

# PIMC for quantum spin models

## PIMC for transverse-field Ising model

$$H = \sum_{\langle ij \rangle} J_{ij} \sigma_i^z \sigma_j^z - \sum_i g_i \sigma_i^x$$

on a lattice  $\sigma^z \rightarrow -\sigma^z$   $\mathbb{Z}_2$  symmetry

for XXZ model

$$H = -J \sum_{\langle ij \rangle} \left[ \delta_i^x \delta_j^x + \delta_i^y \delta_j^y + \Delta \delta_i^z \delta_j^z \right]$$

nearest neighbors on a lattice

$\Delta = 1$   
 $\Delta \neq 1$

$SU(2)$   
 $U(1) \times \mathbb{Z}_2$

$$= \sum_{\langle ij \rangle} \left( H_{ij}^{xy} + H_{ij}^{zz} \right)$$

$$= \sum_{\langle ij \rangle_{\text{even}}} H_{ij} + \sum_{\langle ij \rangle_{\text{odd}}} H_{ij}$$

$\left[ H_{\text{even}} \right]$        $\left[ H_{\text{odd}} \right]$



$$= \sum_{\langle ij \rangle_{\text{even}}} \left( H_{ij}^{xy} + H_{ij}^{zz} \right) + \sum_{\langle ij \rangle_{\text{odd}}} \left( H_{ij}^{xy} + H_{ij}^{zz} \right)$$

$$|\bar{\sigma}\rangle = |\sigma_1, \dots, \sigma_N\rangle$$

$$\sigma_i^z |\sigma_i\rangle = \sigma_i |\sigma_i\rangle$$

$\uparrow \pm 1$

$$\text{Tr} [ e^{-\beta \mathcal{H}} ] = \sum_{\mathcal{N}_{\text{even}} + \mathcal{N}_{\text{odd}}} \prod_{\text{odd}} \langle \sigma_j \rangle e^{-\beta \mathcal{H}_{ij}} \prod_{\text{odd}} \langle \sigma_j \rangle e^{-\beta \mathcal{H}_{ij}^{\text{xy}}} e^{-\beta \mathcal{H}_{ij}^{\text{zz}}} e^{-\beta \mathcal{H}_{ij}^{\text{xy}}} e^{-\beta \mathcal{H}_{ij}^{\text{zz}}}$$

$$= \lim_{M \rightarrow \infty} \text{Tr} \left[ \left( e^{-\beta_M \mathcal{H}_{\text{odd}}} e^{-\beta_M \mathcal{H}_{\text{even}}} \right)^M \right]$$

$$= \lim_{M \rightarrow \infty} \text{Tr} \left[ \left( \prod_{\text{odd}} \langle \sigma_j \rangle e^{-\beta_M \mathcal{H}_{ij}^{\text{zz}}} e^{-\beta_M \mathcal{H}_{ij}^{\text{xy}}} \prod_{\text{even}} \langle \sigma_j \rangle e^{-\beta_M \mathcal{H}_{ij}^{\text{xy}}} e^{-\beta_M \mathcal{H}_{ij}^{\text{zz}}} \right)^M \right]$$

$\sum_{\vec{\sigma}} |\vec{\sigma}\rangle \langle \vec{\sigma}|$

$$= \lim_{M \rightarrow \infty} \sum_{\{\vec{\sigma}^{(k)}\}} \langle \vec{\sigma}^{(0)} | \prod_{\text{odd}} \langle \sigma_j \rangle e^{-\beta_M \mathcal{H}_{ij}^{\text{xy}}} | \vec{\sigma}^{(1)} \rangle \langle \vec{\sigma}^{(1)} | \prod_{\text{even}} \langle \sigma_j \rangle e^{-\beta_M \mathcal{H}_{ij}^{\text{xy}}} | \vec{\sigma}^{(2)} \rangle$$

$k = 0, \dots, 2M-1$

$$\dots \langle \vec{\sigma}^{(2M-2)} | \prod_{\text{odd}} \langle \sigma_j \rangle ( ) | \vec{\sigma}^{(2M-1)} \rangle$$

