

FYS4411/9411 FEB 17, 2022

$$W_i(t+\Delta t) \rightarrow W(\vec{y}, t+\Delta t)$$

$$= \int_{x \in \mathbb{D}} \underbrace{W(\vec{y}, t+\Delta t | \vec{x}, t) W(\vec{x}, t)}_{\text{unknown}} d\vec{x}$$

continuous Markov chain

How do these expressions relate to quantum mech?

$$\hat{H} |\phi_0\rangle = E_0 |\phi_0\rangle$$

$$\hat{H} \hat{H}^{-1} = \hat{H}^{-1} \hat{H} = \underline{1}$$

$$\hat{H}^{-1} \hat{H} |\underline{\phi}_0\rangle = E_0 \hat{H}^{-1} |\underline{\phi}_0\rangle$$

\hat{H} is a differential operator

$$\hat{H} = -\frac{\nabla^2 \psi^2}{2m} - V(\vec{x}) \quad (\text{1 single particle})$$

\hat{H}^{-1} is an integral operator

insert $\int_{-\infty}^{\infty} |\vec{x}\rangle \langle \vec{x}| d\vec{x}$

and multiply from the left with $\langle \vec{y} |$

$$\langle \vec{y} | \Phi_0 \rangle = \Phi(\vec{y}) =$$

$$E_0 \int_{-\infty}^{\infty} \langle \vec{y} | H^{-1} | \vec{x} \rangle \langle \vec{x} | \Phi_0 \rangle d\vec{x}$$

$$= E_0 \int_{-\infty}^{\infty} \langle \vec{y} | H^{-1} | \vec{x} \rangle \Phi_0(\vec{x}) d\vec{x}$$

Green's function/
propagator

$$G(\vec{y}, \vec{x}) = \langle \vec{y} | H^{-1} | \vec{x} \rangle$$

$$\Phi_0(\vec{y}) = \hat{H}(\vec{y}) \hat{H}^{-1} \Phi_0(\vec{y})$$

$$= \int_{-\infty}^{\infty} \hat{H}(\vec{y}) G(\vec{y}, \vec{x}) \Phi_0(\vec{x}) d\vec{x}$$

$$\hat{H}(\vec{y}) G(\vec{y}, \vec{x}) = \delta(\vec{y} - \vec{x})$$

$$\hat{H}(\vec{x}) G(\vec{x}, \vec{y}) = \delta(\vec{x} - \vec{y})$$

Connection with Diffusion equation (1-Dim)

$$\frac{\partial w(x, t)}{\partial t} = D \frac{\partial^2 w(x, t)}{\partial x^2}$$

$$w(x, t) = \frac{1}{\sqrt{4\pi Dt}} \exp\left\{-\frac{x^2}{4Dt}\right\}$$

$$w(y, t+\Delta t) = \int_{-\infty}^{\infty} W(y, t+\Delta t | x, t) w(x, t) dx$$

$$\frac{\partial W(y, t+\Delta t | x, t)}{\partial t} =$$

$$D \frac{\partial^2}{\partial x^2} W(y, t+\Delta t | x, t)$$

$$W(y, t+\Delta t | x, t) =$$

$$\frac{1}{\sqrt{4\pi D \Delta t}} \exp \left\{ -\frac{(y-x)^2}{4D \Delta t} \right\}$$

Factor g in Metropolis's Hastings ;

$$g = \frac{W(y_{t+\Delta t} | x_t) W(y_{t+\Delta t})}{W(x_t | y_{t+\Delta t}) W(x_t)}$$

without quantum force

$$g = \frac{W(y_{t+\Delta t})}{W(x_t)}$$

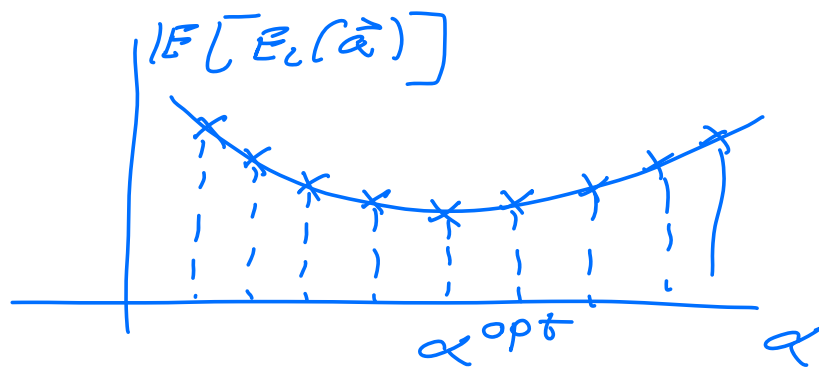
Basic elements in a VMC code

- Metropolis's sampling ✓
- Analytical expression for $E_L(\vec{R}; \vec{\alpha})$ ✓

