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Wednesday 880.05

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On left board (for reference):

"anharmonic oscillator"

$$\hat{H}(\hat{p}, \hat{x}) = :\hat{H}(\hat{p}, \hat{x}): = \frac{\hat{p}^2}{2m} + V(\hat{x}) \quad \text{example } \frac{\hat{p}^2}{2m} + \frac{1}{2}a\hat{x}^2 + \frac{1}{4}\lambda\hat{x}^4$$

(note: this will look like a function field theory examples, but very different since long-distance confining forces)

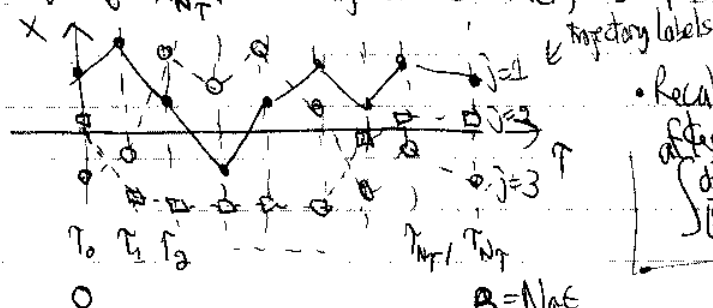
$$Z(\beta) = \text{Tr } e^{-\beta \hat{H}} = \int dx \langle x | e^{-\beta \hat{H}} | x \rangle$$

$$= \int dx \int_{x(0)=x}^{x(\beta)=x} \mathcal{D}(x(\tau)) e^{-\frac{1}{\hbar} \int_0^\beta d\tau \left[\frac{m}{2} \left(\frac{dx(\tau)}{d\tau} \right)^2 + V(x(\tau)) \right]}$$

$$= \int \mathcal{D}[x(\tau)] e^{-\frac{1}{\hbar} \int_0^\beta d\tau \left[\frac{m}{2} \left(\frac{dx(\tau)}{d\tau} \right)^2 + V(x(\tau)) \right]} = \int \mathcal{D}[x(\tau)] e^{-S_E[x(\tau)]/\hbar}$$

$$\left[\int \mathcal{D}x_0 \mathcal{D}x_1 \mathcal{D}x_2 \dots \mathcal{D}x_{N_T} \right] = \int \prod_{k=1}^{N_T} dx_k \left(\frac{m}{2\pi\hbar\epsilon} \right)^{3N_T/2} e^{-\frac{\epsilon}{\hbar} \sum_{i=1}^{N_T} \left[\frac{m}{2} \left(\frac{x_i - x_{i-1}}{\epsilon} \right)^2 + V(x_{i-1}) \right]}$$

where $\beta = N_T \epsilon$ (Monday N_T was M but notation is too confusing)
and $x_0 = x_{N_T}$. The trajectories $x(\tau) \rightarrow \{x(\tau_i)\}$ begin and end at same x



Recall that the $\frac{1}{2m}$ went away after

$$\int \frac{dx}{2\pi\hbar} e^{-\frac{\epsilon p^2}{2m} + i p_0(x_n - x_{n-1})} = \left(\frac{m}{2\pi\hbar\epsilon} \right)^{1/2} e^{-\frac{\epsilon p_0^2}{2m}}$$

can be absorbed in defining Z or hidden in \mathcal{D} .

Note: Using $\tau \in [p_1, p_2]$ instead of $\tau \in [0, \beta]$ is often simpler.

Plan for today:

- ① Quick comments on MATLAB PS#1
- ② Recap of intro to stochastic evaluation of path integrals
- ③ Brief follow-ups to one-particle path integral
 - spectral decomposition, perturbation theory, correlation functions
- ④ Generalization to many-particle states — (anti)symmetrization
 - issues for numerical evaluation

Monday → ⑤ Intro to alternative based on coherent states and field operators

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MATLAB PS#2 Comments [any questions on diagrams?]

- Well executed solutions - we'll build on that when possible, (Eg. MATLAB code to illustrate SVM for PS#2.)

- Most (if not all) of you encountered an error evaluating $\logm(A)$ when A had real negative eigenvalue.

"Principal matrix logarithm not defined for A with nonpositive real eigenvalues. A non-principal matrix log is returned."

- let's remember what is going on in complex plane

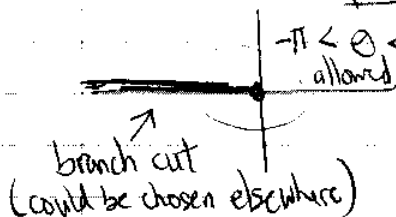


\mathbb{C} polar form $z = re^{i\theta}$

$$\Rightarrow \ln z = \ln r + i\theta = \ln r + i\theta$$

branch point at $z=0$, multivalued as $\theta \rightarrow 0 \rightarrow 2\pi \rightarrow 4\pi \rightarrow \dots$ keeps changing

So define principal logarithm by excluding negative real axis:



Not defined here for $\theta = \pi$ or $\theta = -\pi$ (although typically $-\pi < \theta \leq \pi$ is used)
 ie is $\logm(-1) = \ln 1 + i\pi$ or $-i\pi$ (chooses $+i\pi$)
 matrix

For a matrix, runs into this if eigenvalues lie on branch cut,

- MATLAB for loops

$$M = \text{randn}(5) + i * \text{randn}(5)$$

$$\text{or } M = 3 / \text{factorial}(6)$$

one way: $\text{eye}(5) + M + M \times M / 2 + M \times M \times M / 6 + \dots$

another way:

$$\text{sum} = 0$$

$$\text{for } i = 0:10$$

$$\text{sum} = \text{sum} + M^i / \text{factorial}(i)$$

end

$$\text{disp}(\text{sum});$$

print ('display') final result

exclude ; to print each intermediate answer

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• Recap & intro to stochastic evaluation of path integrals

• See Jaquín Dant notes.

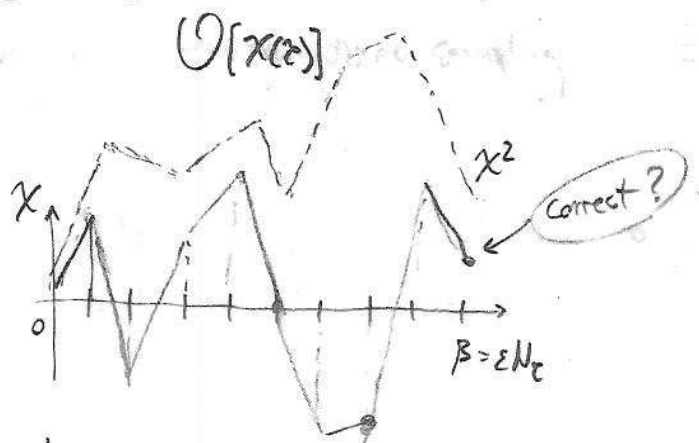
Recap from Monday

(I) Numerical (in particular stochastic) methods are necessary to determine the properties of interacting many-body quantum systems, because pert. theory and saddle point approximations can fail (and they do in many interesting cases).

(II) Stochastic methods use the path integral formulation of QM. This means that expectation values are computed like so:

$$\langle O \rangle = \frac{1}{Z} \int_{X(0)=X(\beta)} D[X(z)] e^{-S_E[X(z)]}$$

• Example: $O = \frac{1}{\beta} \int_0^\beta X^2(z) dz$



(For each trajectory we compute $\frac{1}{\beta} \int_0^\beta X^2(z) dz$ and add it with weight $\frac{1}{Z} \exp\{-S_E[X(z)]\}$)

Get notation right!

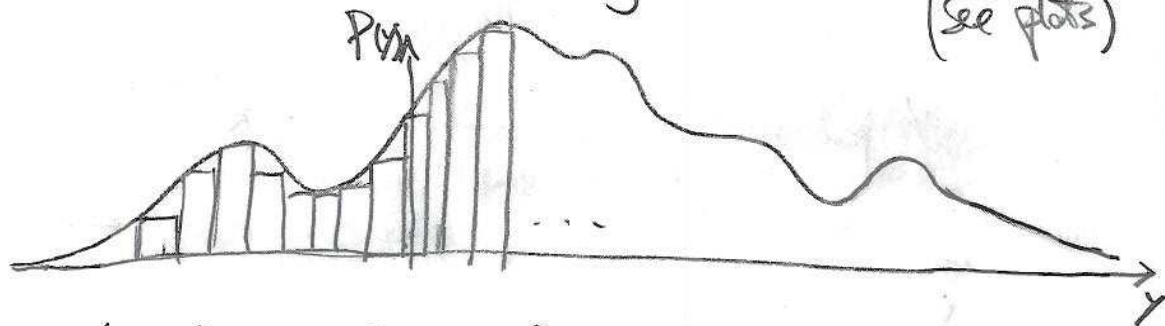
III Stochastic methods are based on the central limit theorem

Example:

Consider y distributed according to $P(y)$ (see plots)

$$P(y) > 0$$

$$\int P(y) dy = 1$$



Drawing samples of y according to $P(y)$ and putting them in bins according to their value will eventually reproduce $P(y)$ as a histogram.

Now consider $\langle O \rangle = \int dy O(y) P(y) = \text{"True mean"}$

$$\sigma^2 = \langle O^2 \rangle - \langle O \rangle^2 = \text{"True standard deviation"} \text{ (squared)}$$

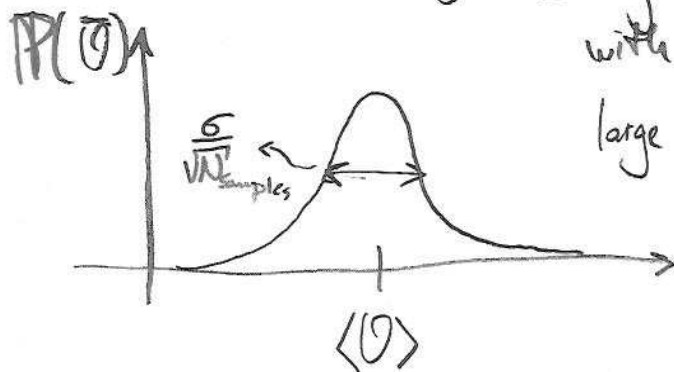
Question: Can I approximate $\langle O \rangle \approx \frac{1}{N_{\text{samples}}} \sum_{n=1}^{N_{\text{samples}}} O(y_n) \equiv \bar{O}$

taking y distributed according to $P(y)$?