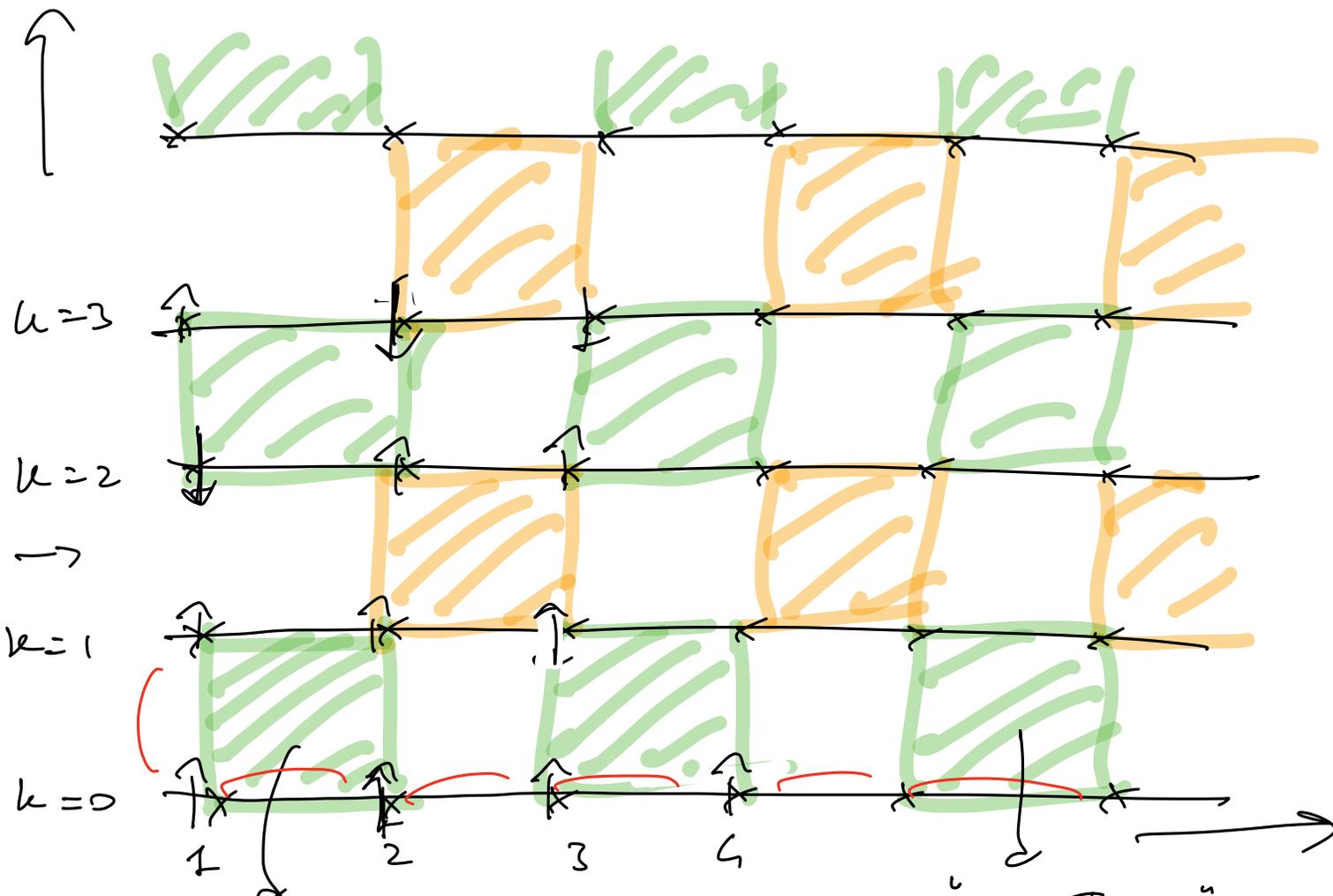
$$\langle \vec{\sigma}^{(2M-1)} | \prod_{\text{even}} \langle \sigma_j \rangle ( ) | \vec{\sigma}^{(0)} \rangle$$

$$e^{+\frac{\beta J}{8M}} \Delta \sum_{\alpha=0}^{2M-1} \sum_{\langle ij \rangle} \sigma_i^{(\alpha)} \sigma_j^{(\alpha)}$$