- importance sampling

$$\vec{F} = \frac{2}{\psi_T} \vec{\nabla} \psi_T \quad \checkmark$$

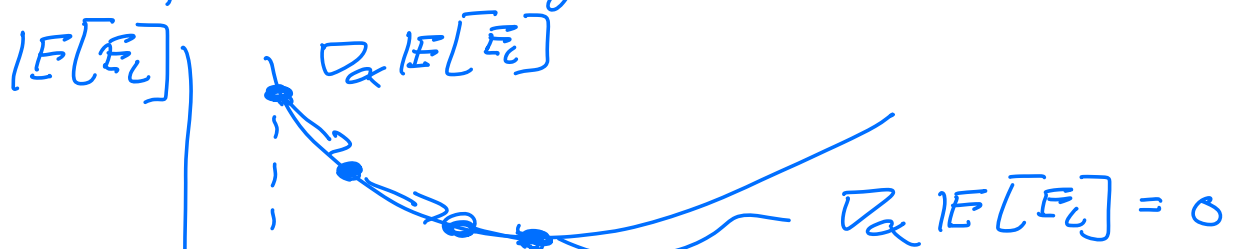
- optimization of

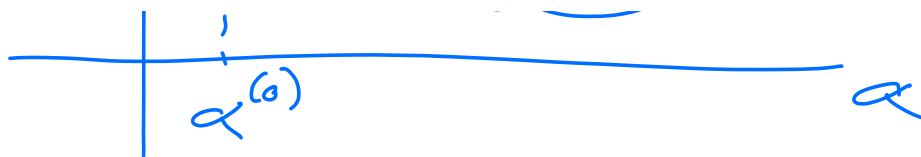
$$E[E_L(\vec{\alpha})]$$


$$\alpha^{\text{opt}} = \arg \min_{\alpha \in \mathbb{R}^M} E[E_L(\vec{\alpha})]$$

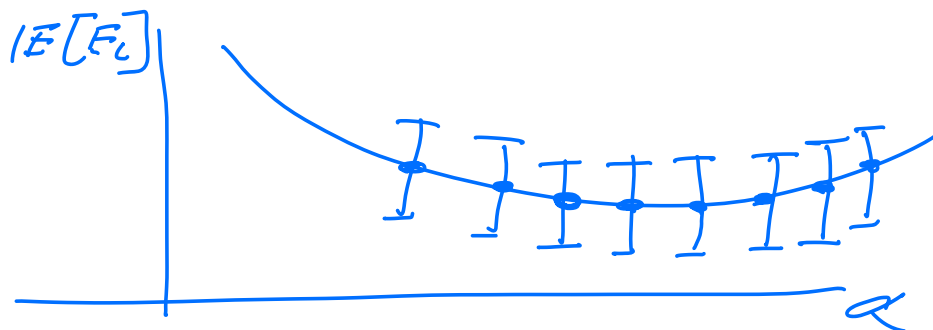
can we optimize the search for α^{opt} with less Monte Carlo cycles?

yes \rightarrow gradient methods





- Gradient optimization
- After this : stochastic resampling method : we want proper estimation of errors



error (standard deviation) estimate.

Post analysis with resampling methods :
Blocking, Bootstrap, Jackknife

- parallelization
 - Gradient optimization
- $\dots \dots \dots (\vec{\theta}_1, \vec{\theta}_2) F(\vec{\theta}_1, \vec{\theta}_2)$

$$E[E_L(\vec{\alpha})] = \int d\vec{R} P(\vec{R}; \vec{\alpha}) E_L(\vec{R}; \vec{\alpha})$$

$$\simeq \frac{1}{M} \sum_{i=1}^M E_L(\vec{R}_i; \vec{\alpha})$$

$$P_T(\vec{R}; \vec{\alpha}) = \frac{|\psi_T(\vec{R}; \vec{\alpha})|^2}{\int d\vec{R} |\psi_T|^2}$$

$$E_L(\vec{R}; \vec{\alpha}) = \frac{1}{\psi_T} \hat{H} \psi_T(\vec{R}; \vec{\alpha})$$

$$\psi_T(\vec{R}; \vec{\alpha}) \rightarrow \psi_T(\vec{\alpha})$$

$$P_T(\vec{R}; \vec{\alpha}) \rightarrow P_T(\vec{\alpha})$$

$$E_L(\vec{R}; \vec{\alpha}) \rightarrow E_L(\vec{\alpha})$$

$$\frac{\partial E[E_L(\vec{\alpha})]}{\partial \vec{\alpha}} = 0 \quad ?$$

$$= \int d\vec{R} \left[\frac{dP_T(\vec{\alpha})}{d\vec{\alpha}} E_L(\vec{\alpha}) + P_T(\vec{\alpha}) \frac{dE_L(\vec{\alpha})}{d\vec{\alpha}} \right]$$

Example: HO in 1 Dim

$$\psi(\vec{\alpha}) = \psi(\alpha) = e^{-\frac{1}{2}\alpha^2 x^2}$$

'T

$$\frac{dP(\alpha)}{d\alpha}$$

$$\wedge \frac{dE_L(\alpha)}{d\alpha}$$

$$\frac{dp(\alpha)}{d\alpha} = \frac{d}{d\alpha} \left[\frac{e^{-\frac{1}{2}\alpha^2 x^2}}{\int dx e^{-\alpha^2 x^2}} \right]$$

$$\frac{d(E[E_L(\alpha)])}{d\alpha} =$$

$$2 \left[E \left[\frac{d \ln \psi(\alpha)}{d\alpha} E_L(\alpha) \right] \right]$$

$$- \left[E \left[\frac{d \ln \psi(\alpha)}{d\alpha} \right] E[E_L(\alpha)] \right]$$

$$= 0$$

new integral

$$\int dx \frac{d \ln \psi(x, \alpha)}{d\alpha} E_L(x; \alpha)$$

new integral

$$\int dx \frac{d \ln \psi(x, \alpha)}{d\alpha}$$

$$\frac{\left(\frac{d \psi(\alpha)}{d \alpha} \right)}{\psi(\alpha)}$$

- set up 3 integrals, two new ones and $\int dx p(x; \alpha) E_L(x; \alpha)$
- use automatic differentiation