(see plots)

CLT: Yes!

\bar{O} is "gaussianly distributed" around $\langle O \rangle$ (True mean) with std deviation $\frac{\sigma}{\sqrt{N_{\text{samples}}}}$, in the limit of large N_{samples} .



$$\Rightarrow \langle O \rangle \approx \bar{O} \pm \frac{\sigma}{\sqrt{N_{\text{samples}}}}$$

We don't know the true $\langle O \rangle$, so this theorem tells us how to estimate it.

We also don't know the true σ , but we can estimate it as

$$\sigma \approx \overline{O^2} - \bar{O}^2$$

In QM, where $\langle O \rangle = \frac{1}{Z} \int \mathcal{D}[x(\tau)] e^{-S_E[x(\tau)]} O[x(\tau)]$

$$= \int \mathcal{D}[x(\tau)] O[x(\tau)] P[x(\tau)]$$

We can estimate

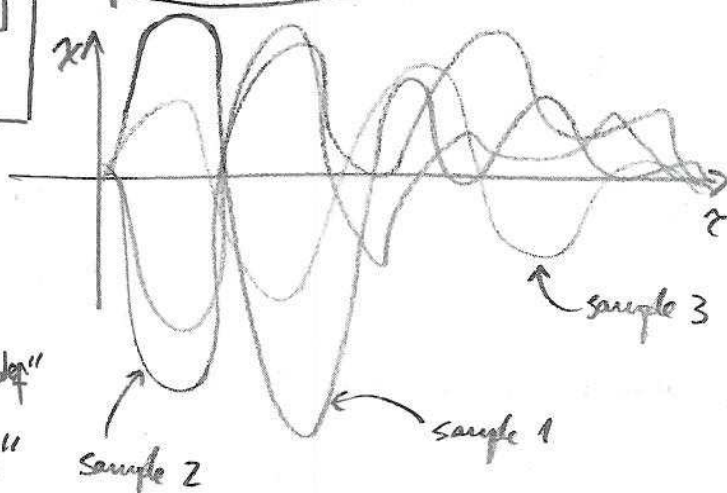
$$\langle O \rangle \approx \bar{O} = \frac{1}{N_{\text{samples}}} \sum_{n=1}^{N_{\text{samples}}} O[x_n(\tau)]$$

$$P[x(\tau)] = \frac{e^{-S_E[x(\tau)]}}{Z}$$

Remember: This works only if

P is well defined as a probability: $P[x] \geq 0$ "positive semidef"

$$\int P[x] dx = 1 \quad \text{"normalizable"}$$



Main problem: How to generate configurations distributed according to $P[x(\tau)]$.

→ Sampling strategies.

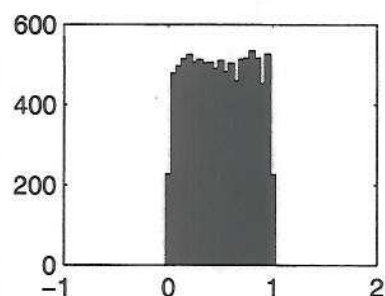
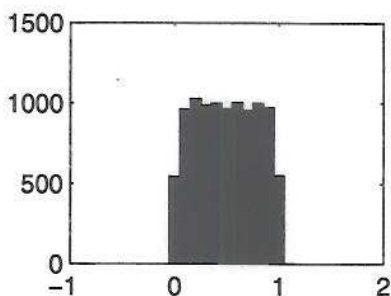
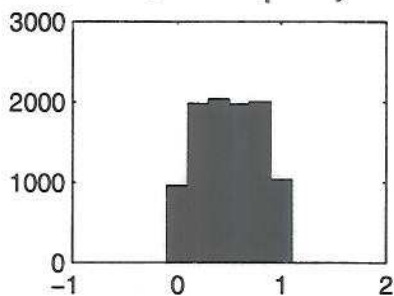
IV Sampling Strategies

- Heat-bath
 - Metropolis ← typical for Ising model
 - Molec. dynamics
 - Hybrid
- } — Markov chain-based