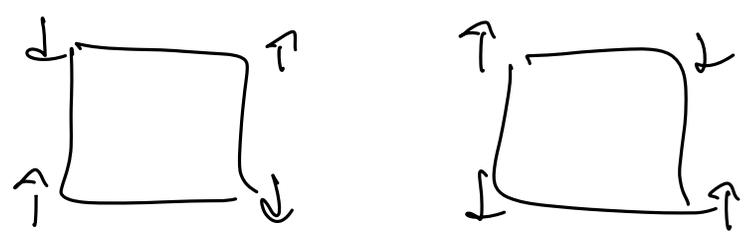
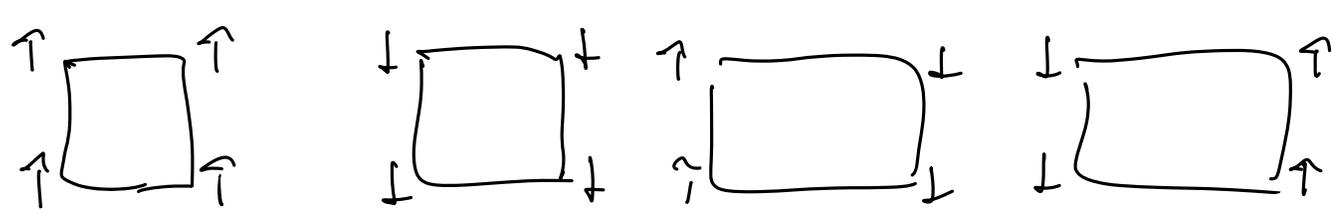
$$\sigma_i^{(\alpha)} = \pm 1$$

$$\prod_{\langle ij \rangle \text{ odd}} \langle \sigma_i^{(\alpha)} \sigma_j^{(\alpha)} | e^{-\beta_M \mathcal{H}_{ij}^{\text{xy}}} | \sigma_i^{(\alpha)} \sigma_j^{(\alpha)} \rangle$$

$$\mathcal{L} \left( \sigma_i^{(\alpha)} \sigma_j^{(\alpha)} ; \sigma_i^{(\alpha')} \sigma_j^{(\alpha')} \right) \rightarrow$$



$$W(\underbrace{\sigma_1^{(0)} \sigma_2^{(0)}}_{\text{vertices}}; \underbrace{\sigma_1^{(1)} \sigma_2^{(1)}}_{\text{plaquettes}}) \rightarrow \text{"vertices" / "plaquettes"}$$



