

# 2025-2-28 与自旋压缩态相关的一些背景

## 压缩系数

$$\xi_S^2 = \frac{\min \Delta J_{\vec{n}_\perp}^2}{j/2} = \frac{4 \min \Delta J_{\vec{n}_\perp}^2}{N}$$

定义平均自旋方向  $\vec{n}_0 = \frac{\langle \vec{J} \rangle}{|\langle \vec{J} \rangle|}$

与平均自旋垂直的方向  $\vec{n}_1, \vec{n}_2, \vec{n}_\perp = \vec{n}_1 \cos \phi + \vec{n}_2 \sin \phi$

垂直方向  $\vec{n}_\perp$  的自旋算符  $J_{\vec{n}_\perp} = J_{\vec{n}_1} \cos \phi + J_{\vec{n}_2} \sin \phi$

垂直方向  $\vec{n}_\perp$  的自旋涨落  $(\Delta J_{\vec{n}_\perp})^2 = \langle J_{\vec{n}_\perp}^2 \rangle = \vec{n}_\perp \Gamma \vec{n}_\perp^T$

协方差矩阵  $\Gamma$

$$\Gamma = \begin{bmatrix} \langle J_{\vec{n}_1}^2 \rangle & \text{Cov}(J_{\vec{n}_1}, J_{\vec{n}_2}) \\ \text{Cov}(J_{\vec{n}_1}, J_{\vec{n}_2}) & \langle J_{\vec{n}_2}^2 \rangle \end{bmatrix}, \quad \text{Cov}(J_{\vec{n}_1}, J_{\vec{n}_2}) = \frac{1}{2} \langle [J_{\vec{n}_1}, J_{\vec{n}_2}]_+ \rangle - \langle J_{\vec{n}_1} \rangle \langle J_{\vec{n}_2} \rangle = \frac{1}{2} \langle [J_{\vec{n}_1}, J_{\vec{n}_2}]_+ \rangle$$

协方差矩阵的最小本征值即为垂直方向  $\vec{n}_\perp$  的最小自旋涨落

$$\min(\Delta J_{\vec{n}_\perp})^2 = \lambda_-$$

$$\xi_S^2 = \frac{4}{N} = \frac{2}{N} \left[ \langle J_{\vec{n}_1}^2 + J_{\vec{n}_2}^2 \rangle - \sqrt{(\langle J_{\vec{n}_1}^2 - J_{\vec{n}_2}^2 \rangle)^2 + 4 \text{Cov}(J_{\vec{n}_1}, J_{\vec{n}_2})} \right]$$

对于自旋相干态CSS,  $\xi_S^2 = 1$

$$\xi_R^2 = \frac{(\Delta \phi)^2}{(\Delta \phi)_{CSS}^2} = \frac{N(\Delta J_{\vec{n}_\perp})^2}{|\langle \vec{J} \rangle|^2}$$

与平均自旋垂直的方向  $\vec{n}_1, \vec{n}_2, \quad \vec{n}_1 \times \vec{n}_2 = \vec{n}_0$

绕  $\vec{n}_1$  方向转动  $\phi$  角度, 沿  $\vec{n}_2$  方向的自旋算符

$$J_{\vec{n}_2}^{out} = \exp(i\phi J_{\vec{n}_1}) J_{\vec{n}_2} \exp(-i\phi J_{\vec{n}_1}) = \cos \phi J_{\vec{n}_2} - \sin \phi J_{\vec{n}_0}$$

$$\langle J_{\vec{n}_2}^{out} \rangle = -\sin \phi \langle J_{\vec{n}_0} \rangle$$

$$(\Delta J_{\vec{n}_2}^{out})^2 = \cos^2 \phi (\Delta J_{\vec{n}_2})^2 + \sin^2 \phi (\Delta J_{\vec{n}_0})^2 - \frac{1}{2} \sin(2\phi) \langle [J_{\vec{n}_2}, J_{\vec{n}_0}]_+ \rangle$$

$$\text{转动角度趋于0时的相位敏感度 } \Delta \phi = \frac{\Delta J_{\vec{n}_2}}{|\partial \langle J_{\vec{n}_2}^{out} \rangle / \partial \phi|} = \frac{\Delta J_{\vec{n}_2}}{\langle J_{\vec{n}_0} \rangle} \left( = \frac{\Delta J_{\vec{n}_\perp}}{|\langle \vec{J} \rangle|} \right)$$

$$\text{自旋相干态的相位敏感度 } (\Delta \phi)_{CSS} = \frac{1}{\sqrt{N}}$$

$$\text{当垂直方向 } \vec{n}_\perp \text{ 选为最小自旋涨落方向时 } \xi_R^2 = \left( \frac{j}{\langle \vec{J} \rangle} \right)^2 \xi_S^2 \geq \xi_S^2$$

