

# The Theory of Configuration Path Integral Monte Carlo

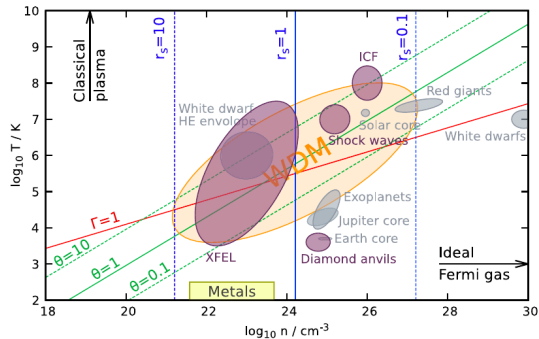
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Kai Hunger

January 29, 2019

# Motivation

- Warm Dense Matter



Applications for calculations of the WDM state<sup>1</sup>

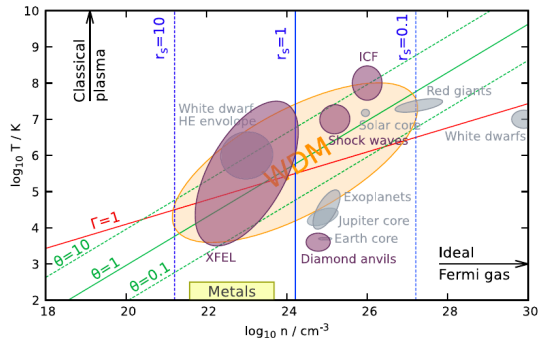
in a.u.:  $r_s^3 = 3/(4\pi n)$ ,  $\Theta = (k_B T/E_F)$ ,  $\Gamma = (r_s k_B T)^{-1}$

<sup>1</sup>T. Dornheim, S. Groth, and M. Bonitz. Phys. Rep., 744:1–86, May 2018.

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- no exact large-scale ab initio approach
- no "small" parameter allowing approximation
  - to interaction strength ( $r_s \rightarrow 0, r_s \rightarrow \infty$ )
  - by ground state ( $\Theta \rightarrow 0$ )
  - by classical methods ( $\Gamma \rightarrow 0$ )



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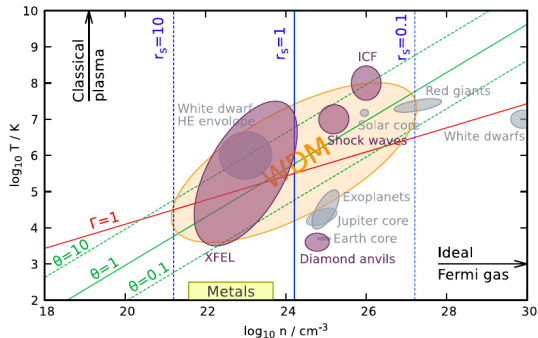
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  - CPIMC: Complementary behavior of FSP with  $r_s \rightarrow$  no FSP at high densities

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Motivation

Introduction to PIMC

CPIMC Formalism

CPIMC Algorithm

CPIMC for the HEG

Outlook

Main Literature:

- S. Groth. *Strongly degenerate nonideal fermi systems: Configuration path integral monte carlo simulation*. Master's thesis, ITAP, CAU, 2014.
- T. Schoof. *Configuration Path Integral Monte Carlo: Ab initio simulations of fermions in the warm dense matter regime*. PhD thesis, ITAP, CAU, 2016.

Many-Particle System characterized by Canonical Ensemble<sup>1</sup> ( $N, \beta, V$ ):

$$\hat{\rho} = \frac{1}{Z} e^{-\beta \hat{H}}, \quad Z = \text{Tr} e^{-\beta \hat{H}}$$

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Expectation Values of operator  $\hat{O}$ :

- from density operator:

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- from thermodynamic relations:

$$\langle \hat{H} \rangle = -\frac{\partial}{\partial \beta} \log(Z), \quad F = -T \log(Z),$$

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Time evolution operator:

$$\hat{U}(t := t_2 - t_1) = e^{-i\hat{H}t}$$

$$\Rightarrow \hat{\rho} = \hat{U}(-i\beta)$$

$$\Rightarrow Z = \text{Tr} \hat{U}(-i\beta)$$

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understand inverse temperature  
as imaginary time

$$\beta \leftrightarrow \tau := i t$$

---

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## Introduction to PIMC — Path Integral Formulation

high temperature decomposition:  $\tau := (\beta/M)$

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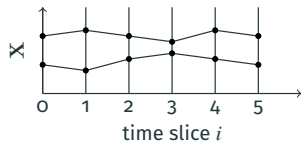
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*Paths*  $C = (X, X_1, \dots, X_{M-1}, \tau_0, \tau_1, \dots, \tau_{M-1})$

