

$$A(s)$$

$$\langle A \rangle = \int_{t_t}^{t_t + t_e} \frac{\Delta(t) dt}{t_e}$$

$$h$$

$$P(s=+1) = \frac{1}{1 + e^{\beta \Delta E} \frac{1}{T}}$$

7 random entre 1 y 0

$$s_i \quad 0 \leq q \leq 1 \quad S_i = +1$$

$$-1 \leq q \leq 0 \quad S_i = -1$$

Maxima Boltzman

es una maxima probabilistica

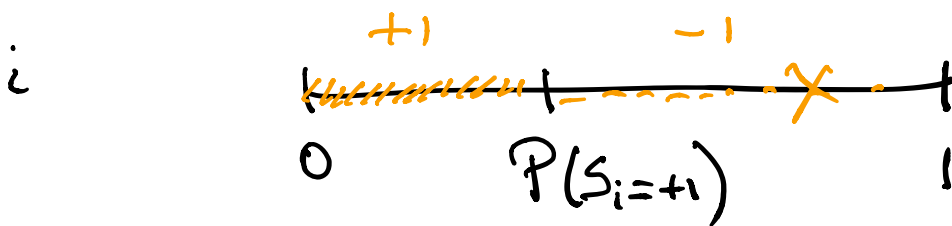
Supongamos que los acoplamiento

son simetricos.  $w_{ij} = w_{ji} = \frac{1}{N} \sum_{\mu} \sum_i^{\mu} \sum_j^{\mu}$

$$P(S_i = +1) = \frac{1}{2} \left[ 1 + \tanh\left(\frac{h_i}{T}\right) \right]$$

$$\begin{aligned} P(S_i = -1) &= 1 - P(S_i = +1) \\ &= \frac{1}{2} \left[ 1 - \tanh\left(\frac{h_i}{T}\right) \right] \end{aligned}$$

$$\begin{aligned} P(S_i) &= \frac{1}{2} \left[ 1 + \tanh\left(\frac{h_i S_i}{T}\right) \right] \\ &= \frac{1}{2} \left[ 1 + S_i \tanh\left(\frac{h_i}{T}\right) \right] \end{aligned}$$



for  $i := 1$  to  $N$

$h := 0$

for  $j := 1$  to  $N$

$h := h + w(i,j) * S[j];$

end

$P_+ := 0.5 * [1 + \tanh(h/T)];$

$r := \text{rand}$  ( $r$  in  $(0,1]$ )

if ( $r \leq P_+$ ) then

$S[i] := +1$

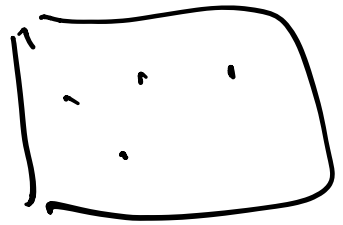
else

$S[i] := -1$

end

end

— conj. particular



¿  $P(\{S\})$ ?

$$\frac{\partial P(\{S\}, t)}{\partial t} = \sum_{\{S'\}} T(S' \rightarrow S) P(S') - \sum_{\{S'\}} T(S \rightarrow S') P(S)$$

En el estado estacionario  $\frac{\partial P}{\partial t} = 0$

o. toda la suma es cero

o hay un equilibrio de flujos

$$\forall S', S \quad [T(S' \rightarrow S) P(S') - T(S \rightarrow S') P(S)] = 0$$

$$T(S' \rightarrow S) = \frac{1}{2} \left[ 1 + \tanh \left( \frac{S'h}{T} \right) \right]$$

$$T(S \rightarrow S') = \frac{1}{2} \left[ 1 + \tanh \left( \frac{S'h}{T} \right) \right]$$

$$\frac{P(s)}{P(s')} = \frac{T(s' \rightarrow s)}{T(s \rightarrow s')}$$

$$= \frac{\frac{1}{2} [1 + \tanh(s h/T)]}{\frac{1}{2} [1 + \tanh(s' h/T)]}$$

$$1 + \tanh(x) = 1 + \frac{e^x - e^{-x}}{e^x + e^{-x}} = \frac{e^x + \cancel{e^{-x}} + e^x - \cancel{e^{-x}}}{e^x + e^{-x}}$$

$$= \frac{2e^x}{e^x + e^{-x}} \quad x = \frac{hs}{T}$$

$$\frac{P(s)}{P(s')} = \frac{e^x}{e^x + e^{-x}} \cdot \frac{1}{\frac{e^{+x'}}{e^{x'} + e^{-x'}}$$

$$x' = -x$$

$$= e^{x - x'}$$

$$\frac{P(s)}{P(s')} = \frac{e^{\frac{hs}{T}}}{e^{\frac{hs'}{T}}}$$

Dame  $\{S\}$  y te digo la probabilidad en el estado estacionario

$$P(s) \propto e^{-\frac{hs}{T}}$$

$$\sum_{\{s\}} P(s) = Z \quad \text{— función partición}$$

$$P(s) = \frac{e^{-\frac{hs}{T}}}{Z}$$

Ahora  $\sum_{\{s\}} P(s) = 1 \quad F = -k_B T \ln(Z)$

Modelo: el sistema hace un  
paseo aleatorio entre todas  
las posibles conf.