## QFI (Quantum Fisher Information)

### 定义与性质

#### 条件概率分布

待测参数为 $\theta$ ，算符 $E(\epsilon)$ 对应的测量值为 $\epsilon$ 的条件概率分布 $P(\epsilon|\theta) = \text{Tr}(\rho(\theta)E(\epsilon))$   
 $m$ 个独立子系统测量结果 $\epsilon = (\epsilon_1, \dots, \epsilon_m)$

## 对数似然函数

$$L(\epsilon|\theta) = \ln P(\epsilon|\theta) = \sum_i \ln P(\epsilon_i|\theta)$$

## 估计子 $\Theta(\epsilon)$

从测量结果 $\epsilon$ 得到 $\theta$ 的一个估计值 $\Theta$ ，相应也有一个分布 $P(\Theta|\theta)$

$$\text{估计子均值} \langle \Theta \rangle = \sum_{\epsilon} \Theta(\epsilon) P(\epsilon|\theta)$$

$$\text{估计子涨落} (\Delta \Theta)^2 = \sum_{\epsilon} P(\epsilon|\theta) (\Theta(\epsilon) - \langle \Theta \rangle)^2$$

## 无偏估计

$$\langle \Theta \rangle = \theta, \forall \theta$$

$$\frac{\partial \langle \Theta \rangle}{\partial \theta} = 1$$

## 极大似然估计子

$$\Theta_{MLE}(\theta) = \arg[\max_{\theta} P(\epsilon|\theta)]$$

## Fisher信息

$$F(\theta) = \left\langle \left( \frac{\partial L(\theta)}{\partial \theta} \right)^2 \right\rangle = \sum_{\epsilon} \frac{1}{P(\epsilon|\theta)} \left( \frac{\partial P(\epsilon|\theta)}{\partial \theta} \right)^2$$

可加性

$$\begin{aligned} F(\theta) &= \left\langle \sum_i \frac{1}{P(\epsilon_i)} \frac{\partial P(\epsilon_i|\theta)}{\partial \theta} \sum_j \frac{1}{P(\epsilon_j)} \frac{\partial P(\epsilon_j|\theta)}{\partial \theta} \right\rangle \\ &= \sum_i \sum_{\epsilon_i} \frac{1}{P(\epsilon_i|\theta)} \left( \frac{\partial P(\epsilon_i|\theta)}{\partial \theta} \right)^2 + \sum_i \sum_{\epsilon_i} \frac{\partial P(\epsilon_i|\theta)}{\partial \theta} \sum_j \sum_{\epsilon_j} \frac{\partial P(\epsilon_j|\theta)}{\partial \theta} \\ &= \sum_i \sum_{\epsilon_i} \frac{1}{P(\epsilon_i|\theta)} \left( \frac{\partial P(\epsilon_i|\theta)}{\partial \theta} \right)^2 + 0 = \sum_i F_i(\theta) \end{aligned}$$

$$m \text{ 个全同子系统 } F(\theta) = m F_{single}(\theta)$$

## Cramer-Rao 下界

$$(\Delta \Theta)^2 \geq (\Delta \Theta_{CR})^2 = \frac{(\frac{\partial \langle \Theta \rangle}{\partial \theta})^2}{F(\theta)}$$

## 测量样本数 $m$ 趋于无穷的极大似然估计

$$P(\Theta_{MLE}|\theta) = \sqrt{\frac{mF(\theta)}{2\pi}} e^{-\frac{mF(\theta)}{2}(\theta - \Theta_{MLE})^2}, \quad m \rightarrow \infty$$

$$\text{渐近无偏 } \frac{\partial \langle \Theta_{MLE} \rangle}{\partial \theta} = 1, \quad m \rightarrow \infty$$

$$\text{方差可以达到CR下界 } (\Delta \Theta_{MLE})^2 = \frac{1}{mF(\theta)}, \quad m \rightarrow \infty$$

