

# Markov - Chain Monte Carlo

M → P

$$\mathbb{I} = \int_V d^D x \frac{p(\vec{x})}{N} g(\vec{x}) = \langle g \rangle_p$$

$\vec{x}_n :$   $\vec{x}_1 \rightarrow \vec{x}_2 \rightarrow \dots \rightarrow \vec{x}_n$

$$\frac{n(\vec{x}_n)}{M} \xrightarrow[M \rightarrow \infty]{} \frac{p(\vec{x}_n)}{N} d^D x$$



$$I_M = \frac{1}{M} \sum_{n=1}^M g(\vec{x}_n) = \underset{\xrightarrow[M \rightarrow \infty]{} I}{\sum_{\vec{x}} \frac{n(\vec{x})}{M} g(\vec{x})}$$

rule  $\Rightarrow T(\vec{x} \rightarrow \vec{y}) = T_{p \leftarrow p}(\vec{x} \rightarrow \vec{y}) A(\vec{x} \rightarrow \vec{y})$

$$A(\vec{x} \rightarrow \vec{y}) = \min \left( 1, \frac{p(\vec{y})}{p(\vec{x})} \right)$$

$$\frac{|I_M - I|}{M} \xrightarrow[M \rightarrow \infty]{} 0 \quad \text{How?}$$

$I_M$  is a sum of random variables

$g(\vec{x}_n)$  are random variables

$\vec{x}_n$  are distributed according to  $p(\vec{x})$

$$\overline{g(\vec{x})} = \int d\vec{x} x \frac{p(\vec{x})}{N} g(\vec{x}) = I$$

$$\Rightarrow \overline{I}_M = I$$

$$\sigma_{I_M}^2 \sim (I - \overline{I}_M)^2$$

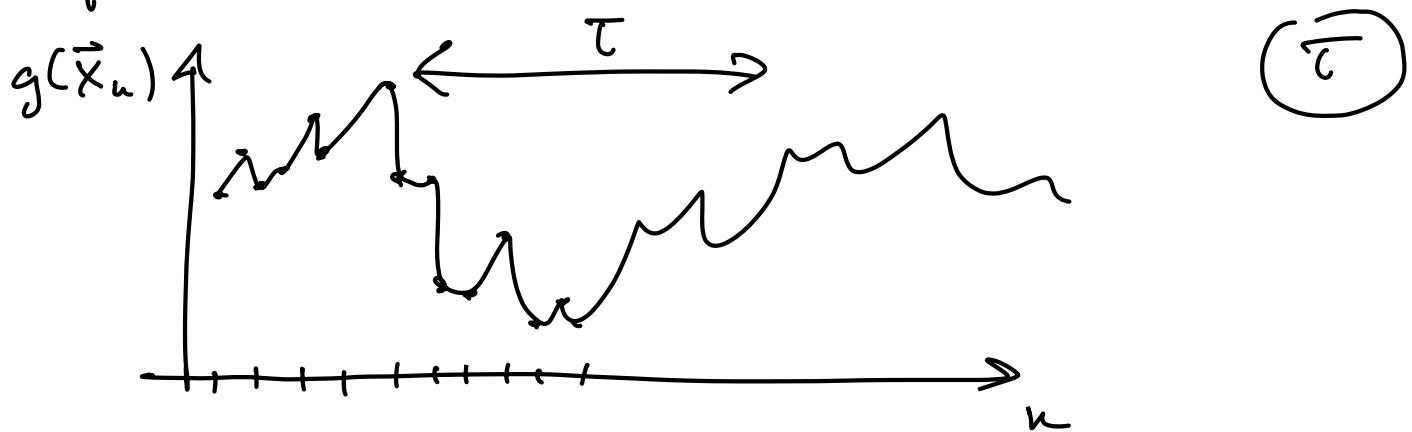
$\overline{g(\vec{x}_n)}$  is a random variable

