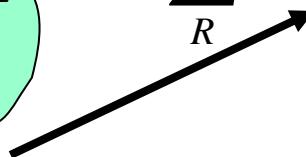
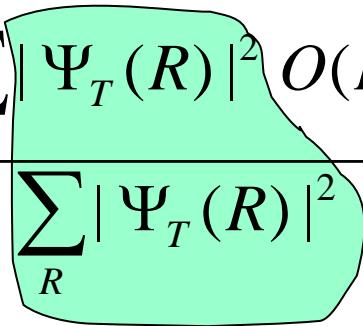


TRIAL MANY BODY WAVE FUNCTION

$$\Psi_T(r_1, r_2, \dots, r_N) \equiv \Psi(R)$$

$$\langle \hat{O} \rangle = \frac{\langle \Psi_T | \hat{O} | \Psi_T \rangle}{\langle \Psi_T | \Psi_T \rangle} = \frac{\sum_R |\Psi_T(R)|^2 O(R)}{\sum_R |\Psi_T(R)|^2} = \sum_R P(R) O(R)$$


**Many interacting particles
Multi dimensional integrals**

Classical Stat Mech

$$H = -J \sum_{\langle ij \rangle} S_i S_j \quad S_i = \pm 1$$

N sites: each site can be \uparrow or \downarrow $= 2^N$ configurations

$$Z = \text{Tr} e^{-\beta H} = \sum_{\{C\}} e^{-\beta E(C)}$$

$$E = \frac{1}{Z} \sum_{\{C\}} e^{-\beta E(C)} E(c)$$

HOW CAN WE EVALUATE MULTIDIMENSIONAL INTEGRALS??

USE MONTE CARLO METHODS

Evaluating Multi-dimensional integrals

Speed

Monte Carlo is an efficient way to evaluate integrals
For large numbers of dof

Quadrature method: 10 divisions
Nparticles=8 dimensions d=2

Evaluate integrand at 10^{2*8} points

Computation speed= 10 MFLOPS = 10^7 Evaluations/sec

Integrand evaluation takes $\sim 10^9$ sec ~ 40 years

Monte Carlo < 1 minute for 1% accuracy

ACCURACY

of points at which integrand evaluated=M

degrees of freedom=f=d* Nparticles

Error: $\sigma_{quad} \sim M^{-2/f}$

$\sigma_{MC} \sim M^{-1/2}$

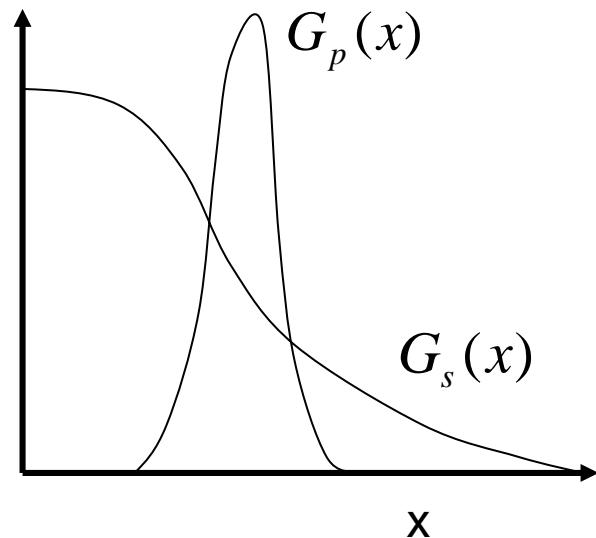


MC more efficient for $f = dN \geq 4$

Evaluating integrals

$$I = \int_0^1 dx f(x) = \int_0^1 dx G_p(x) G_s(x)$$

$$G(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{\frac{-(x-a)^2}{2\sigma^2}}$$



Monte Carlo I:

(a) Pick x at random in the range $[0, 1]$

(b) Evaluate $f(x) = G_p(x) G_s(x)$

at the randomly picked x

$$I = \sum_{x_i} f(x_i)$$

CODES PROVIDED

Exact result: exact.f

Quadrature method: quad.f

Monte Carlo I: mc1.f

Monte Carlo II: mc2.f

(with importance sampling)

Monte Carlo II with importance sampling

- a) Pick x with probability $G_p(x)$ which is sharply peaked
- b) Make a list of points $X = \{x_1, x_2, \dots, x_M\}$ such that the probability of finding a particular x_i in this set $P(x_i) = G_p(x_i)$

$$I = \sum_{x \in X} G_s(x)$$

How to make such a list? METROPOLIS ALGORITHM

Computational Physics
Koonin, Addison-Wesley

Start with some initial x

Let $W(x, x')$ be a transition probability or a rule which takes $x \rightarrow x'$

