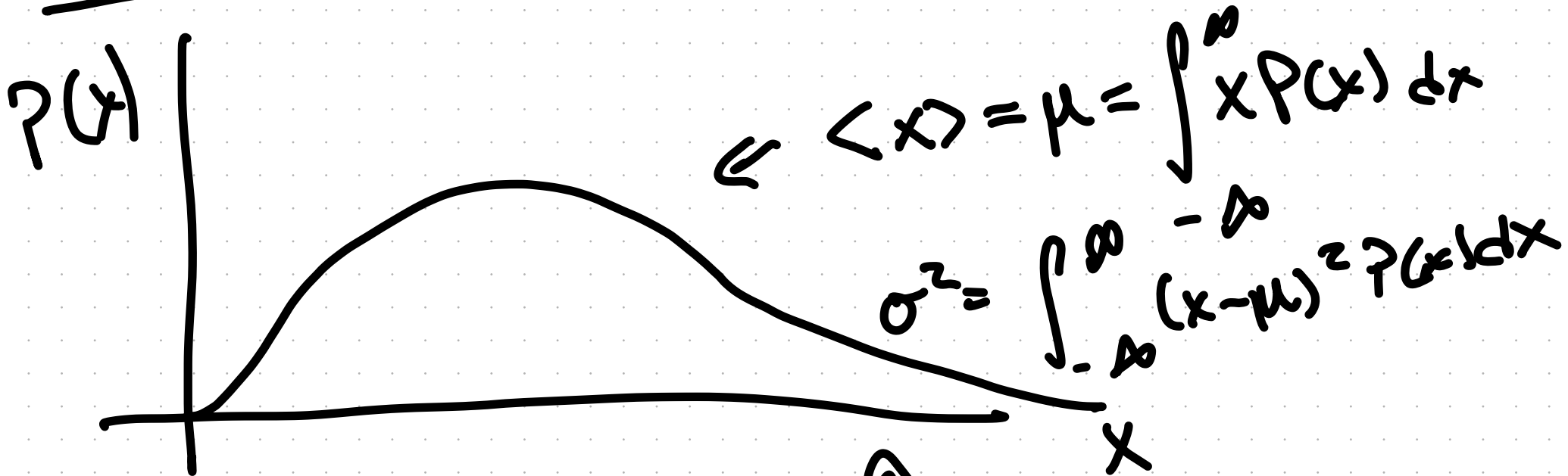


Q:  $f(x) = \int_a^x g(x) dx$        $C_{AB}(t) = \int_0^t A(t) B(0) dt$

$$\langle f(x) \rangle = \int_{-\infty}^{\infty} f(x) P(x) dx$$

$$\text{Var } f(x) = \int_{-\infty}^{\infty} (f(x) - \langle f(x) \rangle)^2 P(x) dx$$

# Lecture 3



$x_1, x_2, \dots, x_N$

sampled from  
this

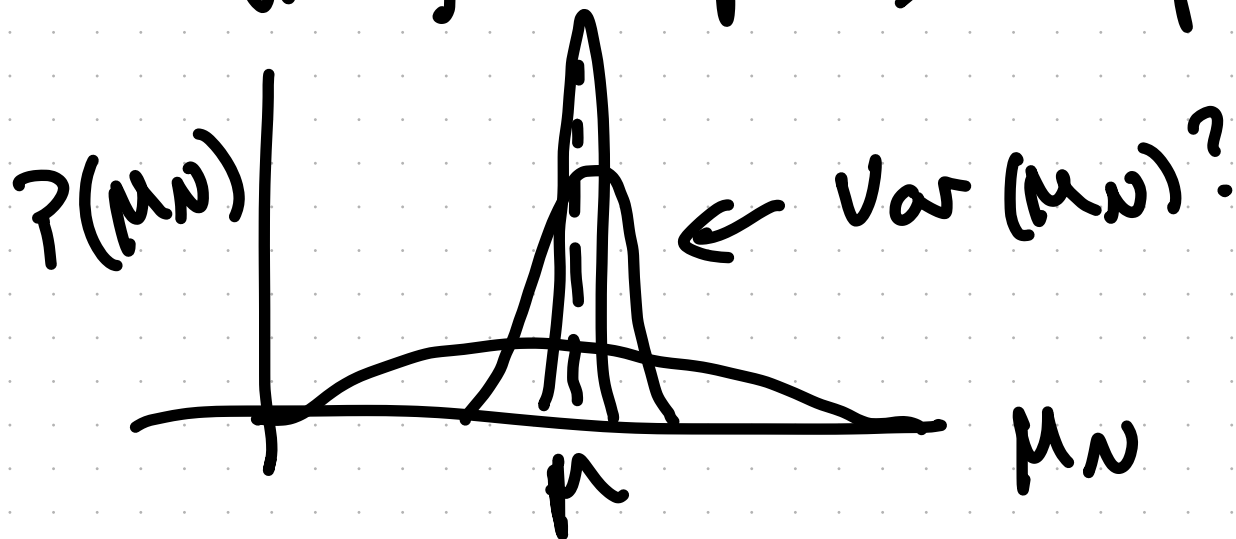
$$\mu_N = \frac{1}{N} \sum_{i=1}^N x_i$$

