

Ludwig Boltzmann



- [Ludwig Boltzmann - Wikipedia, the free encyclopedia](#)

Boltzmann distribution

- [Boltzmann distribution - Wikipedia, the free encyclopedia](#)

Maxwell-Boltzmann distribution

- 馬克斯威-波茲曼分佈

- Maxwell_Boltzmann_statistics

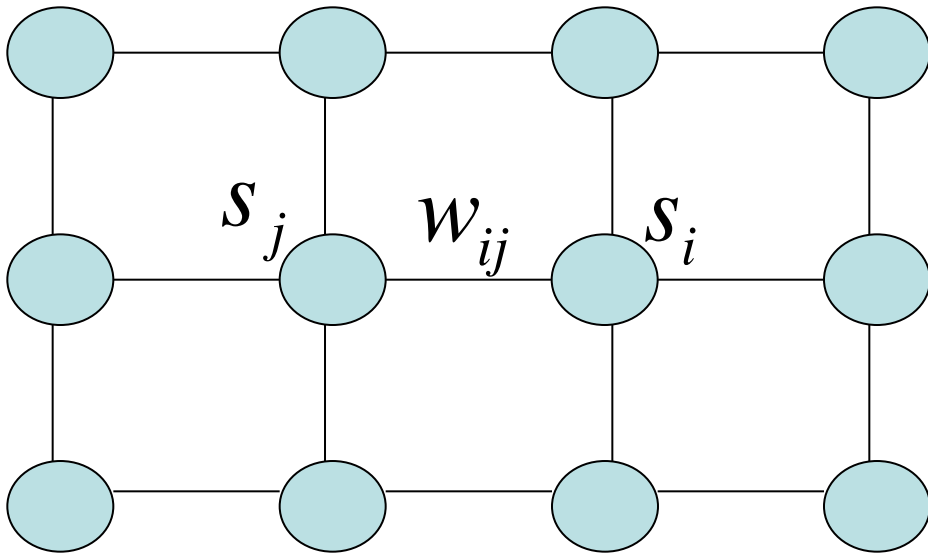
Fermi_Dirac distribution

- [Fermi_Dirac_statistics](#)

Logistic differential equation

- [Logistic function - Wikipedia, the free encyclopedia](#)

- Sigmoid differential equation



$$s_i \in \{-, +\}$$

$$v_i = \langle s_i \rangle \in [-, +]$$

$$s_i \in \{0,1\}$$

$$u_i = \langle s_i \rangle \in [0,1]$$

$$h_i = \sum_{j \in NB_i} w_{ij} v_j$$

logistic differential equation $\frac{du_i}{dh_i} = \beta u_i (1 - u_i)$

$$u_i = f(h_i) = \frac{1}{\exp(-\beta h_i) + 1}$$

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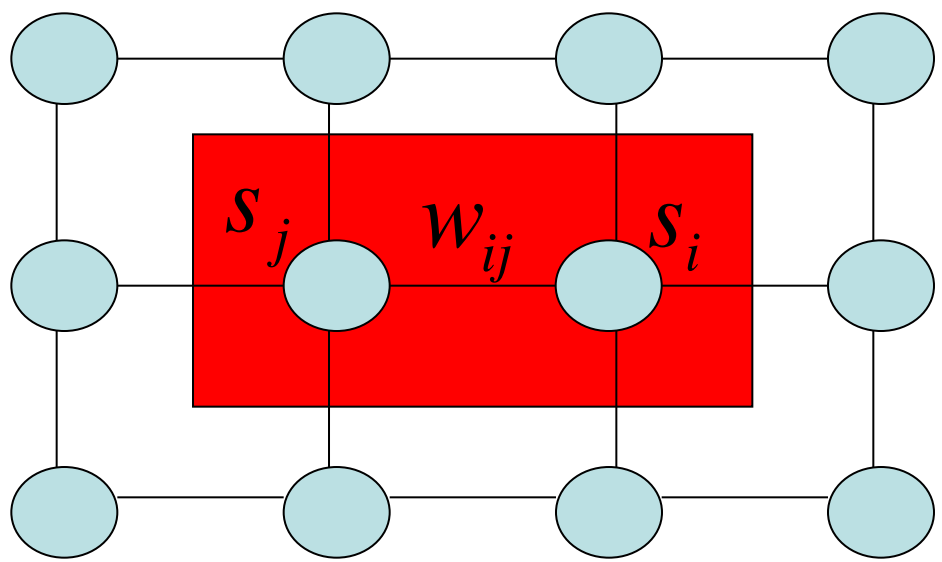
$$v_i = 2u_i - 1,$$