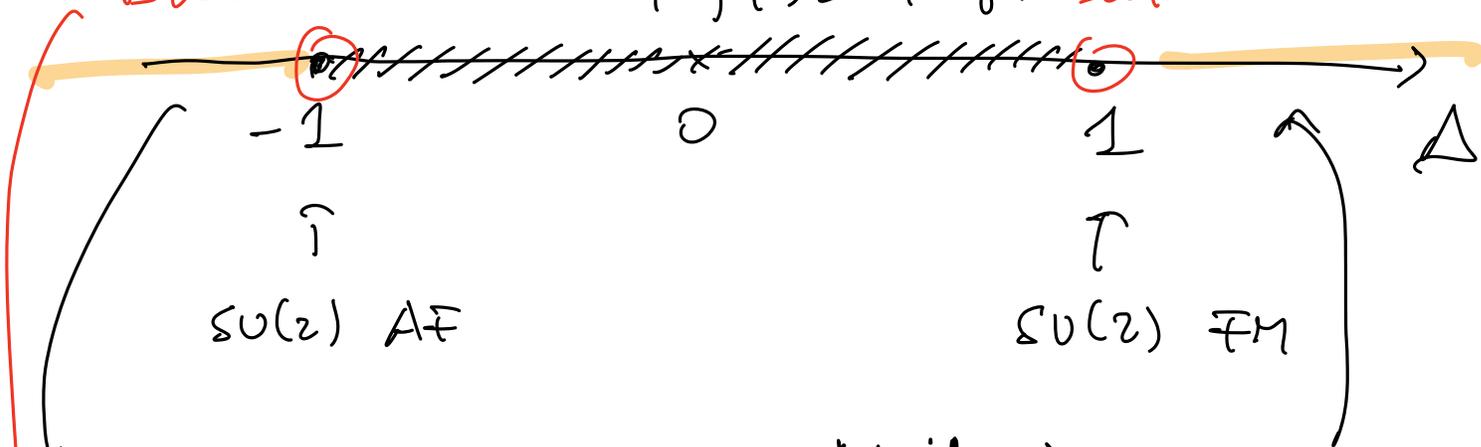
6-vertex model

$$\begin{array}{c}
 \uparrow \uparrow \\
 \uparrow \downarrow \\
 \downarrow \uparrow \\
 \downarrow \downarrow
 \end{array}
 \left(
 \begin{array}{ccccc}
 \uparrow \uparrow & \uparrow \downarrow & \downarrow \uparrow & \downarrow \downarrow & \\
 1 & 0 & 0 & 0 & \\
 0 & \cosh\left(\frac{\beta J}{2M}\right) & \sinh\left(\frac{\beta J}{2M}\right) & 0 & \\
 0 & \sinh\left(\frac{\beta J}{2M}\right) & \cosh\left(\frac{\beta J}{2M}\right) & 0 & \\
 0 & 0 & 0 & 0 & 1
 \end{array}
 \right)$$

$$\beta \rightarrow \infty$$

$$\beta > 0$$

$$\langle S_i^x S_j^x \rangle \sim \frac{1}{|i-j|} \rightarrow \infty |i-j|^\alpha \quad \text{BKT}$$