## 量子Fisher信息

改变测量算符 $E(\epsilon)$ 所得的Fisher信息最大值

$$F_Q[\rho(\theta)] = \max_{E(\epsilon)} F[\rho(\theta), \{E(\epsilon)\}] = \text{Tr}(\rho(\theta)L_\theta^2)$$

可加性 $F_Q[\rho(\theta)] = \sum_i F_Q[\rho_i(\theta)]$

对称对数导数 $L(\theta)$

$$\frac{\partial \rho}{\partial \theta} = \frac{1}{2}(\rho(\theta)L_\theta + L_\theta \rho(\theta))$$

显式形式

$$\begin{aligned} \rho(\theta) &= \sum_k p_k |k\rangle \langle k|, \quad \langle k|k'\rangle = \delta_{k'k} \\ L_\theta &= \sum_k \frac{\partial_\theta p_k}{p_k} |k\rangle \langle k| + 2 \sum_{k,k'} \frac{p_k - p_{k'}}{p_k + p_{k'}} |k\rangle \langle \partial_\theta k|k'\rangle \langle k'| \\ F_Q[\rho(\theta)] &= \sum_k \frac{(\partial_\theta p_k)^2}{p_k} + 2 \sum_{k,k'} \frac{(p_k - p_{k'})^2}{p_k + p_{k'}} |\langle \partial_\theta k|k'\rangle|^2 \end{aligned}$$

么正演化下的显式形式

$$\begin{aligned} \rho(\theta) &= e^{-iH\theta} \rho_0 e^{iH\theta} \\ L_\theta &= 2i \sum_{k,k'} \frac{p_k - p_{k'}}{p_k + p_{k'}} |k\rangle \langle k|H|k'\rangle \langle k'| \\ F_Q[\rho(\theta)] &= 2 \sum_{k,k'} \frac{(p_k - p_{k'})^2}{p_k + p_{k'}} |\langle k|H|k'\rangle|^2 \end{aligned}$$

纯态情形

$$\begin{aligned} L_\theta &= 2i|\psi(\theta)\rangle \langle \psi(\theta)|H - 2iH|\psi(\theta)\rangle \langle \psi(\theta)| \\ F_Q[\rho(\theta)] &= 4(\Delta H)^2 \end{aligned}$$

与压缩系数的关系

结合压缩系数 $\xi_R^2$ 的定义,  $H = J_{\vec{n}_1}$

$$\begin{aligned} F_Q[\rho(\theta)] &= 4(\Delta J_{\vec{n}_1})^2 \\ \text{由不确定关系 } (\Delta J_{\vec{n}_1})^2 (\Delta J_{\vec{n}_2})^2 &\geq \frac{1}{4} \langle J_{\vec{n}_0} \rangle^2 \\ \frac{N}{F_Q[\rho(\theta)]} &= \frac{N}{4(\Delta J_{\vec{n}_1})^2} \leq \frac{N(\Delta J_{\vec{n}_2})^2}{\langle J_{\vec{n}_0} \rangle^2} = \xi_R^2 \end{aligned}$$

自旋演化半经典数值方法

TWA (truncated Wigner approximation)

truncated Wigner approximation

位置动量空间

相点算符

$$\langle x'|\hat{A}(x,q)|x''\rangle = \delta\left(x - \frac{x'+x''}{2}\right)e^{ip\cdot(x'-x'')/\hbar}, \quad \langle p'|\hat{A}(x,p)|p''\rangle = \delta\left(p - \frac{p'+p''}{2}\right)e^{-ix\cdot(p'-p'')/\hbar}$$

外尔符号

$$O_W(x, p) = \text{Tr}[\hat{O}\hat{A}(x, p)] = \int d^D\xi \langle x - \frac{\xi}{2} | \hat{O}(\hat{x}, \hat{p}) | x + \frac{\xi}{2} \rangle \exp \left[ \frac{i}{\hbar} p \cdot \xi \right]$$

$$X_W(x, p) = x, P_W(x, p) = p$$

简单形式的哈密顿量  $\hat{H}(\hat{X}, \hat{P}) = T(\hat{P}) + V(\hat{X})$ ,  $H_W(x, p) = T(p) + X(x)$  为经典哈密顿量

