Determinant quantum Monte Carlo algorithm for simulating Hubbard models in collaboration with Edwin Huang, Elizabeth Nowadnick, Yvonne Kung, Steven Johnston, Brian Moritz and Thomas Devereaux

Christian B. Mendl

July 5, 2016





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Mathematical and Numerical Analysis of Electronic Structure Models Roscoff, France July 2016

### Quantum ensemble averages

Goal: compute canonical ensemble averages

$$\langle \hat{A} \rangle = Z^{-1} \operatorname{tr} \big[ \hat{A} e^{-\beta H} \big], \quad Z = \operatorname{tr} \big[ e^{-\beta H} \big]$$

for a Hubbard-type Hamiltonian

$$H = K + V$$

$$K = -t \sum_{\langle i,j \rangle,\sigma} \left( c_{i\sigma}^{\dagger} c_{j\sigma} + c_{j\sigma}^{\dagger} c_{i\sigma} \right)$$

$$-\mu \sum_{i} (n_{i\uparrow} + n_{i\downarrow})$$

$$V = \frac{U}{V} \sum_{i} (n_{i\uparrow} - \frac{1}{2})(n_{i\downarrow} - \frac{1}{2})$$

- t kinetic hopping amplitude
- $\mu$   $\,$  chemical potential  $\,$
- U el-el interaction strength



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# Traces in quantum Fock-space

For a Hamiltonian of quadratic form

$$H=\sum_{i,j}\mathbf{h}_{ij}\,c_i^{\dagger}c_j$$

the following exact identity holds:

$$tr[e^{-\beta H}] = det[1 + e^{-\beta h}]$$
full quantum Fock-space / single-particle space

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Simple to check for a single "orbital":  $H = \epsilon c^{\dagger} c$ ,

$$\mathrm{tr}\big[\mathrm{e}^{-\beta H}\big] = \langle 0|\mathrm{e}^{-\beta\epsilon c^{\dagger}c}|0\rangle + \langle 1|\mathrm{e}^{-\beta\epsilon c^{\dagger}c}|1\rangle = 1 + \mathrm{e}^{-\beta\epsilon}.$$

In the general case, consider a basis in which h is diagonal.

Blankenbecler et al. PRD 24, 2278 (1981)

# Hubbard-Stratonovich transformation

However, Hubbard interaction term contains four-fermion operators  $\ldots$  May the 4th be with you!

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#### Hubbard-Stratonovich transformation

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$$e^{-\Delta \tau U \sum_{i} (n_{i\uparrow} - \frac{1}{2})(n_{i\downarrow} - \frac{1}{2})} = e^{-\Delta \tau U/4} \frac{1}{2} \sum_{s=\pm 1} e^{-\Delta \tau s \lambda (n_{i\uparrow} - n_{i\downarrow})}$$
four-fermion ops.   
( $n_{i\sigma} = c^{\dagger}_{i\sigma} c_{i\sigma}$ )

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 $\lambda$  determined by  $\cosh(\Delta \tau \lambda) = e^{\Delta \tau U/2}$ 

White, Scalapino et al. PRB 40, 506 (1989)

### From quantum ensemble averages to classical Monte Carlo

Applying Trotter splitting ( $\beta = \Delta \tau L$ ) and the Hubbard-Stratonovich transformation eventually leads to

$$Z = \operatorname{tr}\left[e^{-\beta H}\right] = \sum_{\{s_{i\ell} = \pm 1\}} \det\left[M^{\uparrow}(s)\right] \det\left[M^{\downarrow}(s)\right]$$

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with

$$M^{\sigma}(s) = \mathbb{1} + B^{\sigma}_{L-1}(s) B^{\sigma}_{L-2}(s) \cdots B^{\sigma}_{0}(s),$$

where

$$B^{\sigma}_{\ell}(s) = \mathrm{e}^{\mp \Delta au \lambda \mathbf{v}(s_{\cdot \ell})} \, \mathrm{e}^{-\Delta au k}, \quad \ell = 0, 1, \dots, L-1,$$

k is the kinetic single-particle matrix and

$$\mathbf{v}(s_{\ell}) = \begin{pmatrix} s_{1\ell} & 0 & 0 & \dots \\ 0 & s_{2\ell} & 0 & \dots \\ 0 & 0 & s_{3\ell} & \dots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

White, Scalapino et al. PRB 40, 506 (1989)

# From quantum ensemble averages to classical Monte Carlo

Idea: interpret as probability density

 $Z^{-1} \det \left[ M^{\uparrow}(s) \right] \det \left[ M^{\downarrow}(s) \right]$ 



Figure:  $\uparrow$ :  $s_{i\ell} = 1$ ,  $\downarrow$ :  $s_{i\ell} = -1$ 

 $\rightarrow$  sample over Hubbard-Stratonovich field configurations  $s_{i\ell}$  using classical Monte-Carlo

# Imaginary time Green's function

Matsubara Green's function (with  $\tau, \tau' \in [0, \beta]$ ):

$$G^{\sigma}(\tau,\tau')_{ij} = \langle \mathcal{T}c_{i\sigma}(\tau)c_{j\sigma}^{\dagger}(\tau') \rangle.$$

