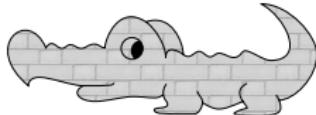


# A general backflow transformation

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**TCM**

# Quantum Monte Carlo

## The Schrödinger equation for an electronic system

$$\left[ \frac{1}{2} \sum_i \nabla_i^2 + \sum_{i < j} \frac{1}{|\mathbf{r}_i - \mathbf{r}_j|} - \sum_I \sum_i \frac{Z_I}{|\mathbf{r}_i - \mathbf{R}_I|} \right] \Phi_0(\mathbf{R}) = E_0 \Phi_0(\mathbf{R})$$

- Solvable exactly for very few systems.
- In general we don't know the exact  $\Phi_0(\mathbf{R})$ .
- We can give an approximate  $\Psi(\mathbf{R})$  instead.

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# Variational Monte Carlo

## Variational estimate of the energy

$$\begin{aligned} E_0 \leq E_{\text{V}}[\Psi] &= \frac{\int \Psi(\mathbf{R}) \hat{H}(\mathbf{R}) \Psi(\mathbf{R}) d\mathbf{R}}{\int \Psi^2(\mathbf{R}) d\mathbf{R}} \approx \\ &\approx \frac{1}{M} \sum_{m; \Psi^2}^M \frac{\hat{H}(\mathbf{R}_m) \Psi(\mathbf{R}_m)}{\Psi(\mathbf{R}_m)} = E_{\text{VMC}}[\Psi] \end{aligned}$$

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- is a very simple method
- along with the variational principle, allows for optimization of a parametrized  $\Psi(\mathbf{R}; \alpha)$
- is entirely dependent on the quality of  $\Psi(\mathbf{R})$

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The time-dependent Schrödinger equation

$$\hat{H}\Phi(\mathbf{R}, t) = i \frac{\partial\Phi(\mathbf{R}, t)}{\partial t}$$

$$\Phi(\mathbf{R}, t) = \sum_n c_n \Phi_n(\mathbf{R}) e^{-iE_n t}$$

In imaginary time  $\tau = it$ , with energy shift  $E_T$

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- Set  $\Phi(\mathbf{R}, 0) = \Psi(\mathbf{R})$ .
- $\lim_{\tau \rightarrow \infty} \Phi(\mathbf{R}, \tau) = \Phi_0(\mathbf{R})$ .
- DMC projects out  $\Phi_0(\mathbf{R})$  as  $\tau \rightarrow \infty$ .
- Implementation: ensemble of configurations subjected to drift+diffusion+branching in a discretized timeline.

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- is a more **complicated**, **expensive** method
- is a more accurate method
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## The Slater-Jastrow wave function

$$\Psi(\mathbf{R}) = \exp[J(\mathbf{R})] \Psi_S(\mathbf{R})$$

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$$J(\mathbf{R}) = J_{e-e}(\mathbf{R}) + J_{e-N}(\mathbf{R}) + J_{e-e-N}(\mathbf{R}) + \dots$$

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The backflow transformation

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$$J_{e-e}(\mathbf{R}) = \sum_{i < j}^M f(r_{ij}) \sum_v^p \lambda_v^{P_{ij}} r_{ij}^v$$

$$J_{e-N}(\mathbf{R}) = \sum_i^N \sum_I^M f(r_{iI}) \sum_\mu^q \lambda_\mu^{S_{ii}} r_{iI}^\mu$$

$$J_{e-e-N}(\mathbf{R}) = \sum_{i < j}^N \sum_I^M f(r_{iI}) f(r_{jI}) \sum_{v_{ij}}^p \sum_{\mu_{iI} \mu_{jI}}^q \lambda_{v_{ij} \mu_{iI} \mu_{jI}}^{P_{ij} S_{iI} S_{jI}} r_{ij}^{v_{ij}} r_{iI}^{\mu_{iI}} r_{jI}^{\mu_{jI}}$$