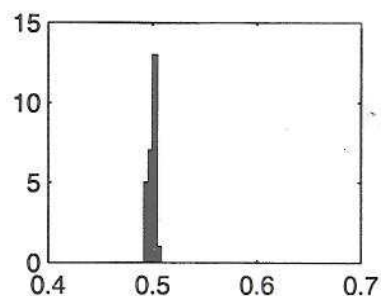
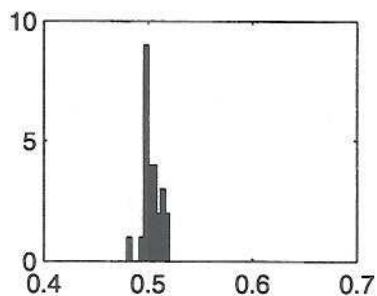
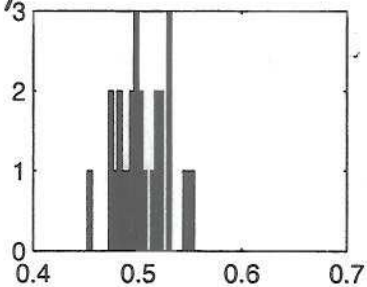
All of them can generate good samples if used properly.

- Equilibrated
- Uncorrelated

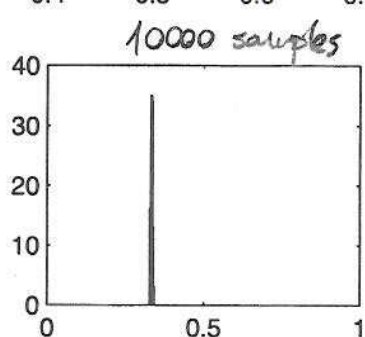
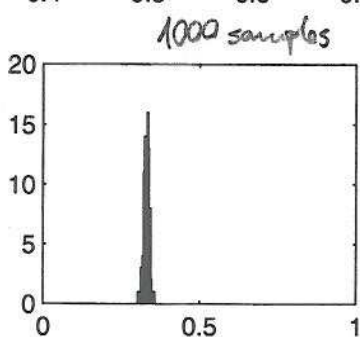
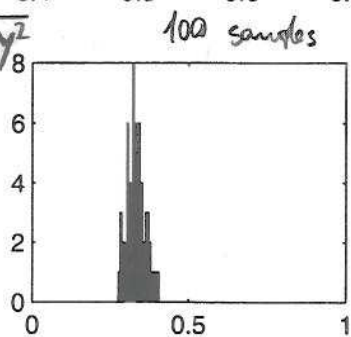
Distribution $P(y)$ (Uniform)



Distrib for \bar{y}_3



Distrib for \bar{y}_8

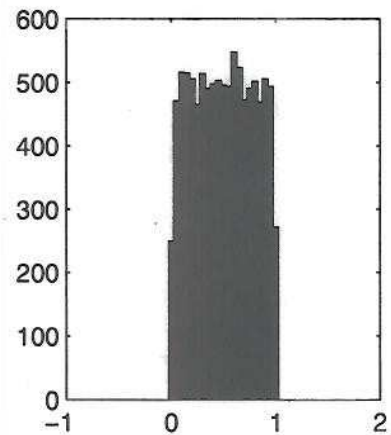
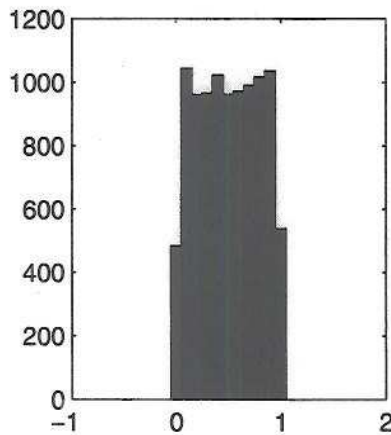
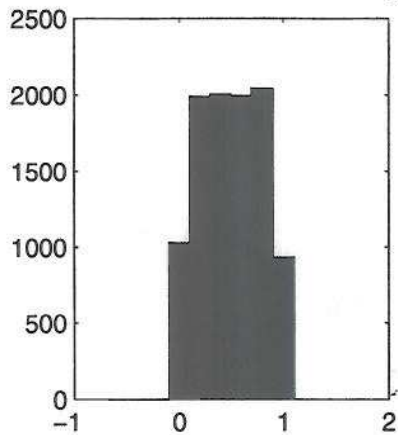


100 samples

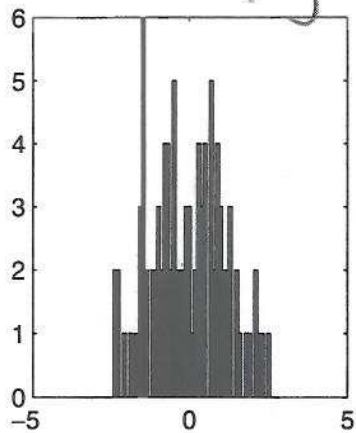
1000 samples

10000 samples

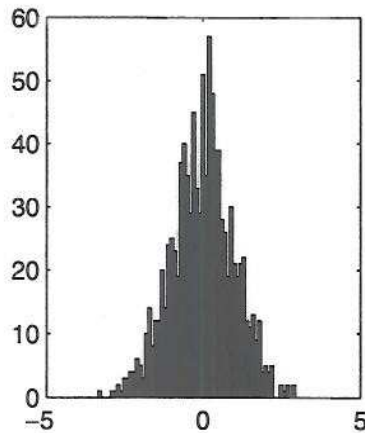
Uniform distribution sampling (10000 samples, variable bin size)