how close is  
 $\mu_N$  to  $\mu$

$\mu_N = \frac{1}{N} \sum_{i=1}^N x_i$       $x_i$  are  $\hat{}$  samples  
independent

$$\langle \mu_N \rangle = \frac{1}{N} \sum_{i=1}^N \langle x_i \rangle = \frac{1}{N} \sum_{i=1}^N \mu = \frac{N}{N} \mu = \mu$$

$$\text{Var}(\mu_N) = \langle \mu_N^2 \rangle - \langle \mu_N \rangle^2$$



$$\text{Var}(\mu_N) = \langle \mu_N^2 \rangle - \langle \mu_N \rangle^2$$

$\mu_N = \frac{1}{N} \sum_{i=1}^N x_i$  ?

$\langle \mu_N \rangle = \mu$

$$\begin{aligned} \langle \mu_N^2 \rangle &= \left\langle \left( \frac{1}{N} \sum_{i=1}^N x_i \right) \left( \frac{1}{N} \sum_{j=1}^N x_j \right) \right\rangle \\ &= \left\langle \frac{1}{N^2} \sum_{i=1}^N \sum_{j=1}^N x_i x_j \right\rangle = \frac{1}{N^2} \sum_{i,j} \langle x_i x_j \rangle \end{aligned}$$

$$\langle x_i x_j \rangle = \begin{cases} \langle x_i \rangle^2 & i=j \\ \langle x_i \rangle \langle x_j \rangle & i \neq j \end{cases} \leftarrow \text{independent}$$

$\underbrace{\langle x_i \rangle \langle x_j \rangle}_{\mu^2}$

$$\begin{aligned} \langle \mu^2 \rangle &= \frac{1}{N^2} \sum_{i=1}^N \sum_{j=1}^N \langle x_i x_j \rangle \\ &= \frac{1}{N^2} \sum_{i=1}^N \left( \langle x_i \rangle^2 + \underbrace{(N-1)\mu^2}_{N\mu^2 - \mu^2} \right) \\ &= \frac{1}{N^2} \sum_{i=1}^N \left( \underbrace{\langle x_i \rangle^2 - \mu^2}_{\text{Var}(x_i)} \right) + \underbrace{\frac{1}{N^2} \sum_{i=1}^N N\mu^2}_{= \mu^2} \end{aligned}$$

$$\text{Var}(\mu_N) = \langle \mu_N^2 \rangle - \mu^2$$

$$= \left[ \frac{1}{N^2} \sum_{i=1}^N \text{Var}(x_i) + \cancel{\mu^2} \right] - \cancel{\mu^2}$$

$$\sigma_N^2$$

$$= \frac{1}{N^2} \sum_{i=1}^N \sigma^2 = \frac{1}{N} \sigma^2$$

$$\sigma_N = \frac{\sigma}{\sqrt{N}}$$

"Standard error of the mean"