$$\overline{g} = I, \quad \sigma_g^2 \text{ Variance}$$

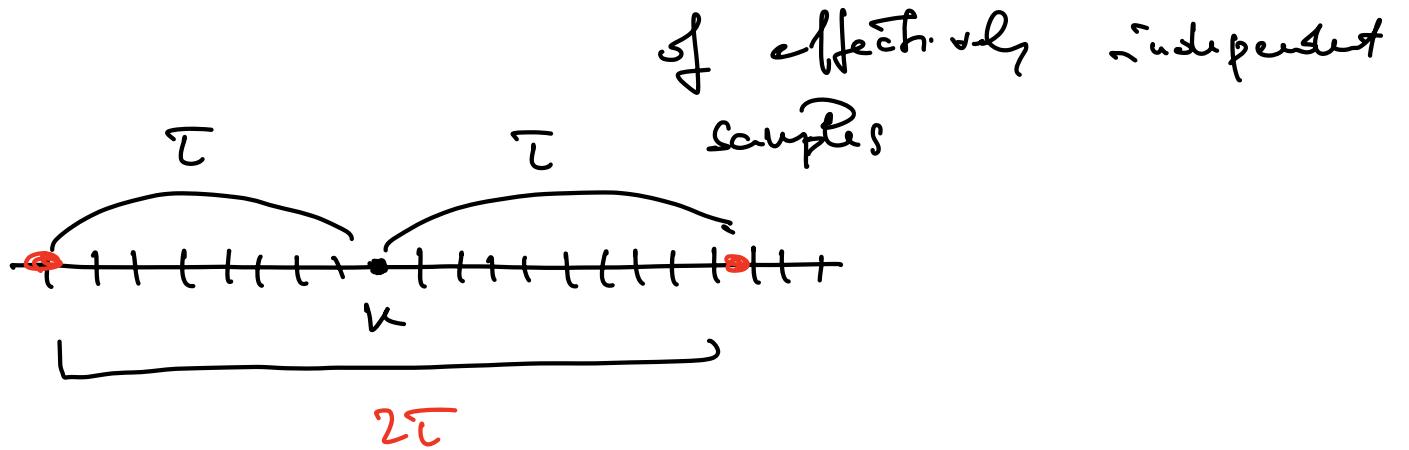
$$\sigma_{I_M}^2 = \frac{M \sigma_g^2}{M^2} = \frac{\sigma_g^2}{M}$$

true if  
 $g(\vec{x}_n)$  are  
independent  
random variables

equilibration time / autocorrelation time



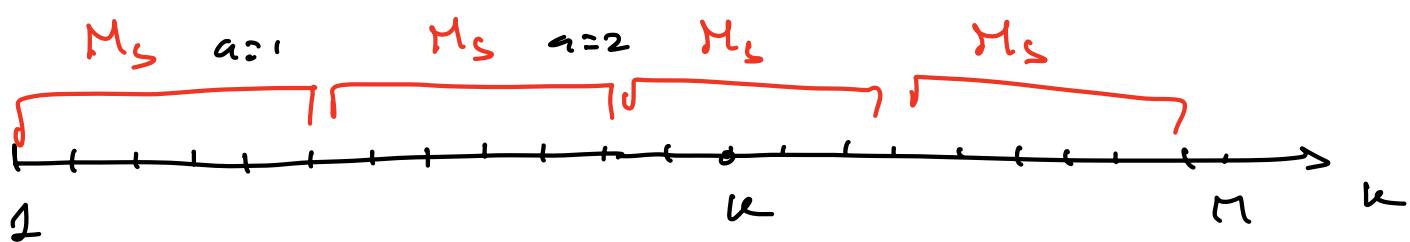
$$M \text{ samples of } g(\vec{x}_n) \rightarrow M_{\text{eff}} = \frac{M}{2\tau}$$



$$\sigma_{\bar{I}_M}^2 = \frac{\sigma_g^2}{M_{\text{eff}}} = \frac{2c}{M} \sigma_g^2 \quad \left| \text{Central limit theorem} \right.$$

How to estimate  $\bar{T}$  ?