$$\frac{dv_i}{dh_i} = 2 \frac{du_i}{dh_i} = 2\beta \left(\frac{1+v_i}{2}\right) \left(\frac{1-v_i}{2}\right)$$

$$= \frac{1}{2} \beta (1+v_i)(1-v_i)$$

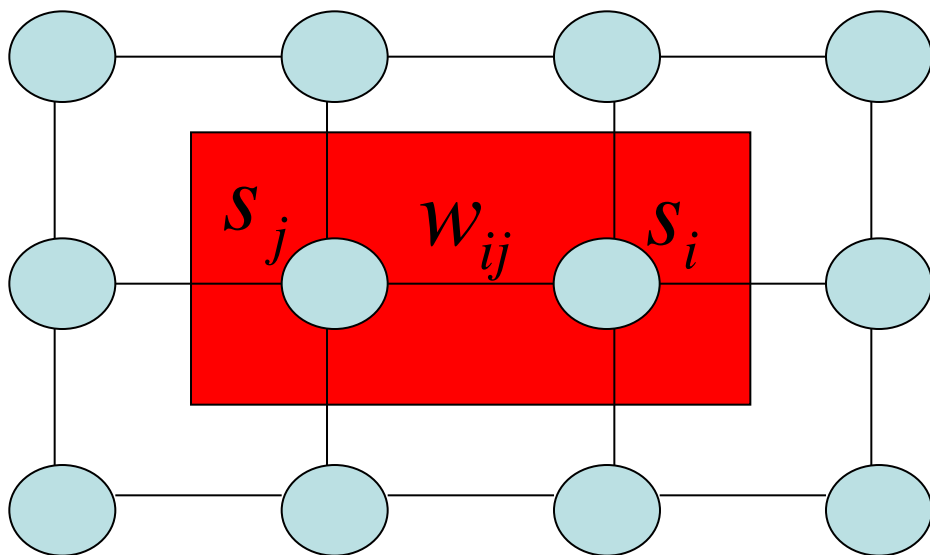
$$v_i = \tanh(\beta h_i)$$

- couple expectation



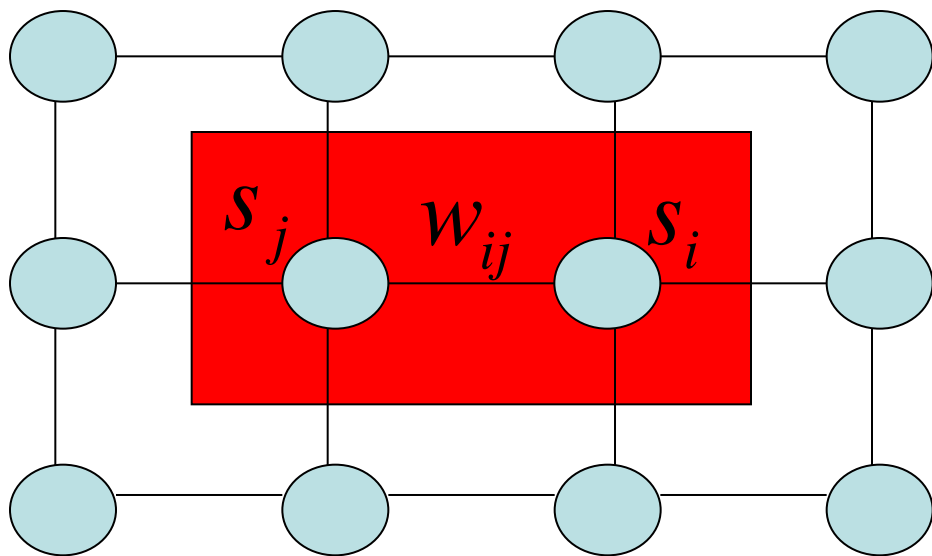
$$s_i \in \{-, +\}$$

$$q_{ij} = s_i s_j \in \{-, +\}$$



$$h_i = \sum_{\substack{k \in NB(i) \\ k \neq i}} w_{ik} v_k$$

$$h_j = \sum_{\substack{k \in NB(j) \\ k \neq j}} w_{jk} v_k$$



$$S_i = +, S_j = +$$

$$S_i = +, S_j = -$$

$$S_i = -, S_j = +$$

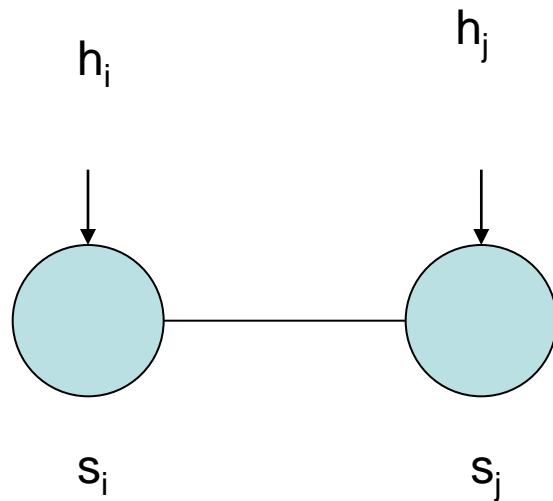
$$S_i = -, S_j = -$$

$$\varepsilon_{++} = h_i + h_j + w_{ij}$$

$$\varepsilon_{+-} = h_i - h_j - w_{ij}$$

$$\varepsilon_{-+} = -h_i + h_j - w_{ij}$$

$$\varepsilon_{--} = -h_i - h_j + w_{ij}$$



$$\Pr(s_i = +, s_j = +) \propto \exp(-\beta \varepsilon_{++})$$

$$\Pr(s_i = +, s_j = -) \propto \exp(-\beta \varepsilon_{+-})$$

