

①. $\int e^{-c} m^2 dm \rightarrow \text{finite}$

②. $\int e^{+c} m^2 dm \rightarrow \infty$

① $\ln z = \frac{z^n - 1}{n}$

②. $\int \text{Tr}(\dots) dJ P(J) = \text{Tr}(\int P(J) \dots dJ)$ 问题: 实用性

③. $\sum_j \sum_i \leftrightarrow \sum_i (\sum_j)$

Hilbert space \Rightarrow Simple \star

Quenched disorder

Annealed disorder

$H = - \frac{J}{2N} \sum_{\langle ij \rangle} \sigma_i \sigma_j - h \sum_i \sigma_i$ 可解

2种解法. 各有启发
本质: 相变 $\sim \frac{1}{T-T_c}$
空间 $2^N \rightarrow$ 1d 积分 (中心极限定理)

1) $Z = \text{Tr} e^{\frac{\beta J}{2N} \sum_{\langle ij \rangle} \sigma_i \sigma_j + \beta h \sum_i \sigma_i}$
 $= \text{Tr} e^{\frac{\beta J}{2N} [\sum_i \sigma_i^2 - N] + \beta h \sum_i \sigma_i}$

$m = \sum_i \sigma_i$ 只是 m 唯一函数

$= e^{-\frac{1}{2} \beta J} \text{Tr} e^{\left(\frac{\beta J}{2N} m^2 + \beta h m\right)}$

P.S: $X = \frac{1}{N} \sum_i X_i$ 和 $m = \sum_i \sigma_i$
 X_i 是随机数 σ_i 也是随机数 $\sigma_i = \pm 1$

$\langle m \rangle = 0$

$\langle m^2 \rangle = \sum_{ij} \langle \sigma_i \sigma_j \rangle = \sum_i \langle \sigma_i^2 \rangle = N$

$Z = e^{-\frac{1}{2} \beta J} \int p(m) e^{\frac{\beta J}{2N} m^2 + \beta h m} dm$

$p(m) \sim e^{-Ax^2}$

Hubbard - Stronovich method:

$\sqrt{\frac{A}{\pi}} \int e^{-Ax^2} dx = 1$

$\sqrt{\frac{A}{\pi}} \int e^{-Ax^2 + Bx} dx = \sqrt{\frac{A}{\pi}} \int e^{-A(x^2 - Bx)} dx = e^{\frac{1}{4} AB^2}$

$\Leftrightarrow B = m \quad A = \frac{\beta J}{N}$

or $\beta^2 J^2 < 1 \Leftrightarrow k_B T > J$

Normal phase!

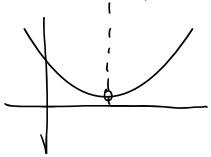
直观图像: $\bar{Z}^N = \text{Tr} \left(e^{\frac{\beta J^2}{2} \sum_{\alpha \neq \beta} (m^{\alpha\beta})^2} \right)$
 $\approx \prod_{\alpha \neq \beta} \text{Tr} e^{\frac{\beta J^2}{2} (m^{\alpha\beta})^2}$
 $= \prod_{\alpha \neq \beta} \int P(m^{\alpha\beta}) e^{\frac{\beta J^2}{2} (m^{\alpha\beta})^2}$
 $= \prod_{\alpha \neq \beta} \int \frac{1}{\sqrt{2\pi U}} e^{-\frac{(m^{\alpha\beta})^2}{2U} + \frac{\beta J^2}{2} (m^{\alpha\beta})^2} dm^{\alpha\beta}$

H-S方法:

$\text{Tr} [e^{\frac{\beta J^2}{2} (m^{\alpha\beta})^2}] \propto \int dq^{\alpha\beta} e^{-X(q^{\alpha\beta})^2 + Y q^{\alpha\beta} m^{\alpha\beta}}$

$\bar{Z}^N \propto \int Dq^{\alpha\beta} e^{-X \sum_{\alpha \neq \beta} (q^{\alpha\beta})^2}$
 $\propto \int Dq^{\alpha\beta} e^{-\beta F}$

Saddle point



$q^{\alpha\beta} = 0$
 $q^{\alpha\beta} \neq 0$

Interaction \Rightarrow 耦合
 $\text{Tr} \left(e^{\sum_{\alpha \neq \beta} Y q^{\alpha\beta} m^{\alpha\beta}} \right)$ N 个独立 site 可变成 $(\quad)^N$
 $= \left(\text{tr} \left[e^{\sum_{\alpha \neq \beta} Y q^{\alpha\beta} \sigma^{\alpha} \sigma^{\beta}} \right] \right)^N$
 $\propto [2 \cosh(Y q^{\alpha\beta})]^N$

★ Edwards *Theoret* chap 1:

从 single spin 出发

简单图像理解: $\sum_j J_{ij} \sigma_i \sigma_j = \sum_i \sigma_i \tilde{\sigma}_i \Rightarrow$ 每个 spin 感受到一个场 $\tilde{\sigma}_i$, 有 (一定) 很大独立性

$P(\sigma) = e^{\beta F - \beta H} = e^{\beta F - \beta \sum_j J_{ij} \sigma_i \sigma_j}$

联合概率: $S_i^{(1)} \mapsto t = \text{now} \Leftrightarrow$ 求 $\langle S_i^{(1)}, S_i^{(2)} \rangle = g \neq 0$
 $S_i^{(2)} \mapsto t = +\infty$

$P(S_i^{(1)}, S_i^{(2)}) = e^{2\beta F - \beta \sum_j J_{ij} (S_i^{(1)} S_j^{(1)} + S_i^{(2)} S_j^{(2)})}$

$= e^{2\beta F - \beta \sum_j [J_{ij}^{(1)} S_i^{(1)} S_j^{(1)} + J_{ij}^{(2)} S_i^{(2)} S_j^{(2)}]}$

求 $\langle S_i^{(1)} \cdot S_i^{(2)} \rangle = g \neq 0$ if $S_i^{(1)}, S_i^{(2)}$ 独立 $\langle S_i^{(1)} S_i^{(2)} \rangle = 0$

else $\neq 0$

$g = \langle S_i^{(2)} S_i^{(1)} \rangle \approx -N e^{-\mu P} S_i^{(1)} S_i^{(2)}$

$\frac{1}{P} = \frac{\cosh(Pg) - \frac{1}{Pg}}{(Pg)}$ $g = \langle S_i^{(2)} S_i^{(1)} \rangle = \int \mu e^{-\mu P g} d\mu / \int e^{-\mu P g} d\mu$