An independent estimate of  $\sigma_{\bar{I}_M}^2$ :



$$n_s = \frac{M}{M_s} \quad M_s \geq T$$

$$\bar{I}_M = \frac{1}{M} \sum_{n=\bar{n}+1}^{\bar{n}+M} g(\vec{x}_n) = \frac{1}{n_s} \sum_{a=1}^{n_s} \frac{1}{M_s} \sum_{j=1}^{M_s} g(\vec{x}_{(a-1)M_s+j})$$

$\downarrow$

after many visited  
 $\vec{x}$  configurations

Shock average

$M_s > T \rightarrow \bar{g}_a$  shock averages are independent

$$I_M = \frac{1}{n_s} \sum_a \vec{g}_a$$

$\Rightarrow$  central limit theorem

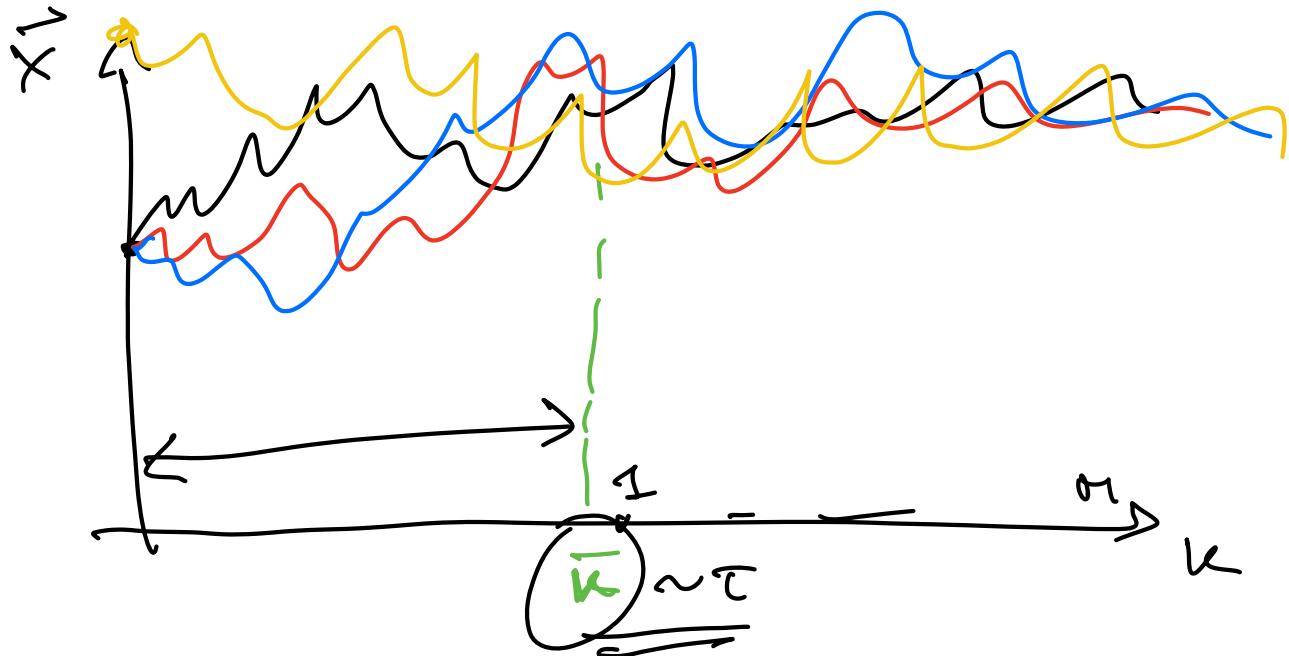
$$\sigma_{I_M}^2 = \frac{\sigma_g^2}{n_s}$$