$W(x, x')$ must satisfy:

$$W(x \rightarrow x') \geq 0$$

$$\sum_{x'} W(x \rightarrow x') = 1$$

$$P(x') = \sum_x P(x)W(x \rightarrow x')$$

Ergodicity: must be able to access all phase space

DETAILED BALANCE

$$P(x)W(x \rightarrow x') = P(x')W(x' \rightarrow x)$$

METROPOLIS ALGORITHM

Is one algorithm that imposes detailed balance

$$W(x \rightarrow x') = \min\left(1, \frac{P(x')}{P(x)}\right)$$

Operationally

1. Pick a starting point x_0
2. Evaluate $P(x_0)$
3. Choose x_{trial} at random
4. Evaluate $P(x_{trial})$
5. If $\frac{P(x_{trial})}{P(x_0)} \geq r$ (random number $[0,1]$)
put $x_1 = x_{trial}$ and put x_1 in the list X
6. If $\frac{P(x_{trial})}{P(x_0)} < r$ put $x_1 = x_0$ and put x_1 in the list X

$$I = \int_0^1 dx G_p(x; a_2, \sigma_2) G_s(x; a_1, \sigma_1)$$

$$a_1 = 0; \sigma_1 = 0.5$$

$$a_2 = 0.5; \sigma_2 = 0.05$$

Exact=0.48393

$$G(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-a)^2}{2\sigma^2}}$$

$$a_1 = 0; \sigma_1 = 0.5$$

$$a_2 = 0.5; \sigma_2 = 0.01$$

Exact=0.48393

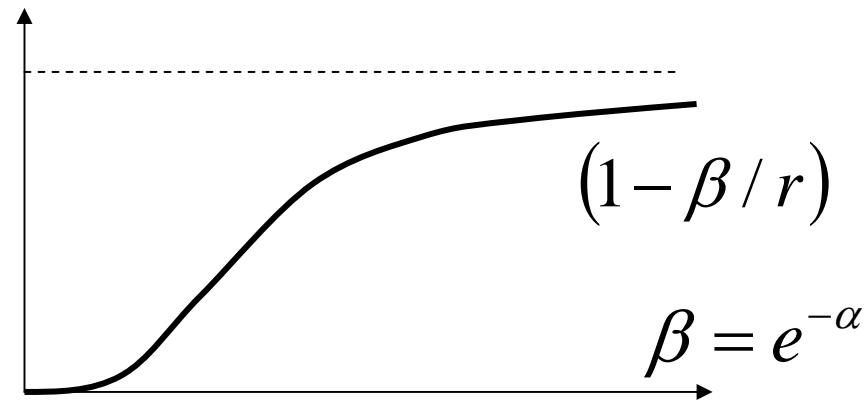
#points	I(MC1)	error	I(MC2)	err	I(MC1)	err	I(MC2)	err
100	0.388	0.28	0.463	0.034	0.383	0.64	0.4931	0.009
1000	0.488	0.11	0.4816	0.014	0.376	0.18	0.4867	0.004
10000	0.507	0.03	0.4841	0.003	0.4977	0.09	0.4841	0.002
100000	0.4837	0.012	0.48415	0.0015	0.4799	0.018	0.4839	0.0004

MC1: without importance sampling

MC2: with imp. sampl.

boson.f

$$\Psi_T = \prod_{i < j} f$$



Long range correlations in ground state

EXPECTATION VALUES AND MONTE CARLO

$$\Psi(R) \equiv \Psi(r_1, r_2, \dots, r_N) = \prod_{i < j} f(r_{ij})$$

$$E_T = \frac{\langle \Psi_T | H | \Psi_T \rangle}{\langle \Psi_T | \Psi_T \rangle} = \frac{\sum_{R,R'} \langle \Psi_T | R \rangle \langle R | H | R' \rangle \langle R' | \Psi_T \rangle}{\sum_R \langle \Psi_T | \Psi_T \rangle}$$