## Wigner函数

$$W(x, p) = \text{Tr}[\hat{\rho}\hat{A}(x, p)] = \rho_W(x, p) = \int d^D\xi \langle x - \frac{\xi}{2} | \hat{\rho} | x + \frac{\xi}{2} \rangle \exp \left[ \frac{i}{\hbar} p \cdot \xi \right], \quad \int \int \frac{d^Dx d^Dp}{(2\pi\hbar)^D} W(x, p) = 1$$

## 算符期望

$$\text{满足 } \langle \hat{O}(\hat{x}, \hat{p}) \rangle = \int \int \frac{d^Dx d^Dp}{(2\pi\hbar)^D} W(x, p) O_W(x, p)$$

## Wigner函数的演化

$$\text{辛算符 } \Lambda = \overleftarrow{\frac{\partial}{\partial p}} \overrightarrow{\frac{\partial}{\partial x}} - \overleftarrow{\frac{\partial}{\partial x}} \overrightarrow{\frac{\partial}{\partial p}}$$

$$\text{泊松括号 } \{A, B\}_{PB} = -A\Lambda B$$

$$\text{Moyal括号 } \{A, B\}_{MB} = -\frac{2}{\hbar} A \sin \left[ \frac{\hbar}{2} \Lambda \right] B = -A\Lambda B + O(\hbar^2) = \{A, B\}_{PB} + O(\hbar^2)$$

$$\text{算符乘积的外尔符号 } (\hat{A}\hat{B})_W(x, p) = A_W(x, p) \exp \left[ -\frac{i\hbar}{2} \Lambda \right] B_W(x, p)$$

$$\text{对易子的外尔符号 } ([\hat{A}, \hat{B}])_W = -2iA_W \sin \left[ \frac{\hbar}{2} \Lambda \right] B_W = i\hbar \{A_W, B_W\}_{MB}$$

$$\text{密度算符演化方程 } \dot{\hat{\rho}}(t) = \frac{1}{i\hbar} [\hat{H}, \hat{\rho}(t)]$$

$$\text{Wigner函数的演化方程 } \frac{\partial}{\partial t} W(x, p, t) = \{H_W(x, p), W(x, p, t)\}_{PB} + O(\hbar^2)$$

## t时刻的期望

对于经典哈密顿量  $H_W(x, p)$  决定的经典轨迹  $x_{cl}(t), p_{cl}(t)$ , 维格纳函数近似满足刘维尔定理

$$\frac{d}{dt} W(x_{cl}(t), p_{cl}(t), t) = O(\hbar^2)$$

t时刻算符期望

$$\langle \hat{O}(\hat{x}, \hat{p}) \rangle(t) = \int \int \frac{d^Dx_{cl}(t) d^Dp_{cl}(t)}{(2\pi\hbar)^D} W(x_{cl}(t), p_{cl}(t)) O_W(x_{cl}(t), p_{cl}(t)) \approx \int \int \frac{d^Dx_0 d^Dp_0}{(2\pi\hbar)^D} W(x_0, p_0) O_W(x_{cl}(t), p_{cl}(t))$$

## 相干态空间

### 外尔符号

$$O_W(\psi, \psi^*) = \int \int \frac{d\eta^* d\eta}{(2\pi)^M} \langle \psi - \frac{\eta}{2} | \hat{O}(\hat{\psi}, \hat{\psi}^\dagger) | \psi + \frac{\eta}{2} \rangle \exp \left[ \frac{1}{2} (\eta^* \psi - \eta \psi^*) \right]$$

在变换  $x = \sqrt{\frac{\hbar}{2}}(\psi + \psi^*), p = i\sqrt{\frac{\hbar}{2}}(\psi^* - \psi)$  下可以回到位置动量空间的形式

## Wigner函数

$$W(\psi, \psi^*) = \int \int \frac{d\eta^* d\eta}{(2\pi)^M} \langle \psi - \frac{\eta}{2} | \hat{\rho} | \psi + \frac{\eta}{2} \rangle \exp \left[ \frac{1}{2} (\eta^* \psi - \eta \psi^*) \right]$$

## 算符期望

$$\langle \hat{O}(\hat{\psi}, \hat{\psi}^\dagger) \rangle = \int \int \frac{d\psi d\psi^*}{\pi^M} W(\psi, \psi^*) O_W(\psi, \psi^*)$$