$$\langle S_i^x S_j^x \rangle \sim e^{-\left(\frac{|i-j|}{\xi}\right)}$$

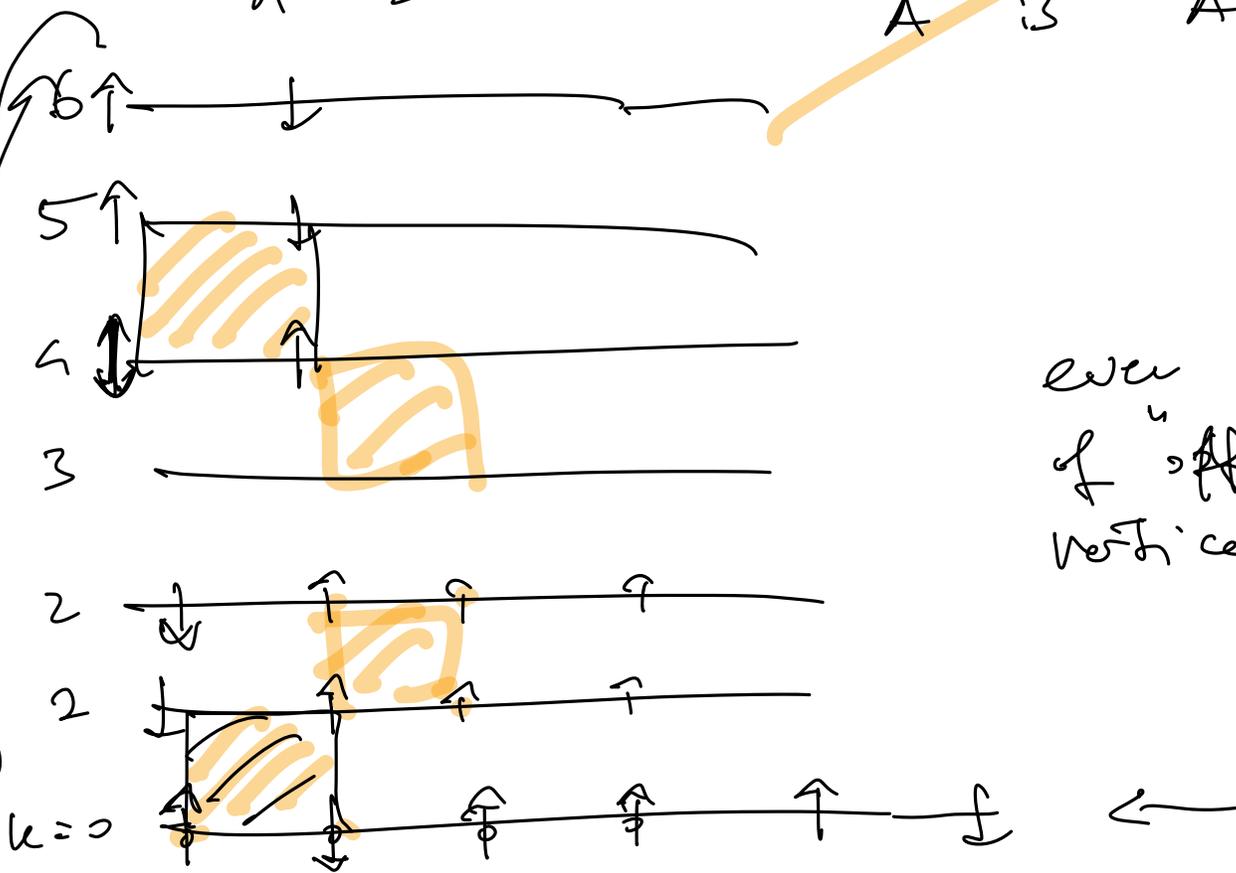
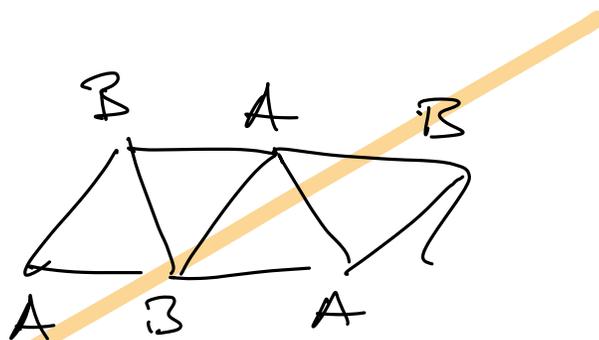
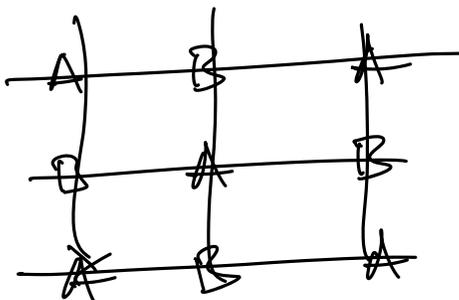
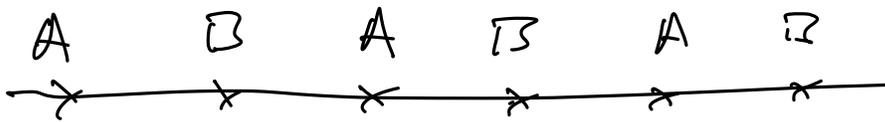
Berezinskii - Kostolitz - Thouless transition

$$\beta < 0$$

$$\sinh\left(\frac{\beta J}{2M}\right) < 0$$

Bipartite lattice

: two sublattices  $A$  &  $B$   
 such that there is <sup>no</sup> bond connecting  
 a site in  $A(B)$  to a site in  $A(B)$



even number  
 of "off-diagonal"  
 vertices

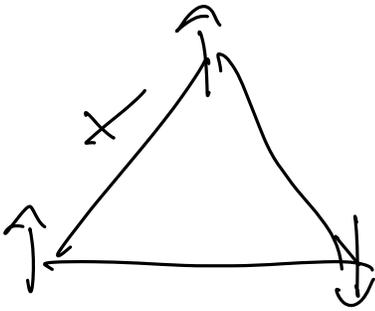
no sign problem

Non-bipartite lattice

(Triangular lattice, ...)