$$\Pr(s_i = -, s_j = +) \propto \exp(-\beta \varepsilon_{-+})$$

$$\Pr(s_i = -, s_j = -) \propto \exp(-\beta \varepsilon_{--})$$

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$$\Pr(s_i = -, s_j = +) \propto \exp(-\beta \varepsilon_{-+})$$

$$\Pr(s_i = -, s_j = -) \propto \exp(-\beta \varepsilon_{--})$$

$$\Pr(q_{ij} = 1) = \Pr(s_i = +, s_j = +) + \Pr(s_i = -, s_j = -)$$

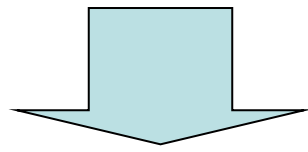
$$= \frac{\exp(-\beta \varepsilon_{++}) + \exp(-\beta \varepsilon_{--})}{\exp(-\beta \varepsilon_{++}) + \exp(-\beta \varepsilon_{+-}) + \exp(-\beta \varepsilon_{-+}) + \exp(-\beta \varepsilon_{--})}$$

$$\Pr(q_{ij} = -1) = \Pr(s_i = +, s_j = -) + \Pr(s_i = -, s_j = +)$$

$$= \frac{\exp(-\beta \varepsilon_{+-}) + \exp(-\beta \varepsilon_{-+})}{\exp(-\beta \varepsilon_{++}) + \exp(-\beta \varepsilon_{+-}) + \exp(-\beta \varepsilon_{-+}) + \exp(-\beta \varepsilon_{--})}$$