Variance of the fluctuations of  $\vec{g}_a$

$$= \frac{2\pi}{M} \sigma_g^2$$

Variance of the fluctuations of  $g(\vec{x}_n)$

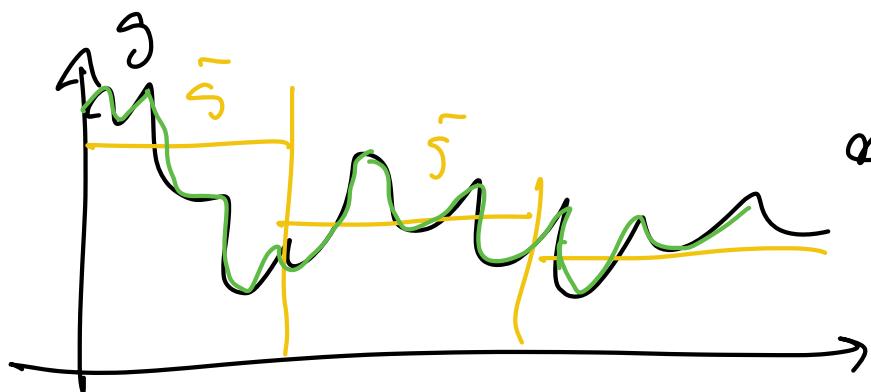
$$\tau = \frac{1}{2} \sigma_g^2 \frac{M_s}{n_s}$$



In practice : guess  $\bar{n}$  ( $\geq \tau$ ) (thermalization time)

guess  $M_S (\geq \tau)$

estimate  $\tau = \frac{1}{2} M_S$



$$\Rightarrow \sigma^2 \tau \leq M_S, \bar{n}$$

✓  $\sigma_{\text{fluct}}^2 = \frac{\sigma_1^2}{n_S}$

$$|\frac{\bar{I} - I_M}{I_M}| \underset{M \rightarrow \infty}{\sim} \frac{\sqrt{\frac{2\tau}{M}}}{\sqrt{\frac{\tau}{M}}} \sigma_0$$

$I = \int \vec{x}^\alpha x^\beta \frac{f(\vec{x})}{n} g(\vec{x})$

$$\nabla = \underline{\underline{L}}^A$$

$$\tau = \tau(D, L) ?$$

luckily :

$$\boxed{\tau \approx A D^{\frac{1}{z}} L^z}$$

$$z \sim 0(1)$$

$\rightarrow z = \text{dynamical exponent}$

typically

typically

$$\underline{\underline{z = 2}}$$

$\Rightarrow$  Efficient estimate of  $\tau$

$$M \rightarrow P$$

Path-integral formulation of quantum statistical mechanics

$\rightarrow$  Path-integral Monte Carlo (PIMC)

single particle in continuum space

$$\hat{H} = \frac{\hat{p}^2}{2m} + V(\hat{x})$$

$$Z = \text{Tr} (e^{-\beta \hat{H}})$$

$$\beta = \frac{1}{k_B T}$$

$$\langle \hat{A} \rangle = \frac{\text{Tr}(\hat{A} e^{-\beta \hat{H}})}{\text{Tr}(e^{-\beta \hat{H}})}$$

$$\vec{x} = (x_1, x_2, \dots, x_d)$$

$$Z = \sum_{\alpha} \langle \alpha | e^{-\beta \hat{H}} | \alpha \rangle$$

$$|\alpha\rangle \rightarrow |\vec{x}\rangle$$

basis of the position operator ( $\vec{x}$ )

$$\vec{x}_1, \vec{x}_2, \dots, \vec{x}_d$$

$$= \int d^d x \underbrace{\langle \vec{x} |}_{\substack{\downarrow \\ \{ }} e^{-\beta \hat{H}} \underbrace{| \vec{x} \rangle}_{\substack{\uparrow \\ \{ }} \leftarrow}$$

$$\underbrace{\langle \vec{x} | e^{-\beta \hat{H}} | \vec{x}' \rangle}_{\substack{\uparrow \\ \{ }} = \rho(\vec{x}, \vec{x}'; \beta)$$

$$it = \beta$$

imaginary-time propagator

$$\frac{1}{i} \hat{H} t$$

$\langle \vec{x} | e^{\frac{i}{\hbar} \hat{H} t} | \vec{x}' \rangle$  = amplitude of prob. of going from  $\vec{x}'$  to  $\vec{x}$  in a time  $t$  (propagates)

$$\langle \vec{x} | e^{-\beta \left( \frac{\vec{p}^2}{2m} + U(\vec{x}) \right)} | \vec{x} \rangle$$