Expectation Values:  $\langle \hat{O} \rangle = \int_C O(C) W(C)$



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WHY USE  
HIGH-TEMPERATURE  
DECOMPOSITION  
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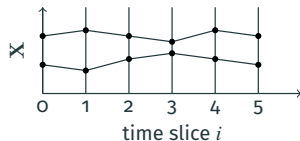
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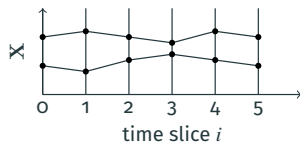
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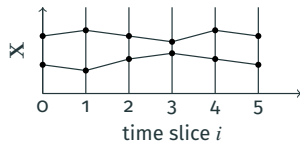
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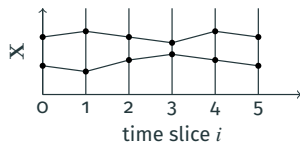
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→ Sample Paths  $C$  using Metropolis-Hastings Algorithm



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<sup>2</sup>Landau, Binder. A Guide to Monte Carlo Simulations in Statistical Physics. Cambr. Univ. Press, 2000.

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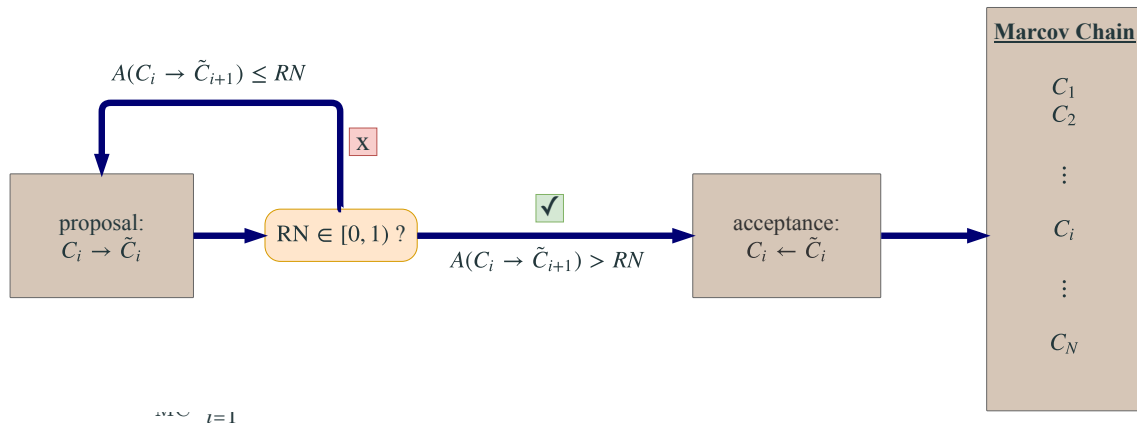
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- Metropolis-Hastings Algorithm:

1. Start with initial Configuration  $C_0$
2. MC-Steps:
  - for**  $1 \leq i \leq N_{\text{MC}}$  **do**
  - Propose  $C'_i$  with probability  $Q(C_i \rightarrow C'_i)$
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  - end for**
3. obtain *Markov-chain*  $(C_0, C_1, \dots, C_{N_{\text{MC}}})$



## Metropolis-Hastings Algorithm



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- Conditions:

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- Ergodic MC Steps  $C_i \rightarrow C'_i$

# Introduction to PIMC — Sign Problem

Fermions  $\rightarrow$  Antisymmetrize Productstates  $|X\rangle = |x_1\rangle |x_2\rangle \dots |x_N\rangle$

$$Z = \int dX \int dX_1 \dots \int dX_{M-1} \langle X | \hat{U}(-i\tau) | X_1 \rangle \dots \langle X_{M-1} | \hat{U}(-i\tau) | X \rangle$$

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$$\langle \hat{O} \rangle = \frac{\langle \hat{O} \hat{s} \rangle}{\langle \hat{s} \rangle}$$



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2nd Quantization Fock-States already include correct particle statistics !