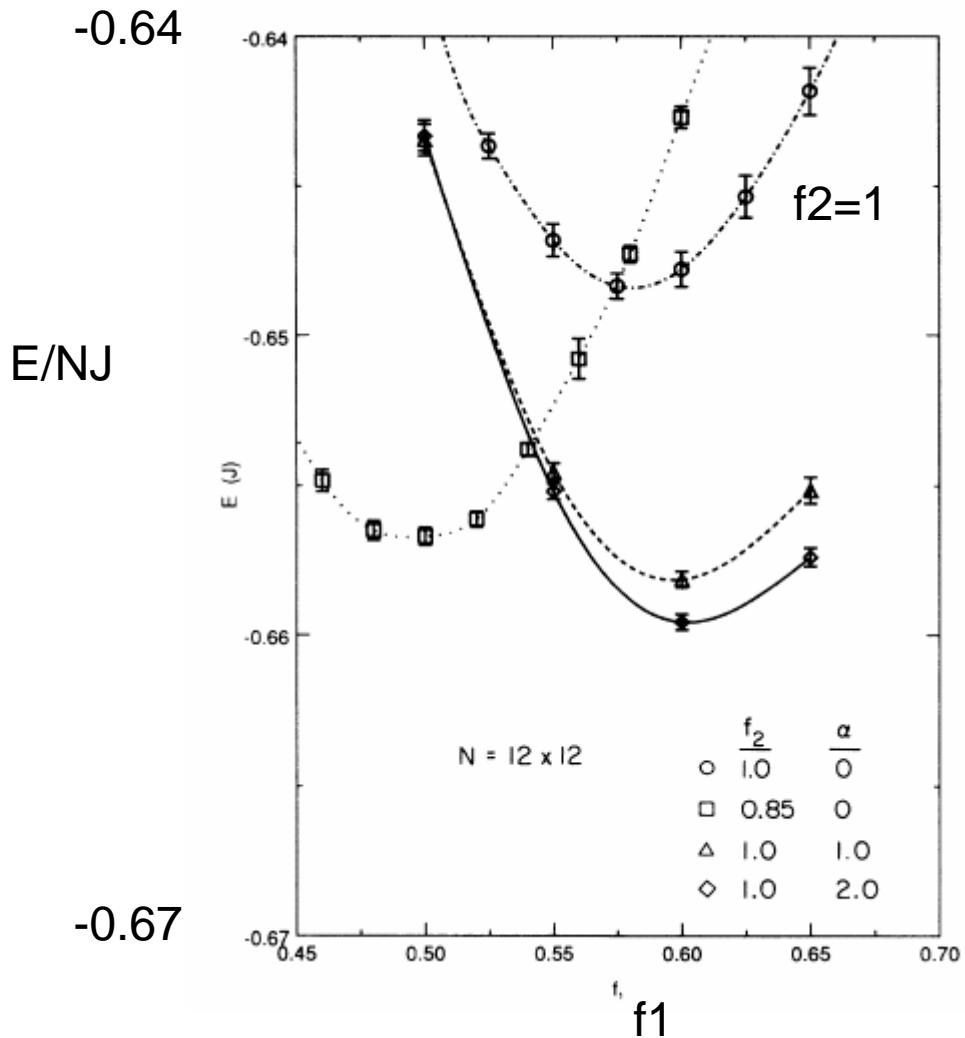
Many interacting particles
Multi dimensional integrals

$$E_T = \sum_R P(R) \underbrace{\sum_{R'} \langle R | H | R' \rangle}_{\varepsilon(R)} \frac{\Psi_T(R')}{\Psi_T(R)} = \sum_{R \in P(R)} \varepsilon(R)$$

Generate a list of configurations $\{R\}$ with $P(R) = \frac{|\Psi_T(R)|^2}{\sum_R |\Psi_T(R)|^2}$

Calculate the energy $\varepsilon(R)$ In each configuration R

Eneel/NJ=-0.5



NT and D.M. Ceperley
PRB 41, 4552 (1990)

GREEN FUNCTION / PROJECTOR QUANTUM MONTE CARLO

AIM: Filter out of a trial wave function the component of the true ground state

$$H|\Phi_\alpha\rangle = \varepsilon_\alpha |\Phi_\alpha\rangle \quad \text{Exact eigenstates of } H$$

$$\Psi_n = (e^{-\tau H})^n \Psi_T$$

$$\Psi_n(R) = \sum_{\alpha} \Phi_{\alpha}(R) e^{-n\tau\varepsilon_{\alpha}} \langle \Phi_{\alpha} | \Psi_T \rangle$$