$\underbrace{\phantom{e^{-\beta \frac{\vec{p}^2}{2m}}}}$

$$= e^{-\beta \frac{\vec{p}^2}{2m}} e^{-\beta U(\vec{x})}$$

Trotter - Lie decomposition

$$e^{-\beta \left( \frac{\vec{p}^2}{2m} + U(\vec{x}) \right)} = \lim_{M \rightarrow \infty} \left( e^{-\beta \frac{\vec{p}^2}{2m}} e^{-\beta U(\vec{x})} \right)^M$$

$\uparrow \quad \downarrow$

$$e^A = (e^{A_M})^M$$

$$= \lim_{M \rightarrow \infty} \left( e^{-\beta_M \frac{\vec{p}^2}{2m}} e^{-\beta_M U(\vec{x})} \right)^M$$

signel on  $|\vec{x}\rangle$

$$e^{-\beta_M \left( \frac{\vec{p}^2}{2m} + U(\vec{x}) \right)} \approx e^{-\beta_M \frac{\vec{p}^2}{2m}} e^{-\beta_M U(\vec{x})} + O\left(\frac{\beta}{M}\right)^2$$

$$\int d\vec{x} |\vec{x}\rangle \langle \vec{x}| = 1$$

$\uparrow$

completeness  
relations

$$\langle \vec{x} | e^{-\beta (\frac{P^2}{2m} + U)} |\vec{x}' \rangle$$

$$= \lim_{N \rightarrow \infty}$$

$$\langle \vec{x} | e^{-\frac{\beta P^2}{2m}} \sum_{n=1}^{M-1} d\vec{x}_n$$

$$e^{-\frac{\beta P^2}{2m}}$$

$$|\vec{x}_1\rangle \langle \vec{x}_1| e^{-\frac{\beta P^2}{2m}} |\vec{x}_2\rangle \langle \vec{x}_2| e^{-\frac{\beta P^2}{2m}} \dots$$

$$\dots \langle \vec{x}_{M-1} | e^{-\frac{\beta P^2}{2m}} \langle \vec{x}' \rangle$$

$$- \sum_{i=1}^M [U(\vec{x}_1) + U(\vec{x}_2) + \dots + U(\vec{x}_{M-1}) + U(\vec{x}')]$$

↑

evaluate this "Free propagation"

$$\langle \vec{x}_{u+1} | e^{-\frac{\beta P^2}{2m}} |\vec{x}_u \rangle$$

$$|\vec{p}\rangle \rightarrow |\vec{p}'\rangle$$

$$: \int d^3 p' |\vec{p}'\rangle \langle \vec{p}| = 1$$

$$= \int \frac{d^3 p}{(2\pi)^3} \langle \vec{x}_u | \vec{p} \rangle \langle \vec{p} | \vec{x}_{u+1} \rangle e^{-\frac{\beta P^2}{2m}}$$

$$e^{\frac{i}{\hbar} \vec{p} \cdot (\vec{x}_u - \vec{x}_{u+1})}$$

$$\begin{aligned}
 &= \frac{1}{\Gamma} \left( x_u^{(j)} - x_{u+1}^{(j)} \right) \\
 &\stackrel{(j)}{=} \frac{(x_u^{(j)} - x_{u+1}^{(j)})}{\int \frac{dp}{2\pi\hbar}} e^{-\frac{\Delta}{2\mu_m} p^2} \\
 &\stackrel{1}{=} \frac{1}{2\pi\hbar} \left[ \frac{2\sqrt{mM}}{p} \right] e^{-\frac{mM}{2\beta\hbar^2} (x_u^{(j)} - x_{u+1}^{(j)})^2} \\
 &\quad \sim \ell^{-2}
 \end{aligned}$$