Fock states  $|\{n^0\}\rangle$  defined by ONV<sup>1</sup>  $\{n^0\} = \{n_i = 0, 1 \mid i \in \mathbb{N}\} = \{n_1, n_2, n_3, \dots\}$

---

<sup>1</sup>occupation number vector

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## CPIMC Formalism — 2nd Quantization

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Calculating Matrix elements in ONV representation:

- one-particle operators

$$\hat{B}^{(1)} = \sum_{ij} b_{ij} \hat{a}_i^\dagger \hat{a}_j, \quad b_{ij} = \langle i | \hat{b} | j \rangle = \int dx \phi_i^*(x) b(x) \phi_j(x)$$

---

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Calculating Matrix elements in ONV representation:

- one-particle operators  $\rightarrow$  Slater-Condon rules

$$B_{12}^{(1)} = \begin{cases} \sum_k b_{kk} n_k & : \{n^{(1)}\} = \{n^{(2)}\} \\ (-1)^{\alpha_{1,p,q}} b_{pq} & : \{n^{(1)}\} = \{n^{(2)}\}_q^p \\ 0 & : \text{else} \end{cases}$$

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- pair-interaction operators:  $w(x, y) = w(y, x) = w^*(y, x)$

$$\hat{W} = \sum_{ijkl} \tilde{w}_{ijkl} \hat{a}_i^\dagger \hat{a}_j^\dagger \hat{a}_l \hat{a}_k, \quad \tilde{w}_{ijkl} = w_{ijkl} - w_{ijlk}$$

$$w_{ijkl} = \langle ij | \hat{w} | kl \rangle = \int dx \int dy \phi_i^*(x) \phi_j^*(y) w(x, y) \phi_k(x) \phi_l(y)$$

---

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- pair-interaction operators  $\rightarrow$  Slater-Condon rules

$$W_{12}^{(2)} = \begin{cases} \sum_{i=0} \sum_{j=i+1} \tilde{w}_{ijij} n_i^1 n_j^1 & : \{n^{(1)}\} = \{n^{(2)}\} \\ \sum_{p,q \neq i=0} (-1)^{\alpha_{1,p,q}} \tilde{w}_{ipiq} n_i^1 & : \{n^{(1)}\} = \{n^{(2)}\}_q^p \\ (-1)^{\alpha_{1,p,q} + \alpha_{2,r,s}} \tilde{w}_{pqrs} & : \{n^{(1)}\} = \{n^{(2)}\}_{r<s}^{p<q} \\ 0 & : \text{else} \end{cases}$$

<sup>1</sup>occupation number vector



- arbitrary Hamiltonian  $\hat{H} = \hat{B}^{(1)} + \hat{W}^{(2)} \Rightarrow$  only 1- and 2-particle excitations !

## Calculating Matrix elements in ONV representation:

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- arbitrary Hamiltonian  $\hat{H} = \hat{B}^{(1)} + \hat{W}^{(2)} \Rightarrow$  only 1- and 2-particle excitations !
- off-diagonal elements specified by T2 ( $p, q$ ) and T4 ( $p, q, r, s$ )

Calculating Matrix elements in ONV representation:

- one-particle operators → Slater-Condon rules

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Separate Hamiltonian:  $\hat{H} =: \hat{D} + \hat{Y}$

$$H_{ij} := \langle \{n^{(i)}\} | \hat{H} | \{n^{(j)}\} \rangle = \begin{cases} D_i := \langle \{n^{(i)}\} | \hat{D} | \{n^{(i)}\} \rangle & : i = j \text{ diagonal} \\ Y_{ij} := \langle \{n^{(i)}\} | \hat{Y} | \{n^{(j)}\} \rangle & : i \neq j \text{ off-diagonal} \end{cases} = \begin{pmatrix} D_1 & Y_{1,2} & Y_{1,3} & \dots \\ Y_{2,1} & D_2 & Y_{2,3} & \dots \\ Y_{3,1} & Y_{3,2} & D_3 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

## CPIMC Formalism — Partition Function

Separate Hamiltonian:  $\hat{H} =: \hat{D} + \hat{Y}$

$$(D_i := \langle \{n^{(i)}\} | \hat{D} | \{n^{(i)}\} \rangle : i = j \text{ diagonal} \quad \begin{pmatrix} D_1 & Y_{1,2} & Y_{1,3} & \dots \\ Y_{2,1} & D_2 & Y_{2,3} & \dots \\ Y_{3,1} & Y_{3,2} & D_3 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix} \quad 9/17$$

Separate Hamiltonian:  $\hat{H} =: \hat{D} + \hat{Y} \rightarrow$  Interaction Picture:  $\hat{H}(t) = \hat{D} + \hat{Y}(t) = \hat{D} + e^{i\hat{D}t}\hat{Y}e^{-i\hat{D}t}$