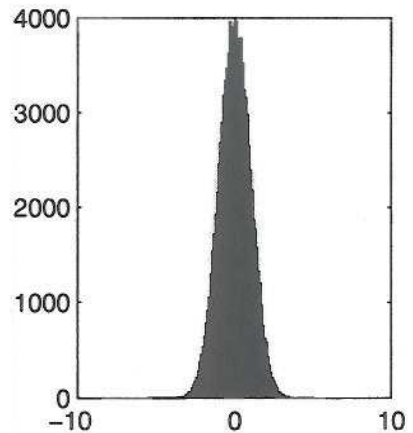
Gaussian distrib. sampling



100 samples



1000 samples



10000 samples

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Followup comments on Z for one particle...

The "real time" version comes from $\tau \rightarrow it$ (or $t \rightarrow -i\tau$) so that

$$U_E(x_f, \tau_f; x_i, \tau_i) = \langle x_f | e^{-\frac{\hat{H}}{\hbar}(\tau_f - \tau_i)} | x_i \rangle \rightarrow U(x_f, t_f; x_i, t_i) = \langle x_f | e^{-\frac{i\hat{H}}{\hbar}(t_f - t_i)} | x_i \rangle$$

Recall that $i\hbar \frac{\partial}{\partial t} \psi(x, t) = \hat{H} \psi(x, t) \Rightarrow \psi(x, t) = e^{-\frac{i\hat{H}(t - t_i)}{\hbar}} \psi(x, t_i)$ for any x

What about the physics content? Do we lose anything going from U to U_E ?

Consider eigenstates of \hat{H} : $[\hat{H}|\psi_n\rangle = E_n|\psi_n\rangle$ and $\langle x|\psi_n\rangle = \psi_n(x)$

$$\Rightarrow U(x_f, t_f; x_i, t_i) = \sum_n \langle x_f | \psi_n \rangle \langle \psi_n | e^{-\frac{i\hat{H}(t_f - t_i)}{\hbar}} | x_i \rangle = \sum_n \psi_n(x_f) \psi_n^*(x_i) e^{-\frac{iE_n(t_f - t_i)}{\hbar}}$$

$$\Rightarrow U_E(x_f, \tau_f; x_i, \tau_i) = \sum_n \langle x_f | \psi_n \rangle \langle \psi_n | e^{-\frac{\hat{H}(\tau_f - \tau_i)}{\hbar}} | x_i \rangle = \sum_n \psi_n(x_f) \psi_n^*(x_i) e^{-\frac{E_n(\tau_f - \tau_i)}{\hbar}}$$

"Spectral representation" shows same content. (that doesn't mean there can't be subtle complications in relating real and imaginary path integrals!)

Note: $\tau_i = 0, \tau_f = \beta$ and $\beta \rightarrow \infty$ limit

$$U_E(x_f, \beta; x_i, 0) \xrightarrow{\beta \rightarrow \infty} \psi_0(x_f) \psi_0^*(x_i) e^{-\beta E_0 / \hbar}$$

$$\text{if } E_1 - E_0 > 0 \quad \text{so } -\frac{1}{\beta} \ln U_E \xrightarrow{\beta \rightarrow \infty} E_0 - \frac{1}{\beta} \ln(\psi_0(x_f) \psi_0^*(x_i))$$

in particular $\Rightarrow -\frac{1}{\beta} \ln Z \xrightarrow{\beta \rightarrow \infty} E_0$

project ground state by $\beta \rightarrow \infty$ limit

independent of x_f, x_i so fine for Z

Note that $\tau \rightarrow it$ means

$$S_E[x(\tau)] \rightarrow -S[x(t)] = -\int_{t_i}^{t_f} dt \left[\frac{1}{2} m \dot{x}^2 - V(x(t)) \right]$$

usual action functional

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How does perturbation theory work in this case? [Note: See Zinn-Justin ch. 2 for more general U_6 instead of Z]

In the model partition function $Z \Rightarrow Z_j = \int dx e^{-\left(\frac{1}{2} p^2 + \frac{1}{4} x^4\right) + jx}$ [used $\frac{1}{4} x^4$]

Idea: analog is to add $f(\tau)$ to it in path integral:

$$H(p, x) \Rightarrow H(p, x; t) = \frac{1}{2m} p^2 + V(x) - x f(\tau)$$

This is a time dependent driving (external) force \Rightarrow like a j at each time.
Also $f(t) = f(b)$.

For concreteness, take $V(x) = \frac{1}{2} a x^2 + \frac{1}{4} x^4$ and think about perturbation $\frac{1}{4} x^4$ about harmonic oscillator $\frac{p^2}{2m} + \frac{1}{2} a x^2$.