∫ la prob. de una conf.

es  $P(s) = \frac{e^{-sh/T}}{Z}$

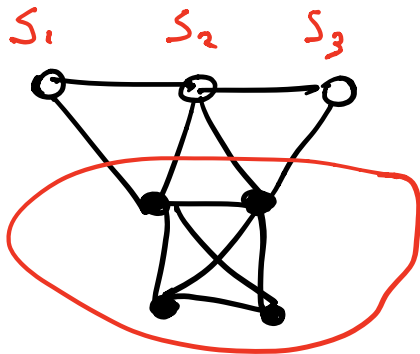
$$Z = \sum_{\{s\}} e^{-\frac{sh}{T}}$$

1983 Hinton y Sejnowski

Divide to red

red visible

red oculta



- neurone visible
- neurone oculta

$\bar{S} = (S_1, S_2, S_3)$  visible

$P=0.7$	+	+	+	}	oculto
$P=0.1$	-	+	+		
$P=0.1$	+	+	-		
$P=0.1$	+	-	+		
0	-	-	+		
0		⋮			

Tarefas  $N_T$  neurones

$K$  ocultos

$N$  visíveis

$$N_T = K + N$$

los denotamos con los índices  
 $\alpha$  recorre los visibles  
 $\beta$  recorre los ocultos

$P_{\alpha\beta}$ : prob de encontrar al sist. en  
 $\{S_{\alpha\beta}\}$

$$P_{\alpha\beta} = P(\{S_{\alpha\beta}\}) = \frac{e^{-H_{\alpha\beta}/T}}{Z}$$

$$H_{\alpha\beta} = -\frac{1}{2} \sum_i \sum_j w_{ij} S_i^\alpha S_j^\beta$$

Ahora minimizar la función

Entropía

Sea  $P_\alpha$  la prob de estado en que  
los visibles aparecen

$$P_\alpha = \sum_\beta P_{\alpha\beta} = \sum_\beta \frac{e^{-H_{\alpha\beta}/T}}{Z}$$

Entropia  $S_e = - \sum_{\alpha} R_{\alpha} \log \left( \frac{R_{\alpha}}{P_{\alpha}} \right) \equiv 0$

descenso por el precedente en  $S_e$

$$\Delta w_{ij} = -\eta \frac{\partial S_e}{\partial w_{ij}}$$

$$w_{ij}^{\text{new}} = w_{ij}^{\text{old}} + \Delta w_{ij}$$

$$\Delta w_{ij} = -\eta \frac{\partial S_e}{\partial w_{ij}} = \eta \sum_{\alpha} \frac{R_{\alpha}}{P_{\alpha}} \frac{\partial P_{\alpha}}{\partial w_{ij}}$$

$$\Delta w_{ij} = \frac{\eta}{T} \left[ \sum_{\alpha} \sum_{\beta} \frac{R_{\alpha}}{P_{\alpha}} s_i^{\alpha\beta} s_j^{\alpha\beta} P_{\alpha\beta} - \sum_{\alpha} R_{\alpha} \langle s_i s_j \rangle \right]$$

$$= \frac{\eta}{T} \left[ \langle s_i s_j \rangle_{\text{classified}} - \langle s_i s_j \rangle_{\text{true}} \right]$$







input

output

$$\Delta W_{ij} = \frac{\eta}{T} \left[ \langle S_i S_j \rangle_{\text{clamped}} - \langle S_i S_j \rangle_{\text{I}} \right]$$

## Project 3

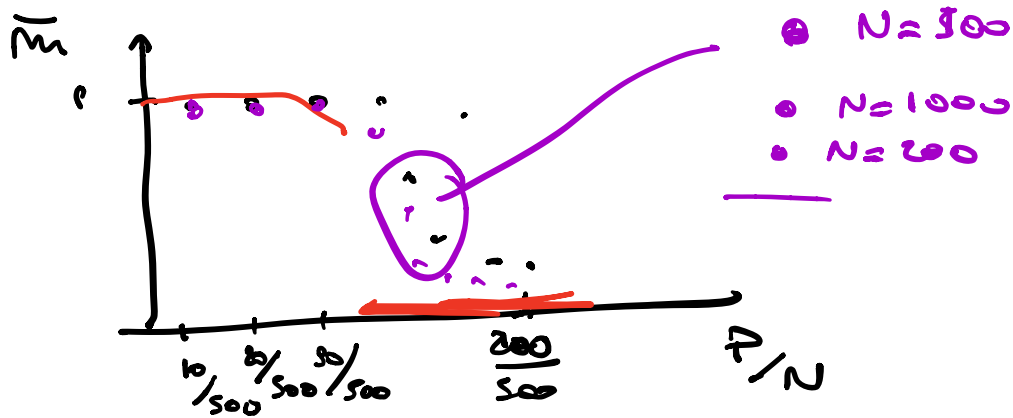
Hopfield 1  $N = 500$  neurons

Figen  $\phi$  2 curiencas de de c/memoria

$$M^M = \frac{1}{N} \sum_i \sum_j S_i^M S_j^M$$

$$\bar{M} = \frac{1}{P} \sum_{P=1}^P M^M$$

$P = 10, 20, 30, \dots, 200$



$$N = 1000$$

$$P = 20, 40, \dots, 400$$

$$N = 2000$$

$$P = 40, 80, \dots, 800$$