- time evolution:

$$\hat{U}(t) = e^{-i\hat{D}t} \underbrace{\mathcal{T} \exp \left\{ - \int_0^t dt' \hat{Y}(t') \right\}}_{\text{time-ordered exponential}} = e^{-i\hat{D}t} \underbrace{\sum_{K=0}^{\infty} (-i)^K \int_0^t dt_1 \int_{t_1}^t dt_2 \dots \int_{t_{K-1}}^t dt_K \hat{Y}(t_K) \dots \hat{Y}(t_2) \hat{Y}(t_1)}_{\text{Dyson series}}$$

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- perform trace:

$$\begin{aligned} Z &= \text{Tr } \hat{U}(-i\beta) = \sum_{\{n^{(0)}\}} \langle \{n^{(0)}\} | \hat{U}(-i\beta) | \{n^{(0)}\} \rangle \\ &= \sum_{\{n^{(0)}\}} \sum_{K=0}^{\infty} \int_0^\beta d\tau_1 \int_{\tau_1}^\beta d\tau_2 \dots \int_{\tau_{K-1}}^\beta d\tau_K \langle \{n^{(0)}\} | (-1)^K e^{-i\hat{D}t} \hat{Y}(\tau_K) \dots \hat{Y}(\tau_2) \hat{Y}(\tau_1) | \{n^{(0)}\} \rangle \end{aligned}$$

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$$\begin{aligned} Z &= \text{Tr} \hat{U}(-i\beta) = \sum_{\{n^{(0)}\}} \langle \{n^{(0)}\} | \hat{U}(-i\beta) | \{n^{(0)}\} \rangle \\ &= \sum_{K=0}^{\infty} \sum_{\{n^{(0)}\}} \sum_{\{n^{(1)}\}} \dots \sum_{\{n^{(K-1)}\}} \int_0^\beta d\tau_1 \int_{\tau_1}^\beta d\tau_2 \dots \int_{\tau_{K-1}}^\beta d\tau_K \times \\ &\quad \times (-1)^K \langle \{n^{(0)}\} | e^{-i\hat{D}\tau_K} \hat{Y}(\tau_K) | \{n^{(K-1)}\} \rangle \dots \langle \{n^{(2)}\} | \hat{Y}(\tau_2) | \{n^{(1)}\} \rangle \langle \{n^{(1)}\} | \hat{Y}(\tau_1) | \{n^{(0)}\} \rangle \end{aligned}$$

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where is the path... ?



# CPIMC Formalism — Path Integral Formulation

Define *Kinks*:

- type 2:  $s_{T_2} := (p, q) \in \mathbb{N}^2$
- type 4:  $s_{T_4} := (p, q, r, s) \in \mathbb{N}^4$
- General type:  $s \in \mathbb{N}^2 \cup \mathbb{N}^4$

$$Z(N, V, \beta) = \overbrace{\sum_{\substack{K=0 \\ K \neq 1}}^{\infty} \sum_{\{n^{(k)}\}} \sum_{s_1} \dots \sum_{s_{K-1}} \int_0^\beta d\tau_1 \int_{\tau_1}^\beta d\tau_2 \dots \int_{\tau_{K-1}}^\beta d\tau_K}^{:= \oint} (-1)^K \exp \left\{ - \sum_{i=0}^K \textcolor{red}{D}_i(\tau_{i+1} - \tau_i) \right\} \prod_{i=1}^K Y_{i,i-1}(s_i) \underbrace{\quad}_{:= W(C)}$$

Ingredients for  $\beta$ -periodic path:

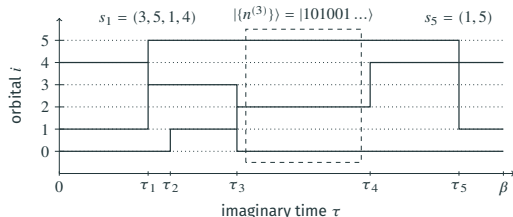
- T2-kinks:



- T4-kinks:



- *stationary links*:



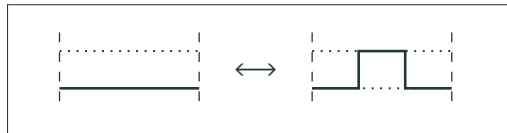
Example configuration Path  $C := (\{n^{(\cdot)}\} s_1, \dots, s_K, \tau_1, \dots, \tau_K)$