In discrete version of $Z[f]$, $f(\tau) \rightarrow \{f(\tau_i)\} \equiv \{f_i\}$, $i = 1, \dots, N_T$.

\Rightarrow $Z[f] = C \int \prod_k dx_k e^{-\sum_{i=1}^{N_T} \left[\frac{m}{2} (x_i - x_{i-1})^2 + \frac{1}{2} a x_i^2 + \frac{1}{4} x_i^4 - x_i f_i \right]}$ (like having a different j at each time step)

So $\frac{\delta Z[f]}{\delta f_i} = C \int \prod_k dx_k x_i e^{-\sum_{i=1}^{N_T} \left[\dots \right]}$ \leftarrow one of these has $i=j$
 $f(\tau_k) \rightarrow \frac{\delta f_j}{\delta f_i}$ a particular j

so $\langle x_{i_1} x_{j_2} \rangle = \frac{1}{Z} \left(\frac{\delta}{\delta f_{i_1}} \frac{\delta}{\delta f_{j_2}} \right) Z[f] \Big|_{f=0}$ "correlation function"

continuum limit $N_T \rightarrow \infty$ $\langle x(\tau_1) x(\tau_2) \rangle = \frac{1}{Z} \left(\frac{\delta}{\delta f(\tau_1)} \frac{\delta}{\delta f(\tau_2)} \right) Z[f] \Big|_{f=0}$ general case $\langle \{ \}^2 \rangle$ from model partition function

[Note $\langle x(\tau_1) x(\tau_2) \rangle = \langle T \hat{x}(\tau_1) \hat{x}(\tau_2) \rangle$ where $\hat{x}(\tau) = e^{-H\tau} x e^{H\tau}$]

\leftarrow "functional derivatives"

Check $\frac{\delta}{\delta f(\tau)} \int d\tau' A(\tau') g(\tau') = \frac{1}{\epsilon} \frac{\delta}{\delta f_i} \sum_k \epsilon f_k g_k = g_i = g(\tau)$ so makes an integral go away!

* What if just $\frac{\delta}{\delta f(\tau)} f(\tau)? \rightarrow \frac{1}{\epsilon} \frac{\delta}{\delta f_i} f_k = \frac{1}{\epsilon} \delta_{ik} \rightarrow \delta(\tau - \tau')$

(Check that $\int \delta(\tau - \tau') d\tau' = 1$ translates $\rightarrow \sum_{k=1}^{N_T} \epsilon \left(\frac{1}{\epsilon} \delta_{ik} \right) = \sum_k \delta_{ik} = 1 \checkmark$)

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Now we can also play the same game as before where we replace

$$x_i^4 \rightarrow \left(\frac{1}{\epsilon} \frac{\partial}{\partial f_i} \right)^4$$

so

$$Z[f] = e^{-\epsilon \sum_{i=1}^{N_f} \frac{1}{4} \left(\frac{1}{\epsilon} \frac{\partial}{\partial f_i} \right)^4} C \int \prod dx_k e^{-\epsilon \sum_{i=1}^{N_f} \left[\frac{m}{2} \left(\frac{x_i - x_{i-1}}{\epsilon} \right)^2 + \frac{1}{2} a x_i^2 - x_i f_i \right]}$$

continuum \Rightarrow

$$e^{-\int_0^\beta dt \frac{1}{4} \left(\frac{\partial}{\partial f(t)} \right)^4} C \int_{x(0)=x(0)}^{\mathbf{x}=\mathbf{x}_0} \mathcal{D} x(t) e^{-\int_0^\beta dt \left[\frac{m}{2} \left(\frac{dx}{dt} \right)^2 + \frac{1}{2} a x^2 - x f(t) \right]}$$

S_E

The discrete form of the integral is a Gaussian form with matrices. If we look up the general result (see Zinn-Justin chaps. 1 handout):

$$I[J] \equiv \int_{-a}^a dy_1 \cdots dy_n e^{-\frac{1}{2} y_i A_{ij} y_j + y_i J_i} = (2\pi)^{n/2} [\det A]^{-1/2} e^{\frac{1}{2} J_i A_{ij}^{-1} J_j}$$

If $J=0$, then $I[0] = (2\pi)^{n/2} [\det A]^{-1/2} \Rightarrow I[J] = I[0] e^{\frac{1}{2} J^T A^{-1} J}$

If we know $I[0]$ from elsewhere, we are done. (We'll assume that here.)