# General Jastrow factor

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$$\begin{aligned} J_{n,m}(\mathbf{R}) &= \sum_{i_1 < \dots < i_n}^N \sum_{I_1 < \dots < I_m}^M \sum_{\{v_{i_\alpha i_\beta}\}_{\alpha < \beta}^n}^p \sum_{\{\mu_{i_\alpha I_\gamma}\}_{\alpha=1, \gamma=1}^{q=n, \gamma=m}}^q \lambda_{\{v\}\{\mu\}}^{\{P\}\{S\}} \times \\ &\times \left( \prod_{\alpha < \beta}^n \Phi_{v_{i_\alpha i_\beta}}^{P_{\sigma(i_\alpha)\sigma(i_\beta)}}(\mathbf{r}_{i_\alpha i_\beta}) \right) \left( \prod_{\alpha}^n \prod_{\gamma}^m \Theta_{\mu_{i_\alpha I_\gamma}}^{S_{\sigma(i_\alpha)I_\gamma}}(\mathbf{r}_{i_\alpha I_\gamma}) \right) \end{aligned}$$

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$$\xi_{e-e}^i(\mathbf{R}) = \sum_{j \neq i}^M f(r_{ij}) \sum_v^p \lambda_v^{P_{ij}} r_{ij}^v \mathbf{r}_{ij}$$

$$\xi_{e-N}^i(\mathbf{R}) = \sum_I^M f(r_{iI}) \sum_\mu^q \omega_\mu^{S_{il}} r_{iI}^\mu \mathbf{r}_{iI}$$

$$\begin{aligned} \xi_{e-e-N}^i(\mathbf{R}) &= \sum_{j \neq i}^N \sum_I^M f(r_{il}) f(r_{jl}) \sum_{v_{ij}}^p \sum_{\mu_{il} \mu_{jl}}^q \left[ \lambda_{v_{ij} \mu_{il} \mu_{jl}}^{P_{ij} S_{il} S_{jl}} r_{ij}^v r_{il}^{\mu_{il}} r_{jl}^{\mu_{jl}} \mathbf{r}_{ij} + \right. \\ &\quad \left. + \omega_{v_{ij} \mu_{il} \mu_{jl}}^{P_{ij} S_{il} S_{jl}} r_{ij}^v r_{il}^{\mu_{il}} r_{jl}^{\mu_{jl}} \mathbf{r}_{il} \right] \end{aligned}$$

# General backflow transformation

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$$\xi_{n,m}^i(\mathbf{R}) = \sum_{i_2 < \dots < i_n}^N \sum_{I_1 < \dots < I_m}^M \sum_{\{v_{i_\alpha i_\beta}\}_{\alpha < \beta}^n}^p \sum_{\{\mu_{i_\alpha I_\gamma}\}_{\alpha=1, \gamma=1}^q}^{q=n, \gamma=m} \times$$

$$\left[ \sum_{k=2}^N [ik] \{P\} \{S\} \left( \prod_{\alpha < \beta}^n [ik] \Phi_{v_{i_\alpha i_\beta}}^{P_{\sigma(i_\alpha) \sigma(i_\beta)}} (\mathbf{r}_{i_\alpha i_\beta}) \right) \left( \prod_{\alpha}^n \prod_{\gamma}^m [ik] \Theta_{\mu_{i_\alpha I_\gamma}}^{S_{\sigma(i_\alpha) I_\gamma}} (\mathbf{r}_{i_\alpha I_\gamma}) \right) \mathbf{r}_{ik} + \right.$$

$$\left. \sum_{K=2}^N [iK] \{P\} \{S\} \left( \prod_{\alpha < \beta}^n [iK] \Delta_{v_{i_\alpha i_\beta}}^{P_{\sigma(i_\alpha) \sigma(i_\beta)}} (\mathbf{r}_{i_\alpha i_\beta}) \right) \left( \prod_{\alpha}^n \prod_{\gamma}^m [iK] \Lambda_{\mu_{i_\alpha I_\gamma}}^{S_{\sigma(i_\alpha) I_\gamma}} (\mathbf{r}_{i_\alpha I_\gamma}) \right) \mathbf{r}_{iK} \right]$$

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