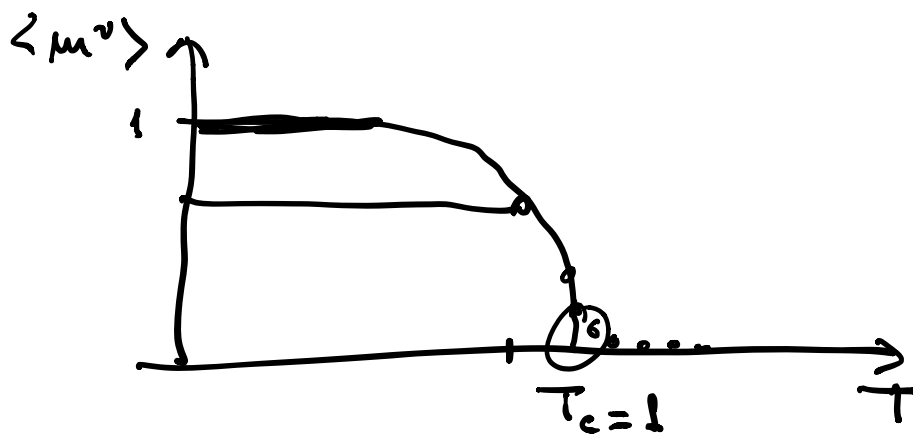
$$\langle m^v \rangle = \frac{1}{N} \sum_i s_i^v \tanh \left(\frac{1}{T} s_i^v \langle m^v \rangle \right)$$

$$= \frac{1}{N} \sum_i \tanh \left(\frac{s_i^v s_i^v \langle m^v \rangle}{T} \right)$$

$$\langle m^v \rangle = \tanh \left(\frac{\langle m^v \rangle}{T} \right)$$

$$\tanh \left(\frac{x}{T} \right) = x$$





$$N \gg 1$$

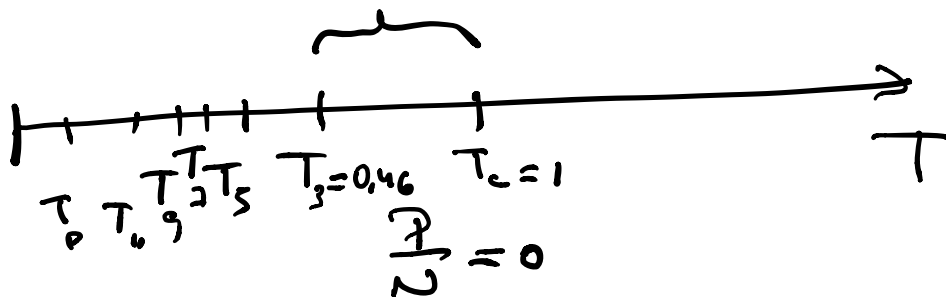
$$N \gg P$$

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

Superficies $\bar{\mu} = (0, \dots, \mu^v, \dots, \mu^x, \dots)$

no \exists estas soluciones

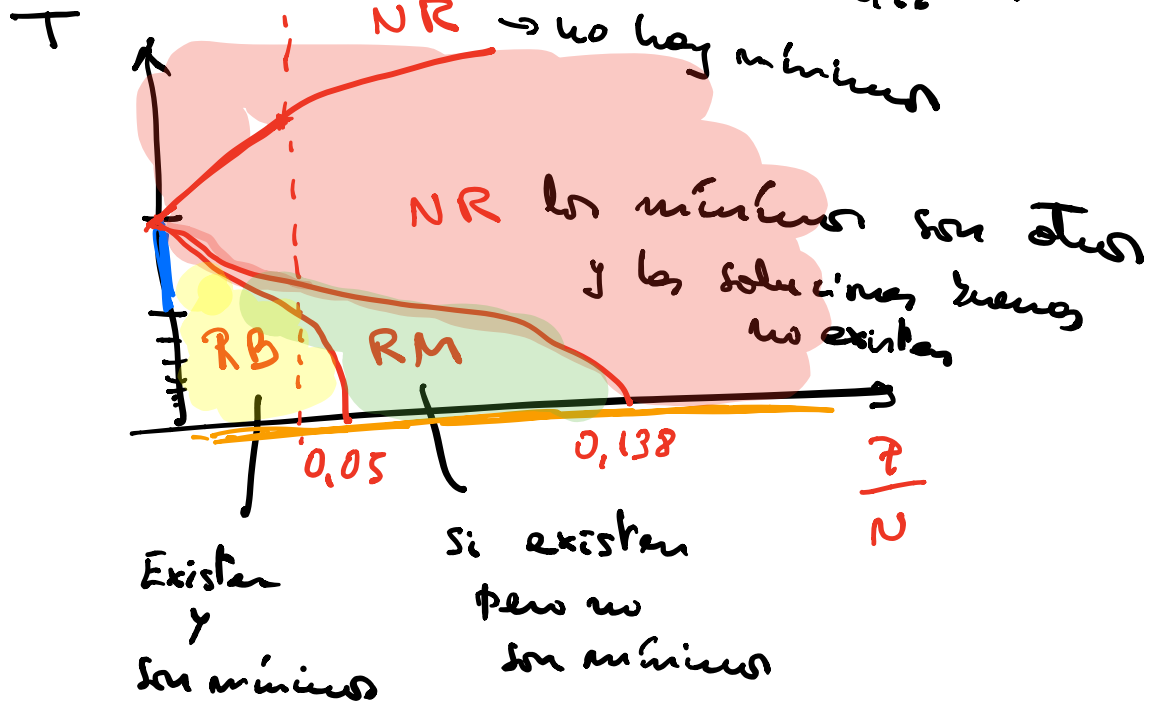
$$\bar{\mu} = (0, \dots, \mu^v, \dots, \mu^x, \dots, \mu^s, \dots)$$



$$\langle \mu^i \rangle = \tanh \left(\frac{1}{T} \langle \mu_{i1} \rangle + \langle \mu_{i2} \rangle \langle \mu_{i2} \rangle \right)$$

$(M^2) = \dots$

L_{in}



$T = 0$