## Wigner函数的演化

$$\text{辛算符 } \Lambda_c = \overleftarrow{\frac{\partial}{\partial \psi}} \overrightarrow{\frac{\partial}{\partial \psi^*}} - \overleftarrow{\frac{\partial}{\partial \psi^*}} \overrightarrow{\frac{\partial}{\partial \psi}}$$

$$\text{泊松括号 } \{A, B\}_{PB} = A \Lambda_c B$$

$$\text{Moyal括号 } \{A, B\}_{MB} = 2A \sinh \left[ \frac{\Lambda_c}{2} \right] B = A \Lambda_c B + \dots = \{A, B\}_{PB} + \dots$$

$$\text{算符乘积的外尔符号 } (\hat{A}\hat{B})_W(\psi, \psi^*) = A_W(\psi, \psi^*) \exp \left[ \frac{\Lambda_c}{2} \right] B_W(\psi, \psi^*)$$

$$\text{对易子的外尔符号 } ([\hat{A}, \hat{B}])_W = 2A_W \sinh \left[ \frac{\Lambda_c}{2} \right] B_W = \{A_W, B_W\}_{MB}$$

$$\text{密度算符演化方程 } \dot{\hat{\rho}}(t) = \frac{1}{i\hbar} [\hat{H}, \hat{\rho}(t)]$$

$$\text{Wigner函数的演化方程 } \frac{\partial}{\partial t} W(\psi, \psi^*, t) = \{H_W(\psi, \psi^*), W(\psi, \psi^*, t)\}_{PB} + \dots$$

$\psi, \psi^*$ 很大时，忽略泊松括号后面的项，因为 $\psi, \psi^*$ 次数低

## t时刻的期望

$$\langle \hat{O}(\hat{\psi}, \hat{\psi}^\dagger) \rangle(t) = \int \int \frac{d\psi_0 d\psi_0^*}{\pi^M} W(\psi_0, \psi_0^*) O_W(\psi_{cl}(t), \psi_{cl}^*(t))$$

## 自旋

### Schwinger表示

$$\hat{s}^z = \frac{\hat{\alpha}^\dagger \hat{\alpha} - \hat{\beta}^\dagger \hat{\beta}}{2}, \quad \hat{s}^+ = \hat{\alpha}^\dagger \hat{\beta}, \quad \hat{s}^- = \hat{\beta}^\dagger \hat{\alpha}$$

利用相干态空间的外尔符号即可得到自旋算符的外尔符号

$$s_W^z = \frac{|\alpha|^2 - |\beta|^2}{2}, \quad s_W^+ = \alpha^* \beta, \quad s_W^- = \beta^* \alpha$$

$$s_W^x = \frac{\alpha^* \beta + \beta^* \alpha}{2}, \quad s_W^y = \frac{\alpha^* \beta - \beta^* \alpha}{2i}, \quad s_W^\perp \stackrel{\text{def}}{=} \sqrt{(s_W^x)^2 + (s_W^y)^2} = |\alpha| |\beta|$$

$$s_W \stackrel{\text{def}}{=} \sqrt{(s_W^z)^2 + (s_W^\perp)^2} = \frac{|\alpha|^2 + |\beta|^2}{2}$$

$$s_W^z + s_W = |\alpha|^2$$

### $|S, S\rangle$ 的Wigner函数

$$|S, S\rangle = \frac{(\alpha^\dagger)^{2S}}{\sqrt{(2S)!}} |0\rangle$$

Wigner函数

$$W(\alpha, \alpha^*, \beta, \beta^*) = 2e^{-2|\alpha|^2 - 2|\beta|^2} L_{2S+1}(4|\alpha|^2)$$

将变量选为 $s_W^z, s_W^\perp$

$$W(s_W^z, s_W^\perp) = \frac{2}{\pi s_W} e^{-4s_W} L_{2S+1}(4[s_W^z + s_W])$$

(分母中的 $s_W$ 应该来源于积分变量代换  $\left| \frac{\partial(s_W^z, s_W^\perp)}{\partial(|\alpha|^2, |\beta|^2)} \right| = s_W$ )

$$\int_{-\infty}^{\infty} ds_W^z \int_0^{\infty} ds_W^\perp \int_0^{2\pi} d\phi W(s_W^z, s_W^\perp) = 1$$

$$\text{大}S\text{极限下 } W(s_W^z, s_W^\perp) \approx \frac{1}{\pi S} e^{-(s_W^\perp)^2/S} \delta(s_W^z - S)$$