$$\begin{aligned} \Pr(q_{ij} = 1) &= \Pr(s_i = +, s_j = +) + \Pr(s_i = -, s_j = -) \\ &= \frac{\exp(-\beta\varepsilon_{++}) + \exp(-\beta\varepsilon_{--})}{\exp(-\beta\varepsilon_{++}) + \exp(-\beta\varepsilon_{+-}) + \exp(-\beta\varepsilon_{-+}) + \exp(-\beta\varepsilon_{--})} \end{aligned}$$

$$\begin{aligned} \Pr(q_{ij} = -1) &= \Pr(s_i = +, s_j = -) + \Pr(s_i = -, s_j = +) \\ &= \frac{\exp(-\beta\varepsilon_{+-}) + \exp(-\beta\varepsilon_{-+})}{\exp(-\beta\varepsilon_{++}) + \exp(-\beta\varepsilon_{+-}) + \exp(-\beta\varepsilon_{-+}) + \exp(-\beta\varepsilon_{--})} \end{aligned}$$



$$\begin{aligned} \langle q_{ij} \rangle &= \Pr(q_{ij} = 1) - \Pr(q_{ij} = -1) \\ &= \frac{\exp(-\beta\varepsilon_{++}) - \exp(-\beta\varepsilon_{+-}) - \exp(-\beta\varepsilon_{-+}) + \exp(-\beta\varepsilon_{--})}{\exp(-\beta\varepsilon_{++}) + \exp(-\beta\varepsilon_{+-}) + \exp(-\beta\varepsilon_{-+}) + \exp(-\beta\varepsilon_{--})} \end{aligned}$$

- $$\text{KL}_i = \sum_{j \in \text{NB}(i)} p_{ij} \log \frac{p_{ij}}{p_i p_j}$$

Minimize KL_i to determine individual p_i

Given $\{v_i\}$, calculate $\langle s_i s_j \rangle$

Given $\langle s_i s_j \rangle$, determine $\langle s_i \rangle$ by minimizing KL_i

- $$\text{KL}_i = \sum_{j \in \text{NB}(i)} p_i p_j \log \frac{p_i p_j}{p_{ij}}$$

Minimize KL_i to determine individual p_i

Given $\{v_i\}$, calculate $\langle s_i s_j \rangle$

Given $\langle s_i s_j \rangle$, determine $\langle s_i \rangle$ by minimizing KL_i

$$\begin{aligned}
\text{KL}_{ij} &= \sum_{s_i, s_j} p_i(s_i) p_j(s_j) \log \frac{p_i(s_i) p_j(s_j)}{p_{ij}(s_i, s_j)} \\
&= \sum_{s_i, s_j} p_i(s_i) p_j(s_j) \{ \log p_i(s_i) + \log p_j(s_j) - \log p_{ij}(s_i, s_j) \} \\
&= \sum_{s_i} p_i(s_i) \log p_i(s_i) + \sum_{s_j} p_j(s_j) \log p_j(s_j) - \sum_{s_i, s_j} p_i(s_i) p_j(s_j) \log p_{ij}(s_i, s_j)
\end{aligned}$$

Minimize KL_i to determine individual p_i

Given $\{v_i\}$, calculate $\langle s_i s_j \rangle$

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Exercise I

$$y(0) = 1$$

$$\frac{dy}{dx} = x$$

The change of y with respect to x is proportional to x .

The initial value of y is one

$$y(x) = ?$$

Exercise II

$$y(0) = 1$$

$$\frac{dy}{dx} = x(1 - x)$$

The change of y with respect to x is proportional to $x(1 - x)$.

$$y(x) = ?$$

Exercise III

$$y(0) = 1$$

$$\frac{dy}{dx} = y$$

The change of y with respect to x is proportional to y .

$$y(x) = ?$$

Exercise IV

$$y(0) = 1$$

$$\frac{dy}{dx} = y(1 - y)$$

The change of y with respect to x is proportional to $y(1 - y)$.

$$y(x) = ?$$