$\Rightarrow$

SIGN  
PROBLEM



Ground-state methods

: Variational MC

"Biased" approaches

choice of a variational "Ansatz"

$|\underline{\Psi}\rangle$  chosen to mimic the g.s.  
of  $\frac{H}{\text{Hamiltonian}}$

$\forall |\underline{\Psi}\rangle$

$$E(\underline{\Psi}) = \frac{\langle \underline{\Psi} | H | \underline{\Psi} \rangle}{\langle \underline{\Psi} | \underline{\Psi} \rangle} \geq E_0 = \langle \underline{\Phi}_0 | H | \underline{\Phi}_0 \rangle$$

ground state

$$\min_{|\Psi\rangle} E(|\Psi\rangle) = E_0$$

$|\vec{x}\rangle \rightarrow |x\rangle$   
 basis of Hilbert space  $\mathcal{H}$

$$|\Psi\rangle = \sum_{\vec{x}} \psi_{\vec{x}} |\vec{x}\rangle$$

$$\frac{\partial E}{\partial \psi_{\vec{x}}} = 0 \quad \text{condition}$$

$$\#(\vec{x}) \sim \mathcal{O}(\exp(N))$$

$$\psi_{\vec{x}} =: f(\vec{x}; \vec{\alpha})$$

$$\vec{\alpha} = (\alpha_1, \dots, \alpha_M) \in \mathbb{R}^M$$

WORKING ASSUMPTION

"Bias"

$$M \sim \text{poly}(N)$$

$f(\vec{x}; \vec{\alpha})$  computable in a time  $\sim \text{poly}(N)$

$$\min_{\vec{\alpha}} E(\vec{\alpha}) = \min_{\vec{\alpha}} \sum_{\vec{x}, \vec{x}'} f(\vec{x}; \vec{\alpha}) f(\vec{x}'; \vec{\alpha}) \langle \vec{x} | \mathcal{H} | \vec{x}' \rangle$$

$$\sum_{\vec{x}} |f(\vec{x}; \vec{\alpha})|^2$$

$$\sum_{\vec{x}} |f(\vec{x}; \vec{\alpha})|^2 \left( \sum_{\vec{x}'} \frac{f(\vec{x}'; \vec{\alpha})}{f(\vec{x}; \vec{\alpha})} \langle \vec{x} | \mathcal{H} | \vec{x}' \rangle \right)$$

$$\sum_{\vec{x}} |f(\vec{x}; \vec{\alpha})|^2$$

$$\Psi(\vec{x}; \vec{\alpha})$$

$$= \min_{\vec{\alpha}} \sum_{\vec{x}} \Psi(\vec{x}; \vec{\alpha}) \quad \underline{E_L(\vec{x})}$$

"local energy"

$$= \min_{\vec{\alpha}} \frac{\langle E_L(\vec{x}) \rangle}{|\Psi|^2} \rightarrow \text{Monte Carlo calculation}$$

$$\vec{x} \rightarrow \vec{x}' \rightarrow \vec{x}''$$

Sparse matrix

$$\langle \vec{x} | \mathcal{H} | \vec{x}' \rangle$$

$$\sum_{\langle ij \rangle} (\delta_i^x \delta_j^x + \delta_i^y \delta_j^y) = \frac{1}{2} (\delta_i^+ \delta_j^- + \delta_i^- \delta_j^+)$$

$$\langle \sigma_1, \sigma_2, \dots, \sigma_n | \mathcal{H} | \sigma_1', \sigma_2', \dots, \sigma_n' \rangle$$

Efficiently calculate  $E(\vec{\alpha})$

→ minimize w.r.t.  $\vec{\alpha}$

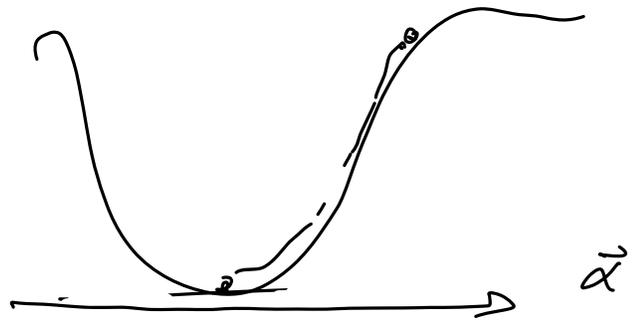
$$\vec{\nabla}_{\vec{\alpha}} E(\vec{\alpha})$$

gradient

$$\vec{\alpha}^{(0)}$$

first

guess



$$\vec{\alpha}^{(1)} = \vec{\alpha}^{(0)} - \epsilon \nabla_{\vec{\alpha}} E(\vec{\alpha}) \quad \epsilon \ll 1$$