$\mu_N \rightarrow \mu$   
like  $\frac{1}{\sqrt{N}}$

# Central Limit Theorem

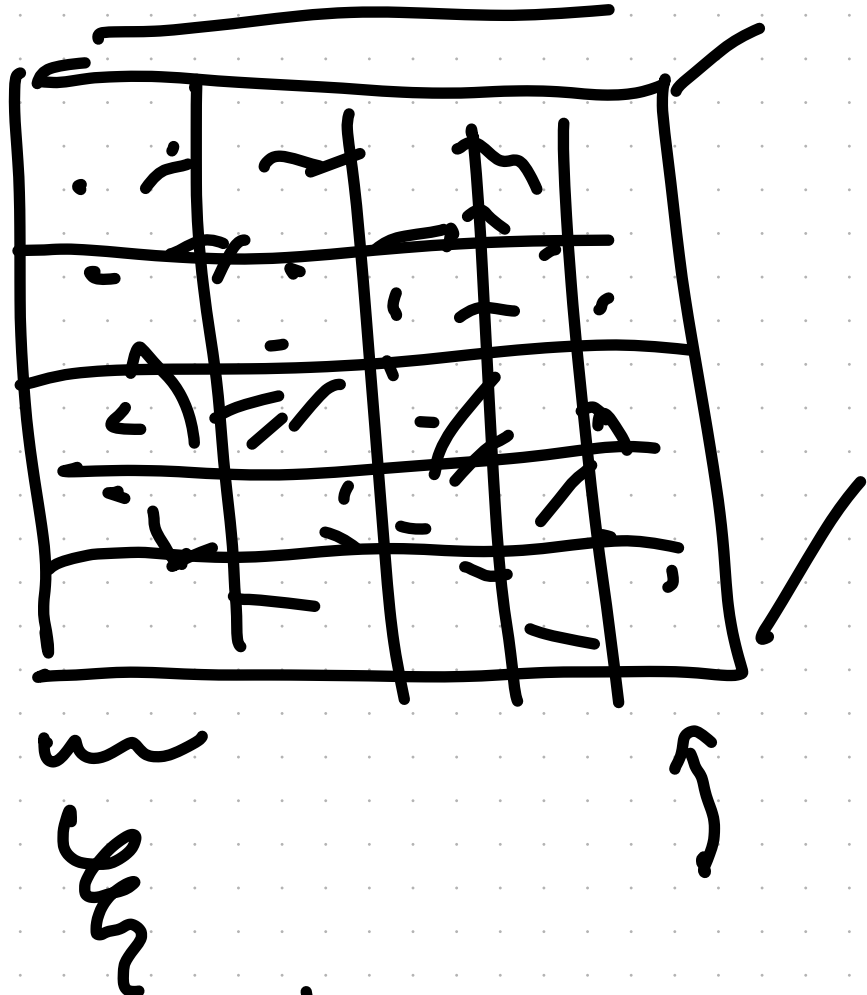
$x_i$  points sampled from  $P(x)$

$$\mu_N = \frac{1}{N} \sum_{i=1}^N x_i$$

$$P(\mu_N - \mu) \xrightarrow{N \rightarrow \infty} \mathcal{N}\left(\underset{\substack{\text{mean} \\ \text{zero}}}{0}, \underset{\text{var}}{\sigma^2/N}\right)$$
$$= \frac{1}{A} e^{-\frac{(\mu_N - \mu)^2}{2(\sigma^2/N)}}$$

# Statistical Mechanics

$$N_{\text{boxes}} = \frac{V}{\underbrace{\epsilon^d}_d}$$



Calculate  $A_i$  in each box

$\langle A_i \rangle = \langle A \rangle$  for system

$$\text{Var } A_i \sim \frac{1}{\sqrt{N_{\text{boxes}}}}$$



# Classical Mechanics

$$\vec{r} = (\vec{r}_1, \dots, \vec{r}_N)$$

$$\vec{v} = (\vec{v}_1, \dots, \vec{v}_N)$$

$$\vec{a} = \frac{d\vec{v}}{dt} = \frac{d^2\vec{r}}{dt^2}$$

$$\vec{a} = \ddot{\vec{r}} = \ddot{\vec{r}}$$

Newton's Equation says  $\vec{F} = m \vec{a}$

N differential equations

$$m_i \ddot{\vec{r}}_i = \vec{F}_i(\vec{r}_1, \dots, \vec{r}_N)$$

if we know  $\vec{r}(0)$ ,  $\vec{v}(0)$  and  $\vec{F}(\vec{r})$   
then everything is determined

If no friction, and know

$$U(\vec{r})$$

$$\vec{F}(\vec{r}) = -\nabla U(\vec{r}) \text{ ie}$$

$$F_i(\vec{r}) = - \frac{\partial U(\vec{r})}{\partial r_i} \quad (\text{no veloc.})$$

$$\nabla A = \left( \frac{\partial A}{\partial r_1}, \frac{\partial A}{\partial r_2}, \dots, \frac{\partial A}{\partial r_n} \right)$$

$$E(\vec{r}, \vec{v}) = \sum_i \underbrace{\frac{1}{2} m_i v_i^2}_{\text{kinetic}} + \underbrace{U(r)}_{\text{potential}}$$