涨落为高斯分布，只能考虑集体自旋的涨落

## DTWA (Discrete TWA)

discrete truncated Wigner approximation

## 离散相空间

$$\alpha = (q, p) \in \{(0, 0), (0, 1), (1, 0), (1, 1)\}$$

$$\vec{r}_{(0,0)} = (1, 1, 1), \vec{r}_{(0,1)} = (-1, -1, 1), \vec{r}_{(1,0)} = (1, -1, -1), \vec{r}_{(1,1)} = (-1, 1, -1)$$

## 相点算符

$$\hat{A}_\alpha = \hat{\rho}(\vec{r}_\alpha) = (1 + \vec{r}_\alpha \cdot \vec{\sigma})/2$$

$$\text{Tr}(\hat{A}_\alpha \hat{A}_\beta) = \delta_{\alpha\beta}$$

$$N\text{个自旋的相点算符 } \hat{A}_{\vec{\alpha}} = \prod_{i=1}^N \hat{A}_{\alpha_i}$$

## 外尔符号

$$O_\alpha^W = \text{Tr}(\hat{O} \hat{A}_\alpha)/2, \quad \hat{O} = \sum_\alpha \hat{A}_\alpha O_\alpha^W$$

## Wigner函数

$$w_\alpha = \text{Tr}(\hat{\rho} \hat{A}_\alpha)/2, \quad \hat{\rho} = \sum_\alpha \hat{A}_\alpha w_\alpha, \quad \sum_\alpha w_\alpha = 1$$

$$p_1^x = w_{(0,0)} + w_{(1,0)}, \quad p_{-1}^x = w_{(0,1)} + w_{(1,1)}$$

$$p_1^y = w_{(0,0)} + w_{(1,1)}, \quad p_{-1}^y = w_{(0,1)} + w_{(1,0)}$$

$$p_1^z = w_{(0,0)} + w_{(0,1)}, \quad p_{-1}^z = w_{(1,0)} + w_{(1,1)}$$

$$\text{直积态的Wigner函数 } w_{\vec{\alpha}} = \prod_{i=1}^N w_{\alpha_i}$$

## 算符期望

$$\langle O \rangle = \sum_{\vec{\alpha}} w_{\vec{\alpha}} O_{\vec{\alpha}}^W$$

## t时刻的算符期望

$$\langle O \rangle(t) = \sum_{\vec{\alpha}} w_{\vec{\alpha}}(0) O_{\vec{\alpha}}^W(t)$$

$$O_{\vec{\alpha}}^W(t) = \text{Tr}(\hat{O}(t) \hat{A}_{\vec{\alpha}})/2 = \text{Tr}(\hat{O}(0) \hat{U}(t) \hat{A}_{\vec{\alpha}} \hat{U}^\dagger(t))/2, \quad U(t) = \exp(-i\hat{H}t/\hbar)$$

$$\hat{A}_{\vec{\alpha}}(t) = \hat{U}(t) \hat{A}_{\vec{\alpha}} \hat{U}^\dagger(t)$$

$$i\dot{\hat{A}}_{\vec{\alpha}}(t) = [\hat{H}, \hat{A}_{\vec{\alpha}}]$$

约化相点算符

$$\hat{A}_j = \text{Tr}_{\not j}(\hat{A}_{\vec{\alpha}}), \quad \hat{A}_{j,k} = \text{Tr}_{\not j, \not k}(\hat{A}_{\vec{\alpha}}), \quad \hat{A}_{j,k,l} = \text{Tr}_{\not j, \not k, \not l}(\hat{A}_{\vec{\alpha}})$$

包含单体作用和两体作用的哈密顿量

$$\hat{H} = \sum_j \hat{H}_j + \sum_{j,l} \hat{H}_{j,l} = \frac{1}{2} \sum_{i=1}^N \vec{\Omega}_i \cdot \vec{\sigma}_i + \frac{1}{2} \sum_{i \neq j}^N \vec{\sigma}_i^T \cdot \mathbf{J}_{ij} \cdot \vec{\sigma}_i$$