de Broglie thermal wavelength

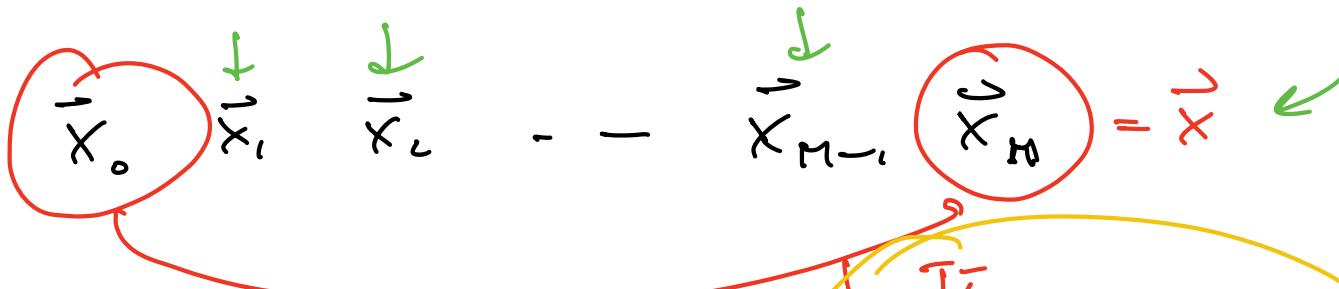
$$\lambda = \frac{\hbar}{\sqrt{2\mu m k_B T}}$$

$$\frac{mM}{2\beta\hbar^2} = \frac{\overline{\mu} M}{\ell^2}$$

$$\begin{aligned}
 &\langle \vec{x} | e^{-\beta \hat{H}} | \vec{x}' \rangle \\
 &\rightarrow = \lim_{M \rightarrow \infty} \left( \frac{M^{\frac{1}{2}}}{\ell} \right)^{dM} \int_{-\infty}^{\infty} \frac{M-1}{\ell} \left( d^d x_u \right) \\
 &\quad - \pi \frac{M}{\ell^2} \sum_{u=0}^{M-1} \left( \vec{x}_u - \vec{x}_{u+1} \right)^2 e^{-\frac{\Delta}{M} \sum_{u=0}^{M-1} U(\vec{x}_u)}
 \end{aligned}$$

$$Z = \int d^d x e^{-\beta H}$$

$$= \lim_{M \rightarrow \infty} \left( \frac{M^{r_2}}{\pi} \right)^{dM} \int \left( \prod_{n=1}^{M-1} dx_n \right) e^{-S[\vec{x}_n]}$$



$$S[\vec{x}_n] = \beta \left[ \frac{\pi M}{\beta \lambda^2} \sum_{n=0}^{M-1} (\vec{x}_n - \vec{x}_{n+1})^2 + \frac{1}{M} \sum_{n=1}^M U(\vec{x}_n) \right]$$