So what is J_i and y_i here? $y_i \rightarrow x_i$, $J_i \rightarrow f_i$
What is A_{ij} ? It is the quadratic part: so match $x_i A_{ij} x_j$ to terms in the exponent with two x_i 's.

$$-\frac{1}{2} x_i A_{ij} x_j = \begin{pmatrix} x_1 & x_2 & x_3 \\ -\frac{1}{2} x \end{pmatrix} \begin{pmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = -\frac{1}{2} (A_{11} x_1^2 + A_{12} x_1 x_2 + A_{21} x_2 x_1 + A_{22} x_2^2 + \dots)$$

$$-\epsilon \sum_{i=1}^{N_f} \frac{1}{2} a x_i^2 = -\frac{1}{2} \epsilon a [x_1^2 + x_2^2 + x_3^2 + \dots] \Rightarrow (A_{ij})_{\text{this term}} = \epsilon a \delta_{ij} = \begin{pmatrix} \epsilon a & 0 \\ 0 & \epsilon a \end{pmatrix}$$

$$-\epsilon \sum_{i=1}^{N_f} \frac{m}{2} \left(\frac{x_i - x_{i-1}}{\epsilon} \right)^2 = -\frac{1}{2} \frac{\epsilon m}{\epsilon^2} [(x_1 - x_0)^2 + (x_2 - x_1)^2 + (x_3 - x_2)^2] = -\frac{1}{2} \frac{m}{\epsilon} [x_1^2 - x x_0 x_1 + x_0^2 + x_2^2 - x_1 x_2 - x_2 x_1 + x_1^2 + x_3^2 - x_2 x_3 - x_3 x_2 + x_2^2 + \dots]$$

$$\Rightarrow A_{11} = \frac{m}{\epsilon} + \epsilon a, A_{12} = -\frac{m}{\epsilon}, A_{21} = -\frac{m}{\epsilon}, A_{22} = \frac{m}{\epsilon} + 2\epsilon a + \dots$$

$A_{10} = A_{3N_f} = -1 = A_{N_f+1}$

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note the boundary conditions

or

$$\underline{A} = \frac{m}{\epsilon} \begin{pmatrix} 2 & -1 & 0 & 0 & -1 \\ -1 & 2 & -1 & & \\ 0 & -1 & 2 & & \\ & & & \ddots & \\ -1 & & & & 2 \end{pmatrix} + \epsilon a \begin{pmatrix} 1 & & & & \\ & 1 & & & \\ & & 1 & & \\ 0 & & & \ddots & \\ & & & & 1 \end{pmatrix}$$

so we can construct \underline{A} explicitly in the discrete representation.
Then we have

$$Z[F] = e^{-\epsilon \sum_i \frac{1}{4} \left(\frac{\Delta x}{\Delta t} \right)^4} z_0 e^{\frac{1}{2} F_i A_{ik}^{-1} F_k}$$

How do we understand the continuum version?

$$\int_{x(0)=x_0}^{\infty} x(\tau) e^{-\int_0^{\beta} d\tau \left[\frac{m}{2} \left(\frac{dx}{d\tau} \right)^2 + \frac{1}{2} a x^2 - x f(\tau) \right]}$$

$$= \int_{x(0)=x_0}^{\infty} x(\tau) e^{-\int_0^{\beta} d\tau \left\{ x \left[-\frac{m}{2} \frac{d^2}{d\tau^2} + \frac{1}{2} a \right] x - x f(\tau) \right\}} \quad \text{(What about surface term? Remember boundary condition, surface term = 0)}$$

$$= \int_{x(0)=x_0}^{\infty} x(\tau) e^{-\int_0^{\beta} d\tau \int_0^{\beta} d\tau' \left\{ x(\tau) \left[-\frac{m}{2} \frac{d^2}{d\tau^2} + \frac{1}{2} a \right] \delta(\tau-\tau') x(\tau') - x(\tau) \delta(\tau-\tau') f(\tau') \right\}}$$

$$= \int_{x(0)=x_0}^{\infty} x(\tau) e^{-\int_0^{\beta} d\tau \int_0^{\beta} d\tau' \frac{1}{2} (x + f A^{-1}) A (x + f A^{-1}) + \frac{1}{2} \int_0^{\beta} d\tau \int_0^{\beta} d\tau' A(\tau) A^{-1}(\tau, \tau') f(\tau')}$$

$$\Rightarrow Z[F] = e^{-\int_0^{\beta} d\tau \frac{1}{4} \left(\frac{\Delta x}{\Delta t} \right)^4} z_0 e^{\frac{1}{2} \int_0^{\beta} d\tau \int_0^{\beta} d\tau' A(\tau) A^{-1}(\tau, \tau') f(\tau')}$$

$\Rightarrow A^{-1}(\tau, \tau')$ is the inverse of the differential operator $\left(-\frac{m}{2} \frac{d^2}{d\tau^2} + \frac{1}{2} a \right) \delta(\tau-\tau')$ or the Green's function, with the periodic boundary condition.

\Rightarrow solution to $\left(-\frac{m}{2} \frac{d^2}{d\tau^2} + \frac{1}{2} a \right) A^{-1}(\tau) = \delta(\tau)$ with $A^{-1}(0) = A^{-1}(\beta)$

(we've used $A^{-1}(\tau, \tau') = A^{-1}(\tau - \tau')$.)

• Can you solve this?