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For discretized version in the field s and  $\tau = \tau'$ , can derive that

$$G^{\sigma}(\ell,s) = \left[\mathbb{1} + B^{\sigma}_{\ell-1}(s) \cdots B^{\sigma}_{0}(s) B^{\sigma}_{L-1}(s) \cdots B^{\sigma}_{\ell}(s)\right]^{-1}$$

with  $\tau = \Delta \tau \ell$  and

$$B_{\ell}^{\sigma}(s) = \mathrm{e}^{\mp \Delta \tau \lambda v(s_{\ell})} \mathrm{e}^{-\Delta \tau k}.$$

Blankenbecler et al. PRD 24, 2278 (1981)

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- Sequentially for all *i* and  $\ell$ , suggest a flip  $s_{i\ell} \rightarrow s'_{i\ell} = -s_{i\ell}$ , acceptance probability  $R = R^{\uparrow}R^{\downarrow}$  with

$$R^{\sigma} = \frac{\det[M^{\sigma}(s')]}{\det[M^{\sigma}(s)]} = 1 + (1 - G^{\sigma}(\ell, s)_{ii}) \left(e^{\pm 2\Delta\tau\lambda s_{i\ell}} - 1\right)$$

(computationally "cheap" since Green's function is known)

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• To avoid gradual loss of precision, still have to recompute Green's function from scratch after several steps (expensive)

White, Scalapino et al. PRB 40, 506 (1989)

Algorithm relies on probability density

$$Z^{-1} \det \left[ M^{\uparrow}(s) 
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To circumvent, factor into  $sign(s) \times absolute$  value, and pull sign towards observable:

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However, for low temperatures,  $\langle \operatorname{sign}(s) \rangle$  becomes small, need many more Monte-Carlo samples to reliably estimate  $\langle \hat{A} \rangle$ .

Many additional "tricks of the trade" to stabilize and speed up algorithm, for example

• Explicitly keep track of *determinants* of Green's functions, otherwise numerical loss of precision when computing them based on matrix entries

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- Long chain of matrix multiplications in

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can lead to extremely large condition numbers; alleviate by sequential QR-decompositions with column pivoting

Bai et al. Linear Algebra Appl. 435, 659-673 (2011)

# Example: occupancy and sign



Average occupancy in dependence of chemical potential  $\mu$ 



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Parameters: 8 imes 8 lattice, t = 1, t' = -0.3, U = 8,  $\Delta au$  = 0.1

# Example: spectral function



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Figure: Doping dependence of the spectral function  $A(\mathbf{k}, \omega)$  along high symmetry cuts through the Brillouin zone

Parameters: 8 imes 8 lattice, t = 1, t' = -0.25, U = 8,  $\beta$  = 3,  $\Delta au$  = 0.1

#### Example: spectral function



Figure: Doping dependence of the spectral function  $A(\mathbf{k}, \omega)$  along high symmetry cuts through the Brillouin zone

Parameters: 8  $\times$  8 lattice, t = 1, t' = -0.25, U = 8,  $\beta = 3$ ,  $\Delta \tau = 0.1$ 

Using maximum entropy analytic continuation to "real" frequencies  $\omega$ ,

$$G(\mathbf{k}, au) = \int_{-\infty}^{\infty} rac{\mathrm{e}^{- au\omega}}{1 + \mathrm{e}^{-eta\omega}} A(\mathbf{k}, \omega) \,\mathrm{d}\omega.$$

Jarrell and Gubernatis, Phys. Rep. 269, 133-195 (1996)

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# Summary

Main ingredients:

• From quantum Fock to single-particle space

$$\mathrm{tr}\big[\mathrm{e}^{-\beta H}\big] = \mathsf{det}\big[\mathbb{1} + \mathrm{e}^{-\beta h}\big]$$

- Introduce auxiliary Hubbard-Stratonovich field  $s_{i\ell}$  to obtain quadratic form for U-term
- Efficient Metropolis sampling of

$$Z^{-1} \det[M^{\uparrow}(s)] \det[M^{\downarrow}(s)]$$

using Green's function



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Introduce an imaginary-time step  $\Delta \tau$  such that  $\beta = \Delta \tau L$ . Trotter splitting  $\rightsquigarrow$ 

$$Z = \operatorname{tr} \left[ e^{-\Delta \tau L H} \right] \simeq \left( \operatorname{tr} \left[ e^{-\Delta \tau V} e^{-\Delta \tau K} \right] \right)^{L}$$

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Remark: the product

$$B_{L-1}(s) B_{L-2}(s) \cdots B_0(s)$$

with

$$B_{\ell}(s) = e^{-\Delta \tau \lambda v(s_{\ell})} e^{-\Delta \tau k}$$

is a discrete approximation of the imaginary time flow

$$U( au, au') = \mathcal{T} \exp \Big[ - \int_{ au'}^{ au} \mathrm{d}\hat{ au} \, H(\hat{ au}) \Big]$$

in the field s, where  $\mathcal{T}$  is the imaginary-time ordering operator.

Blankenbecler et al. PRD 24, 2278 (1981)