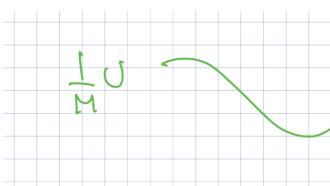
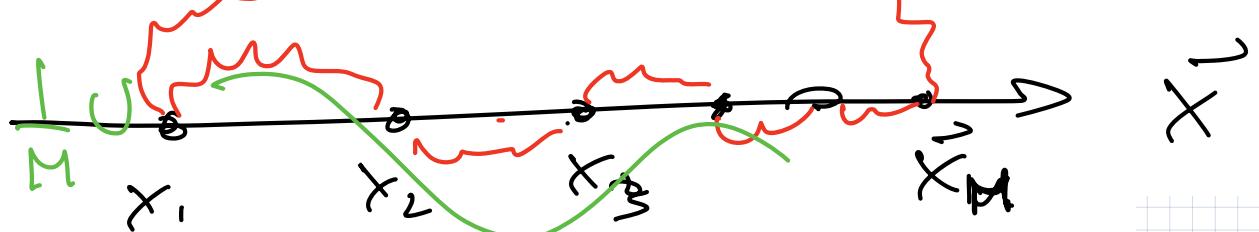
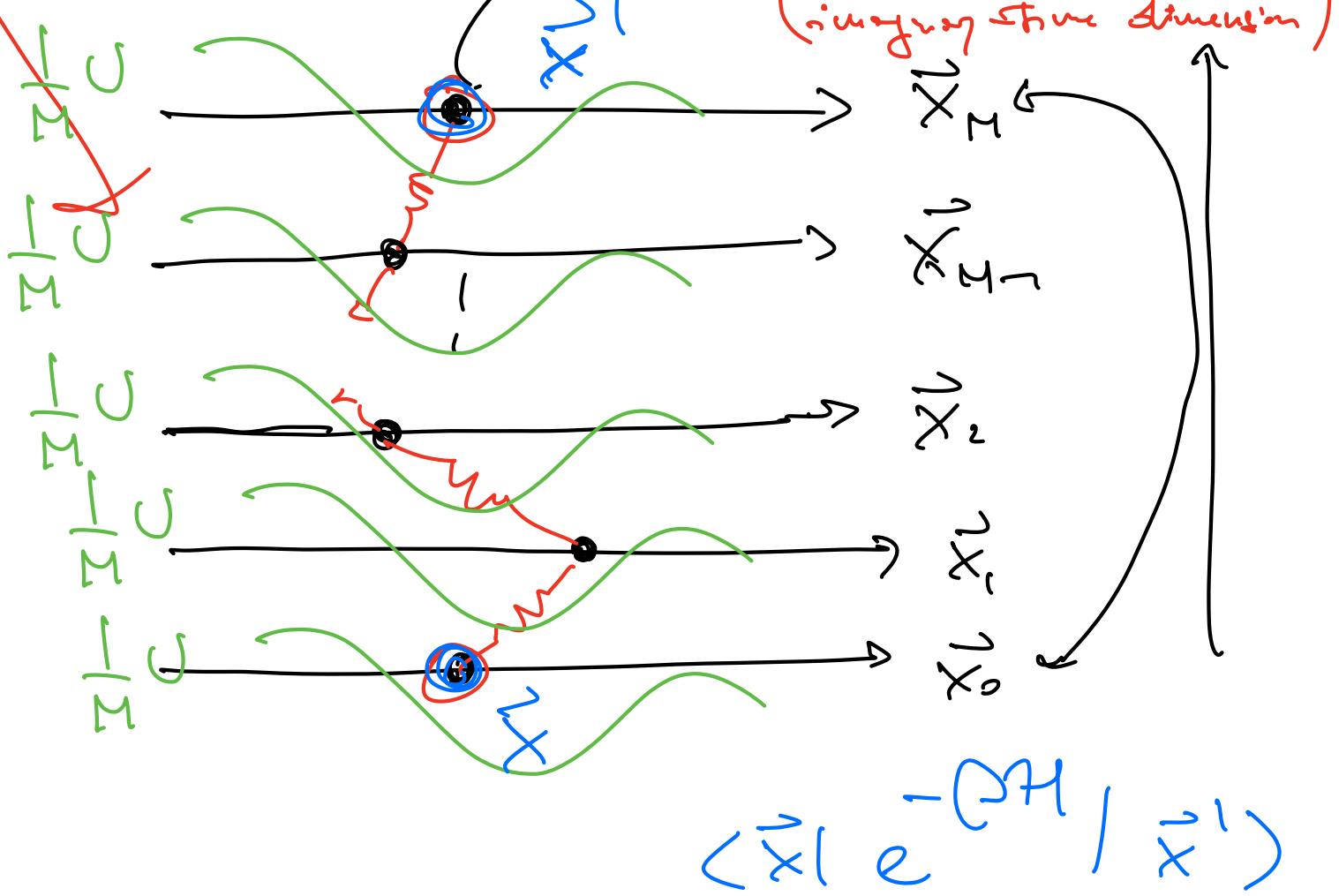
$\downarrow$   
effective action

$$= \beta H_{\text{eff}}(\{\vec{x}_n\})$$

$$\underline{k} = \frac{2\pi M}{\beta \lambda^2}$$

spring constant

"leads" (Froster dimension)



Ring electric polymer

$Z = (\ ) \times \lim_{M \rightarrow \infty}$  partition function of  
a ring polymer of  
 $M$  beads