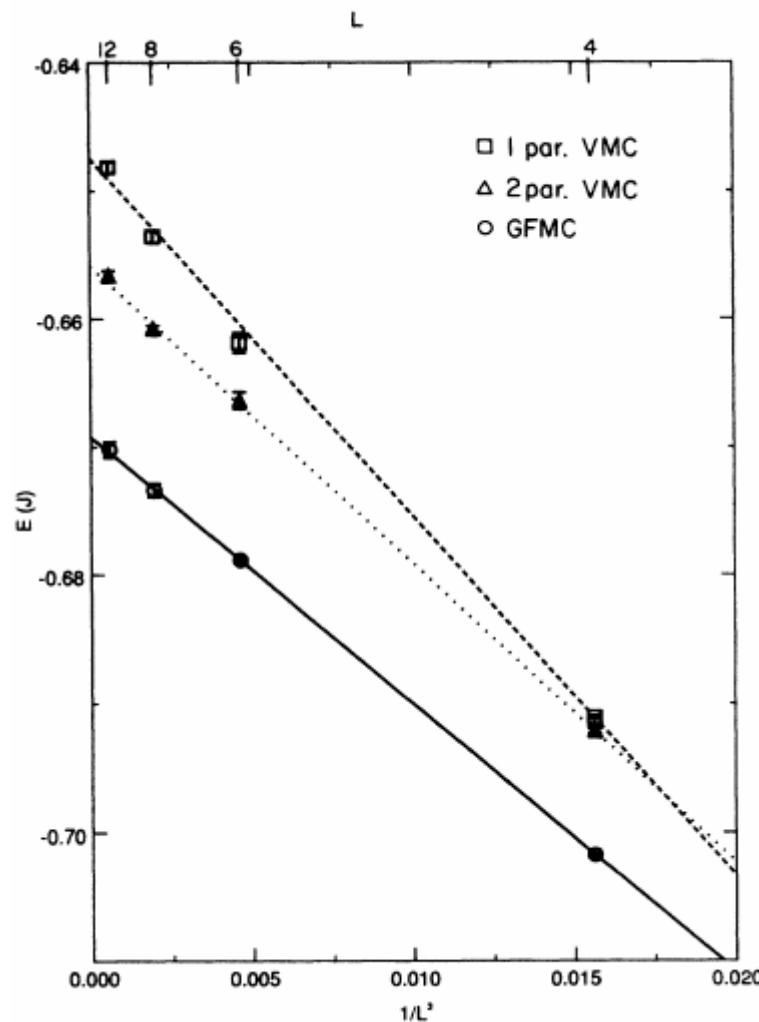
Ground state ε_0 decays most slowly

$$\lim_{n \rightarrow \infty} \Psi_n(R) \rightarrow \Phi_0(R) \langle \Phi_0 | \Psi_T \rangle$$



Overlap of true ground state with trial wave function

Finite size scaling



$$E / NJ = -0.6692 \pm 0.0002$$

Best result obtained by GFMC

SUPERFLUID

MAGNET

Off diagonal long range order

$$\langle b_i^+ \rangle \neq 0$$

Magnetization in XY plane

$$\langle S_i^+ \rangle \neq 0$$

In a subspace with fixed number of particles

$$h(l) = \langle b_i^+ b_{i+l}^- \rangle \xrightarrow{l \rightarrow \infty} const \neq 0$$