...

$$\nabla_{\vec{\alpha}} E(\vec{\alpha}) = \frac{\sum_{\vec{x}, \vec{x}'} f^*(\vec{x}; \vec{\alpha}) f(\vec{x}', \vec{\alpha}) \langle \vec{x} | \mathcal{H} | \vec{x}' \rangle}{\sum_{\vec{x}} |f(\vec{x}; \vec{\alpha})|^2}$$

$$D_u(\vec{x}) = \frac{\partial}{\partial \alpha_u} \log f(\vec{x}; \vec{\alpha}) = \frac{\partial_{\alpha_u} f}{f}$$

$$D_u^*(\vec{x}) = \frac{\partial}{\partial \alpha_u} \log f^*(\vec{x})$$

$$\frac{\partial E(\vec{\alpha})}{\partial \alpha_u} = \frac{2\text{Re} \left[ \langle D_u^*(\vec{x}) | E_L(\vec{x}) \rangle \right]}{|f|^2} - \langle D_u^*(\vec{x}) \rangle \langle E_L(\vec{x}) \rangle$$

Examples of variational w.f.

$$1) \text{ Gaussian } |\vec{x}\rangle \rightarrow |\vec{R}\rangle = (r_1, r_2, \dots, r_n)$$

$$|\Psi\rangle = \int d^3R \psi(\vec{R}) |\vec{R}\rangle$$

$$\psi(\vec{R}) \stackrel{\text{ex.}}{=} \prod_i \psi_0(\vec{r}_i) \quad \text{BEC}$$

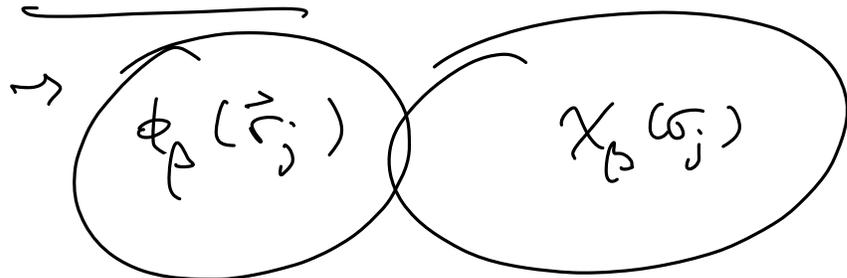
$$\stackrel{\text{ex.}}{=} \prod_{i < j} u(|\vec{r}_i - \vec{r}_j|) = e^{\sum_{i < j} w(|\vec{r}_i - \vec{r}_j|)} \quad \text{Jastrow v.f.}$$

$$u(r) \stackrel{\text{ex.}}{=} \sum_n \tilde{u}_n e^{i q_n r}$$

$$\stackrel{\text{ex.}}{=} u_0 + u_1 r + u_2 r^2 + u_3 r^3 + \dots$$

$\uparrow \quad \uparrow \quad \uparrow \quad \nearrow$   
 $\alpha_0 \quad \alpha_1 \quad \alpha_2 \quad \alpha_3$

## 2) Fermions



$$\Psi(\vec{r}_1 \sigma_1, \vec{r}_2 \sigma_2, \dots, \vec{r}_N \sigma_N) = \text{Det} [ \phi_{p_i}(\vec{r}_j) \chi_{p_i}(\sigma_j) ]$$

Slater determinant

$$\Psi(\vec{R}, \vec{\sigma}) = \sum_d C_d \text{Det} [ \quad ]$$

$$\Psi(\vec{R}, \vec{\sigma}) = \text{Det} [ \quad ] e^{\sum_{i < j} w(|\vec{r}_i - \vec{r}_j|)}$$