Momenta  $\vec{p}_i = m_i \vec{v}_i$

If  $\vec{F} = -\nabla U$ , forces are  
"conservative"

total energy is conserved

$$\mathcal{E} = \frac{1}{2} m v^2 + U(r) = \frac{p^2}{2m} + U(r)$$

$$\frac{d\mathcal{E}}{dt} = \frac{1}{2} m (v\dot{v} + \dot{v}v) + \frac{dU(r)}{dt}$$

$$= m v a + \frac{dU(r)}{dt}$$

$$= v F + \frac{dU(r)}{dt} = v F + \left( \frac{\partial U}{\partial r} \right) \frac{dr}{dt} = 0$$

Chain rule  $\frac{dX}{dt} = \sum_i \dot{x}_i \left( \frac{\partial X}{\partial x_i} \right) \frac{\partial x_i}{\partial t} = \sum_i \dot{x}_i \frac{\partial X}{\partial x_i}$

$$\frac{d\varepsilon}{dt} = \vec{v} \cdot \vec{F} - (\nabla U) \cdot \vec{r}$$
$$= \vec{v} \cdot \vec{F} - \vec{F} \cdot \vec{v} = 0$$

in higher  
dimensions

Energy is conserved

# Lagrangian Mechanics

for conservative systems

$$L(\vec{r}, \dot{\vec{r}}) = K(\dot{\vec{r}}) - U(\vec{r})$$

Euler-Lagrange Equation

$$\frac{d}{dt} \left( \frac{\partial \mathcal{L}}{\partial \dot{r}_i} \right) - \frac{\partial \mathcal{L}}{\partial r_i} = 0$$

(Sec. 6)  
Tuckerm

$$K = \frac{1}{2} m \dot{r}_i^2$$

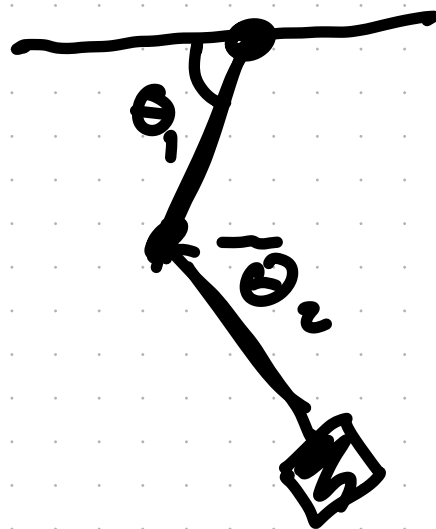
$$\frac{d}{dt} (m \dot{r}_i) = - \frac{\partial U}{\partial r_i}$$

$m \ddot{r}_i = F_i$

Why is this helpful?

Other coordinates

$$q_i = f_i(\vec{r})$$



# Hamiltonian Mechanics

$$H(\vec{q}, \vec{p})$$

in  $xyz$

$$\vec{p} = m\vec{v}$$

In general  $p_i = \frac{\partial \mathcal{H}}{\partial \dot{q}_i}$

$$K(\dot{q}_i) = \frac{1}{2} m \dot{q}_i^2$$

$$\frac{\partial \mathcal{H}}{\partial \dot{q}_i} = m \dot{q}_i = m v_i = p_i$$

total energy

$$H(\vec{q}, \vec{p}) = K + U(q), \quad K = \sum_{i=1}^n \frac{p_i^2}{2m}$$



total energy

$$H(\vec{q}, \vec{p}) = K + U(q), \quad K = \sum_{i=1}^n \dot{q}_i^2 / 2m$$

$$\dot{q}_i = \frac{\partial H}{\partial p_i}$$
$$p_i = - \frac{\partial H}{\partial q_i}$$

← equations of motion

$H$  and  $Q$  are connected  
by a "Legendre transform"

[Sec 1.5]

$$\begin{aligned} H(\vec{p}, \vec{q}) &= \sum_{i=1}^n \frac{\partial Q}{\partial q_i} \dot{q}_i - Q \\ &= \underbrace{\sum_{i=1}^n \dot{q}_i p_i}_{\vec{p} \cdot \vec{q}} - Q \end{aligned}$$

$H$  is total energy

$$\frac{dH}{dt} = \sum_{i=1}^N \left( \frac{\partial H}{\partial q_i} \right) \left( \frac{\partial q_i}{\partial t} \right) + \left( \frac{\partial H}{\partial p_i} \right) \left( \frac{\partial p_i}{\partial t} \right)$$

$$H(\vec{q}, \vec{p}) = \sum_{i=1}^N \left( -\dot{p}_i \right) q_i + \left( q_i \right) \dot{p}_i$$

$$= \sum_{i=1}^N [q_i \dot{p}_i - \dot{p}_i q_i]$$
$$= 0$$