

多体  $\left\{ \begin{array}{l} \text{平均场 (spin, B-H, F-H, ...)} \Rightarrow \text{多体} \Rightarrow \text{单体+自洽} \\ \text{严格近似: } \left\{ \begin{array}{l} \text{spin} \\ \text{Bose} \\ \text{Fermi} \end{array} \right. \begin{array}{l} 1) \text{ 数 Hilbert 空间} \\ \text{给定守恒量} \end{array} \end{array} \right.$

MC  $\Rightarrow$  高维积分  $\rightarrow$  求矩阵元, 另类处理

技巧, 索引表  $|i\rangle \Leftrightarrow |n_1, \dots, n_L\rangle$



trick:  $L=7, N=5$ . Bose.

去掉不太可能出现在态, 使子空间减小.

例如 5 个粒子全在一个 site 上, 这种过于集中的态

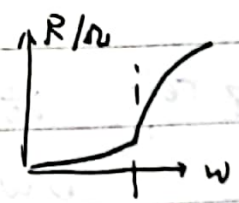
### Anderson Localization (AL)

#### 单体问题

Motivation: (1) 存在于各种材料中存在, 半导体.

AL P.W. Anderson 1957 年最早提出 Anderson Localization. 相变.

$\left. \begin{array}{l} W > W_c \text{ state localized.} \\ W < W_c \text{ extended.} \end{array} \right\}$  for 3D.



折后意味着波函数会传播, 往往意味着金属.

$\left\{ \begin{array}{l} \text{localized} \Rightarrow \text{insulator.} \end{array} \right.$

(2) 比较容易计算.

(3) 应用广泛.

Ref: (1) Hoffmann 计算物理. 转移矩阵一章.

(2) Thouless: J. Phys. C. 1970

(3) Abrahams, Anderson, Licciardella

Ramar Ramakrishnan PRL: 1979.

书 [3] 卷

(4) 李正华 《固体理论》 第12章

方法 = 转移矩阵

Ising model.

$$H = -J \sum_i \sigma_i - \sigma_{i+1} + B \sum_i \sigma_i$$

$$\sigma_i = \pm 1$$

↓  
磁场所

Lenz

$$H = -t \sum_i (c_i^\dagger c_{i+1} + \text{h.c.}) + V_i c_i^\dagger c_i$$

$$H_{N \times N} = \begin{pmatrix} 0 & V_1 & t & & & \\ t & V_2 & t & & & 0 \\ & & t & & & \\ & & & \ddots & & \\ 0 & & & t & & \\ & & & & \ddots & \end{pmatrix}$$

★ ref. 讨论  $V_i$  下的物理.

① white noise

②  $V_i = V \omega_i$   $\omega_i$  无理数.

③ correlated disorder.

1D  $\rightarrow$  推广 高维.

$$H = -t \sum_i (c_i^\dagger c_{i+1} + \text{h.c.}) + v_i c_i^\dagger c_i$$

1d.

$$v_i \neq 0 \text{ white noise } \left\{ \begin{array}{l} \langle v_i \rangle = 0 \\ \langle v_i v_j \rangle = \sigma^2 \delta_{ij} \end{array} \right.$$

投空位  $N=1$ , one particle

$$|k\rangle = \left\{ |i\rangle \right\} \equiv \left\{ |c_i^\dagger |0\rangle \right\} \quad \forall i \in \mathbb{Z}$$

$$= |100 \dots 01 \dots 0\rangle$$

$$|n\rangle = \frac{1}{\sqrt{n!}} (\alpha^\dagger)^n |0\rangle \quad [x \frac{12}{5}]$$

$$\psi = \sum_i \alpha_i |i\rangle = \sum_i \alpha_i c_i^\dagger |0\rangle$$

$$\left( \begin{array}{l} i \dot{c}_i = [c_i, H] = \left( \frac{\partial H}{\partial c_i^\dagger} \right) \\ = -t (c_{i+1} + c_{i-1}) + v_i c_i \\ \text{use } [a, a^\dagger] = 1 \quad \text{or } [a, f] = \left( \frac{\partial f}{\partial a^\dagger} \right) \end{array} \right)$$

$$E \alpha_i = -t (\alpha_{i+1} + \alpha_{i-1}) + v_i \alpha_i$$

$$E \psi = H \psi \quad E \alpha_i |i\rangle = \alpha_j (H |j\rangle)$$

$$E \alpha_i = \alpha_j \langle i | H | j \rangle$$

$$= v_i \alpha_i - t (\alpha_{j+1} + \alpha_{j-1})$$

IF  $v_j = v = \text{const}$ 

$$\alpha_{j+1} = \alpha_{j-1} + \left( \frac{v-E}{t} \right) \alpha_j$$

$$\Rightarrow x^2 = 1 + \frac{v-E}{t} x \quad \Rightarrow \text{递推关系}$$

当  $V_i$  为随机数:

$$\alpha_{i+1} = -\alpha_{i-1} + (V_i - \bar{v}) \alpha_i \quad (k=1)$$

$$\begin{pmatrix} \alpha_{i+1} \\ \alpha_i \end{pmatrix} = \begin{pmatrix} V_i - \bar{v} & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \alpha_i \\ \alpha_{i-1} \end{pmatrix} = \hat{T}_i X_i$$

$$\Rightarrow X_{i+1} = \hat{T}_i X_i = \hat{T}_i \hat{T}_{i-1} X_{i-1}$$

$$\dots = \hat{T}_i \hat{T}_{i-1} \dots \hat{T}_1 X_1$$

$$X_{N+1} = \hat{T}_N \hat{T}_{N-1} \dots \hat{T}_1 X_1 \sim e^{-N/4}$$

QR 分解:  $R =$  上三角阵

$R_1, R_2 \Rightarrow$  上三角

$$\begin{pmatrix} a & b \\ 0 & c \end{pmatrix} \begin{pmatrix} x & y \\ 0 & z \end{pmatrix} = \begin{pmatrix} ax & ay + bz \\ 0 & cz \end{pmatrix}$$

$$\hat{T}_1 X_1$$



$$\hat{T}_2 Q_1 R_1 X_1$$



$$\hat{T}_3 Q_2 R_2 R_1 X_1$$



$$Q_3 R_3 R_2 R_1 X_1$$



$$\hat{T}_N \hat{T}_{N-1} \dots \hat{T}_1 X_1$$

$$Q_N R_N R_{N-1} \dots R_1 X_1$$



求  $X_{n+1}^T X_{n+1}$

$$= (X_1)^T R_1 + R_2 + \dots + R_n + Q_n^T Q_n (R_n \dots R_1 (X_1))$$

$$= (X_1)^T R_1 + R_2 + \dots + R_n + R_n \dots R_1 (X_1) \sim e^{-n/\gamma}$$

$$(R_3 R_2 R_1)_{kk} = R_3^{kk} R_2^{kk} R_1^{kk}$$

(0) (0)

$$X_{n+1}^T X_{n+1} \sim e^{-2(n+1)/\gamma} = f(n) e^{-2(n+1)/\gamma}$$

$$\ln f(n) - \frac{2(n+1)}{\gamma} = \ln (X_1^T + R^T R X_1)$$

$$\Rightarrow \frac{\ln f(n)}{n+1} = + \frac{2}{\gamma} + \frac{1}{n+1} (X_1^T + R^T R X_1)$$

$$\frac{2}{\gamma} = -\lim_{n \rightarrow \infty} \frac{1}{n+1} \ln (X_1^T + R^T R X_1)$$

号关联长度

定理: Useledac 通, 历史: (动力学系统)

$$T = \lim_{n \rightarrow \infty} (T_n T_{n+1})^{1/2n} \text{ 存在, } \Rightarrow \text{ 号存在}$$

极限存在 号存在

上三角阵的本征值:

$$\det \left[ \begin{pmatrix} a & b & c \\ 0 & d & e \\ 0 & 0 & f \end{pmatrix} - X \right] = (a-X)(d-X)(f-X) = 0$$

上三角阵的本征值是主对角元

$$R = R_N \cdots R_1$$

$$\lambda_{kk} = (R_N)_{kk} (R_{N-1})_{kk} \cdots (R_1)_{kk}$$

$$(TT^+) = \rho^+ \lambda^+ \rho \rho^+ \lambda \rho$$

$$= \rho^+ \lambda^+ \lambda \rho$$

$$(\lambda^+ \lambda)_{kk} = \prod_j (R_j)_{kk} (R_j^+)_{kk}$$

$$= \prod_j |(R_j)_{kk}|^2$$

$$f(N) e^{-\frac{2N}{g_k}} = \prod_{j=1}^N |(R_j)_{kk}|^2$$

$$\Rightarrow \frac{1}{g_k} = \lim_{N \rightarrow \infty} \frac{1}{2N} \ln |(R_j)_{kk}|^2$$

$$\frac{\ln f(N)}{2N} - \frac{1}{g_k} = \left( \frac{1}{2N} \sum_{j=1}^N \ln |(R_j)_{kk}|^2 \right)$$

$$\Rightarrow \frac{1}{g_k} = - \lim_{N \rightarrow \infty} \frac{1}{2N} \sum_j \ln |(R_j)_{kk}|^2$$

取最大的  $g_k$  (或最小的  $\lambda_k$ )