$$h(l \rightarrow \infty) = (m_x^+)^2 + (m_y^+)^2$$

Condensate fraction

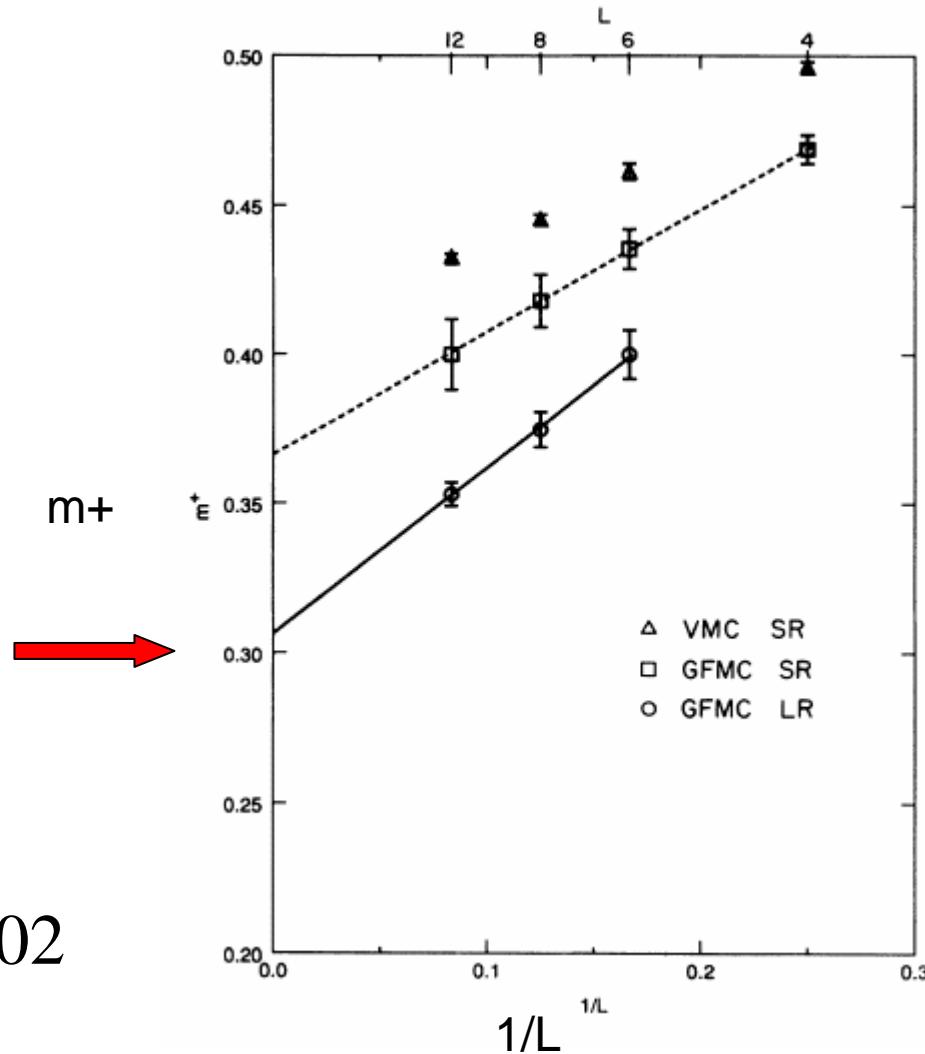
$$g(l) = \langle n_i n_{i+l} \rangle \xrightarrow{l \rightarrow \infty} (m_z^+)^2$$

Sublattice magnetization in XY plane

Diagonal long range order

Sublattice magnetization along z

Staggered magnetization survives inspite of quantum fluctuations



$$m^+ = 0.32 \pm 0.02$$

Quantum Monte Carlo is not one black box

Variational QMC

Green function or Diffusion QMC

Path Integral QMC

Determinantal QMC

Comparisons of the different QMC techniques

SIGN PROBLEM IN DIFFERENT INCARNATIONS

GFMC: Nodes in Wave Function

$$Z_{\text{PfMC}} = \sum_{\{S\}} \det[M_{\uparrow}(\{S\})] \det[M_{\downarrow}(\{S\})] \equiv \sum_{\{S\}} P(\{S\})$$

$$\langle \hat{O} \rangle = \sum_{\{S\}} P(\{S\}) O(\{S\})$$

$$\tilde{P} = |P|$$

$$\langle \hat{O} \rangle = \frac{\langle O \operatorname{sgn}(\tilde{P}) \rangle_{\tilde{P}}}{\langle \operatorname{sgn}(\tilde{P}) \rangle_{\tilde{P}}}$$

small average sign at low T leads to large fluctuations in MC averages

Recent progress: Projected techniques + Det QMC: Imada; Shiwei Zhang

Energy is sensitive only to short range correlations in wave function

For an improved description must include contribution from zero point motion of elementary excitations in the ground state

$$H_{ZP} = \frac{1}{2} \sum_{q < q_c} m_q (\dot{\rho}_q \dot{\rho}_{-q} + \omega_q^2 \rho_q \rho_{-q})$$

$$\Psi_{ZP} = \prod_{q < q_c} \exp\left(-\frac{1}{2} m_q \omega_q \rho_q \rho_{-q}\right) = \prod_{i < j} \exp\left(-\frac{\alpha}{|\vec{r}_i - \vec{r}_j|}\right)$$

$$\rho_q = \frac{1}{N} \sum_i e^{i \vec{q} \cdot \vec{r}_i} \quad \omega_q \sim cq$$

$$m_q \sim \frac{1}{q^2} \quad d = 2$$

$$\hbar c = \frac{\pi \alpha |T_k|}{2} Ja$$

My teaching philosophy is summarized in this saying
by Confucius:

I hear and I forget

I see and I remember

I do and I understand