Next week we'll look at diagrams:



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How do we generalize to a few- or many-body system?

• For the quantum mechanics approach with Hamiltonian:

$$\hat{H} = \sum_{i=1}^N \frac{(\hat{p}_i)^2}{2m} + \frac{1}{2} \sum_{i \neq j}^N V(\hat{x}_i - \hat{x}_j) + (3\text{-body } V) + \dots$$

For N particles with identical masses

• Since they are identical particles, the partition function has to be a sum (trace) over a complete basis that has the correct symmetry (bosons - symmetric, fermions - antisymmetric).

• So going from $|x\rangle$ to $|x^{(1)}\rangle |x^{(2)}\rangle \dots |x^{(N)}\rangle$

as a direct product seems problematic because it is not a definite symmetry.

[Note: The indices written as superscripts here label the different particles. These are typically written as x_1, x_2, \dots, x_N . However, this directly clashes with our use of subscripts to indicate the x 's at different time steps. Negele and Orland just use the same notation for both!]

↑
(note, actually it does work if the states in the trace are symmetrized)

(note the curly bracket)

• So we use

$$\{|x^{(1)}\rangle |x^{(2)}\rangle \dots |x^{(N)}\rangle\} \equiv \frac{1}{N!} \sum_P \{^P |x^{(1)}\rangle |x^{(2)}\rangle |x^{(3)}\rangle \dots |x^{(N)}\rangle\}$$

where the P means a permutation of the particles and $\{^P$ fixes up bosons ($\{^P = 1$, all $\{^P$ have same sign) and fermions ($\{^P = \pm 1$ as P is even/odd).

The completeness relation is

$$\frac{1}{N!} \sum_{x^{(1)} \dots x^{(N)}} |x^{(1)}\rangle \dots |x^{(N)}\rangle \langle x^{(1)} \dots x^{(N)}| = 1$$

• More generally, let $x \rightarrow \infty$ represent a single-particle basis.
• This is not a normalized basis as yet, but that is not important for us.

10/14/09

• At this stage we simply need this basis to define the trace:

$$Z = \frac{1}{N!} \int \prod_{i=1}^N dx^{(i)} \{ x^{(1)} \dots x^{(N)} | e^{-\beta \hat{H}} | x^{(1)} \dots x^{(N)} \}$$

$$= \frac{1}{N!} \sum_P \{ x^{(P1)} \dots x^{(PN)} | e^{-\beta \hat{H}} | x^{(1)} \dots x^{(N)} \}$$

Just as before, the "time interval" from 0 to β can be broken into N_T pieces and we insert complete sets of states.

• We can actually use either (anti)symmetrized or just product states. In the latter case, the final states of correct symmetry imposes the correct Fermi or Bose statistics.

• In the direct product case,

$$Z = \frac{1}{N!} \sum_P \int \prod_{i=1}^N [\psi(x^{(i)}_1) \dots \psi(x^{(i)}_{N_T})] e^{-\int_0^\beta d\tau \left[\sum_{i=1}^N \left(\frac{dx^{(i)}(\tau)}{d\tau} \right)^2 + \frac{1}{2} \sum_{i,j} V(x^{(i)}(\tau) - x^{(j)}(\tau)) \right]}$$

$x^{(1)}_{N_T} = x^{(1)}_1$
 \vdots
 $x^{(N)}_{N_T} = x^{(N)}_1$

When we simulate fermions with this form, $e^{-\beta \hat{H}}$ filters out the lowest state of any symmetry \Rightarrow but the lowest symmetric will be lower than the antisymmetric \Rightarrow noise growing exponentially before projecting on antisymmetric states.

• Alternative is to project at each ϵ step:

$$\{ x^{(1)} \dots x^{(N)} | e^{-\epsilon \hat{H}} | y^{(1)} \dots y^{(N)} \} = \left(\frac{m}{2\pi\epsilon} \right)^{N/2} \frac{1}{Z(\epsilon)} e^{-\frac{m}{2\epsilon} \sum_i (x_i - y_i)^2 - \frac{\epsilon}{2} \sum_{i,j} V(y_i - y_j)}$$

$$= \left(\frac{m}{2\pi\epsilon} \right)^{N/2} \text{Det } M e^{-\frac{m}{2\epsilon} \sum_i (x_i - y_i)^2 - \frac{\epsilon}{2} \sum_{i,j} V(y_i - y_j)}$$

with $M_{ij} = e^{-\frac{m}{2\epsilon} [(x_j - y_i)^2 - (x_i - y_j)^2]}$ but Det M has minus signs! More later on why this is bad.

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10/11/09

Instead of continuing with this basis, we are going to switch to a path integral over fields, by considering a different complete set of states to use, which are built from the occupation number basis.

- This gives rise to Hamiltonians written in terms of creation and destruction operators, a^\dagger, a
- Then the states we want are eigenstates of H (instead of \hat{p}, \hat{x} ^{eigen} states)
⇒ coherent states

- So let's do a bit of grd quantized formalism!