BBGKY级联方程

$$i\dot{\hat{A}}_j = [\hat{H}_j, \hat{A}_j] + \sum_{l=1, l \neq j}^N \text{Tr}_l([\hat{H}_{j,l}, \hat{A}_{j,l}])$$

$$i\dot{\hat{A}}_{j,k} = [\hat{H}_j + \hat{H}_k + \hat{H}_{j,k}, \hat{A}_{j,k}] + \sum_{l=1, l \neq j, k}^N \text{Tr}_l([\hat{H}_{j,l} + \hat{H}_{k,l}, \hat{A}_{j,k,l}])$$

截断关联

$$\hat{A}_{j,k} \approx \hat{A}_j \hat{A}_k$$

$$i\dot{\hat{A}}_j = [\hat{H}_j, \hat{A}_j] + \sum_{l=1, l \neq j}^N \text{Tr}_l([\hat{H}_{j,l}, \hat{A}_j \hat{A}_l])$$

等价于让 $\hat{A}_{\vec{\alpha}}$ 始终保持直积态得到的方程，求解规模得以缩小  
对于每个相空间抽样 $\vec{\alpha}$ ，演化方程可以写为经典自旋 $\vec{r}_{\vec{\alpha}}$ 的演化

$$\frac{d\vec{r}_i}{dt} = \vec{\Omega}_{eff}^i \times \vec{r}_i, \quad \vec{\Omega}_{eff}^i = \vec{\Omega}_i + 2 \sum_{j=1, j \neq i}^N J_{ij} \cdot \vec{r}_j$$

## DDTWA (Dissipative Discrete TWA)

dissipative discrete truncated Wigner approximation

$$\dot{\rho} = -i[H, \rho] + L_{deph}(\rho) + L_{decay}(\rho)$$

$$L_{deph}(\rho) = \frac{\Gamma_{\phi}}{2} \sum_{i=1}^N (\sigma_i^z \rho \sigma_i^z - \rho)$$

$$L_{decay} = \frac{\Gamma}{2} \sum_{i=1}^N (2\sigma_i^- \rho \sigma_i^+ - \sigma_i^+ \sigma_i^- \rho - \rho \sigma_i^+ \sigma_i^-)$$

### 直接使用DTWA的局限

经典自旋演化过程中的模长不再守恒，相互作用项将出现偏差  
目标是尽可能保持模长不变

### Dephasing的处理

$$d\vec{r}_i = \vec{\Omega}_{eff}^i \times \vec{r}_i dt + d\vec{r}_i|_{deph}$$

$$dr_i^x|_{deph} = -\Gamma_{\phi} r_i^x dt - \sqrt{2\Gamma_{\phi}} r_i^y dW_i$$

$$dr_i^y|_{deph} = -\Gamma_{\phi} r_i^y dt - \sqrt{2\Gamma_{\phi}} r_i^x dW_i$$

$$dr_i^z|_{deph} = 0$$

维纳过程  $\langle dW_i \rangle = 0, \langle (dW_i)^2 \rangle = dt$ ，从高斯分布中抽样

模长变化  $\langle d\vec{r}_i^2 \rangle = 0$

### Decay的处理

$$d\vec{r}_i = \vec{\Omega}_{eff}^i \times \vec{r}_i dt + d\vec{r}_i|_{decay}$$

$$dr_i^x|_{decay} = -\frac{\Gamma}{2} r_i^x dt - \sqrt{\Gamma} r_i^y dW_i$$

$$dr_i^y|_{decay} = -\frac{\Gamma}{2} r_i^y dt + \sqrt{\Gamma} r_i^x dW_i$$

$$dr_i^z|_{decay} = -\Gamma(r_i^z + 1)dt + \sqrt{\Gamma}(r_i^z + 1)dW_i$$

维纳过程  $\langle dW_i \rangle = 0, \langle (dW_i)^2 \rangle = dt$ ，从高斯分布中抽样

模长变化  $\langle d\vec{r}_i^2 \rangle = \Gamma(1 - \langle (r_i^z)^2 \rangle)dt$

由于  $\langle d(r_i^z)^2 \rangle = -2\Gamma r_i^z(r_i^z + 1)dt + \Gamma(r_i^z + 1)^2 dt$

$r_i^z$  小于0时， $d(r_i^z)^2$  大于0， $\langle (r_i^z)^2 \rangle$  趋于1，模长变化被抑制