Spring constant

$$\bar{k} = \frac{2\pi M}{(\Delta L)^2}$$

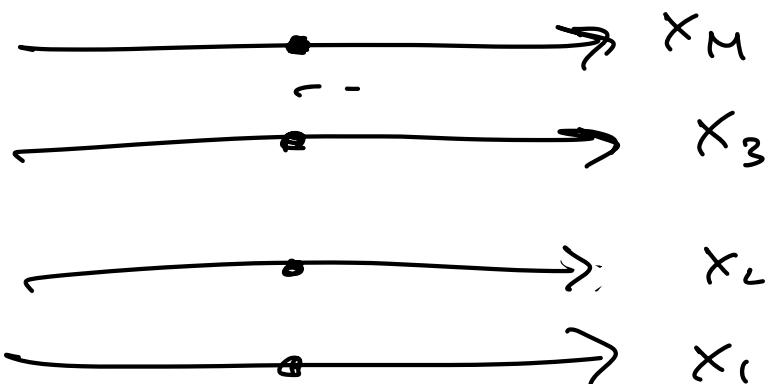
$$= \frac{2\pi M}{h} 2\pi m (u_3 \tau)^2$$

$$h = \frac{h}{2\pi m u_3 \tau}$$

$$= \begin{cases} \rightarrow 0 & \tau \rightarrow 0 \\ \rightarrow \infty & \tau \rightarrow \infty \end{cases}$$

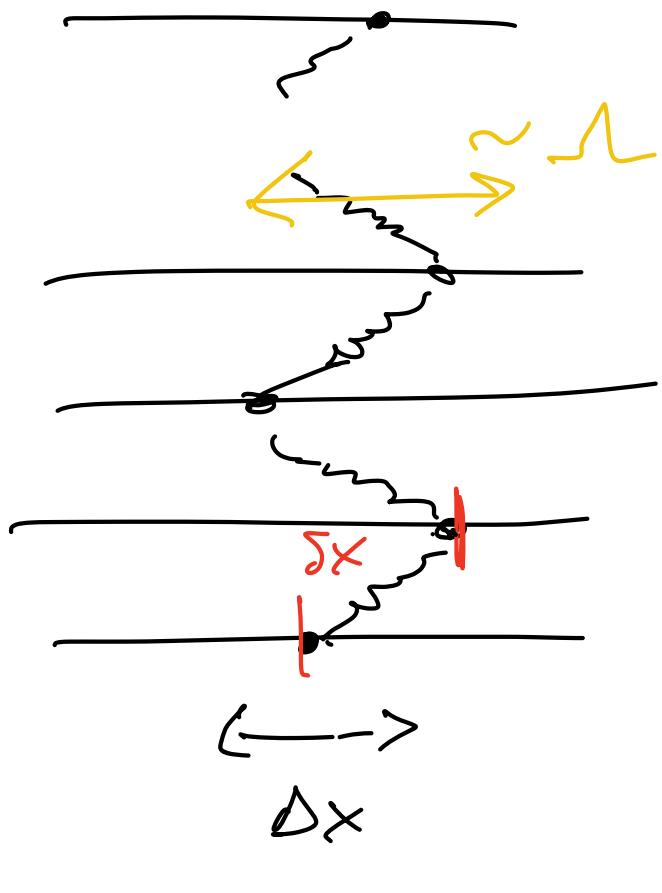


My polymer becomes  
softly stiff



$$Z \sim \int d^d x e^{-\beta U(\vec{x})} \int d^d p e^{-\beta \frac{p^2}{2m}}$$

$T \rightarrow 0$



$$\frac{1}{2\sigma^2}$$

$$\delta x = x_u - x_{u+1}$$

Gaussian variables

$\sim$

$$e^{-\left(\frac{\bar{u}M}{\sigma^2}\right)(\delta x)^2}$$

$$\rightarrow \sigma^2 = \frac{n^2}{2\bar{u}M}$$

$$(\Delta x)^2 = M \sigma^2 = \frac{n^2}{2\bar{u}}$$

$$\Delta x = \sqrt{\frac{n}{2\bar{u}}}$$

$$\vec{x}_1 \rightarrow \vec{x}_2 \rightarrow \dots \rightarrow \vec{x}_M$$

"path" in imaginary time

$Z =$  integral over paths =  
.. over ring polymer  
configurations