- Add/Remove 2 kinks

Detailed Balance:

$$\begin{aligned} & p_{\text{ap}} \frac{1}{2\beta N_0} p(\tau) |W(C)| \nu(C \rightarrow C') \\ &= p_{\text{rp}} \frac{1}{K+2} |W(C')| \nu(C' \rightarrow C) \end{aligned}$$

- Add/Remove 2 kinks
- T2 kink

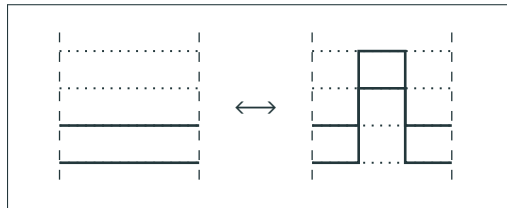


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# CPIMC Algorithm

- Add/Remove 2 kinks
  - T2 kink • T4 kink



Detailed Balance:

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# CPIMC Algorithm

- Add/Remove 2  
kinks
  - T2 kink • T4 kink
- Add/Remove 1  
kink1

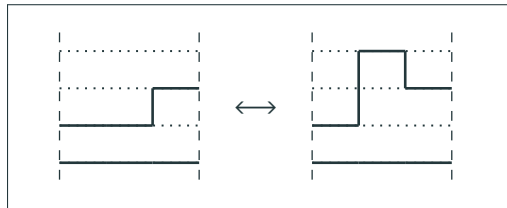


Detailed Balance:

$$\begin{aligned}
 & p_{\text{ak}} \frac{1}{4KN_o} p(\tau) |W(C)| v(C \rightarrow C') \\
 & = p_{\text{rk}} \frac{1}{K+1} \frac{1}{2N_K} |W(C')| v(C' \rightarrow C)
 \end{aligned}$$

# CPIMC Algorithm

- Add/Remove 2 kinks
  - T2 kink • T4 kink
- Add/Remove 1 kink1
  - T2 kink

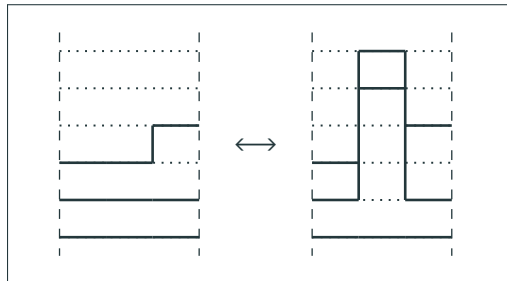


Detailed Balance:

$$\begin{aligned}
 & p_{\text{ak}} \frac{1}{4KN_o} p(\tau) |W(C)| v(C \rightarrow C') \\
 & = p_{\text{rk}} \frac{1}{K+1} \frac{1}{2N_K} |W(C')| v(C' \rightarrow C)
 \end{aligned}$$

# CPIMC Algorithm

- Add/Remove 2 kinks
  - T2 kink • T4 kink
- Add/Remove 1 kink1
  - T2 kink • T4 kink



Detailed Balance:

$$\begin{aligned}
 & p_{\text{ak}} \frac{1}{4KN_o} p(\tau) |W(C)| v(C \rightarrow C') \\
 & = p_{\text{rk}} \frac{1}{K+1} \frac{1}{2N_K} |W(C')| v(C' \rightarrow C)
 \end{aligned}$$



- Add/Remove 2 kinks
  - T2 kink • T4 kink
- Add/Remove 1 kink1
  - T2 kink • T4 kink
- Change 2 kinks

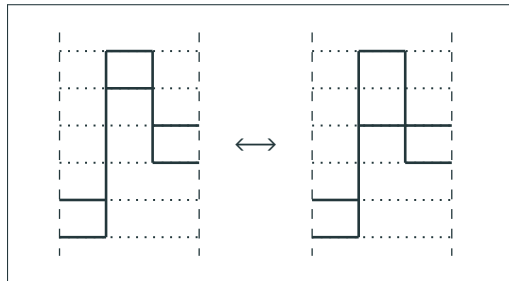


Detailed Balance:

$$\begin{aligned}
 & p_{\text{ck}} \frac{1}{2KN_oN_K} |W(C)| \nu(C \rightarrow C') \\
 &= p_{\text{ck}} \frac{1}{2KN_oN_{K'}} |W(C')| \nu(C' \rightarrow C)
 \end{aligned}
 \quad 11/17$$

# CPIMC Algorithm

- Add/Remove 2 kinks
  - T2 kink • T4 kink
- Add/Remove 1 kink1
  - T2 kink • T4 kink
- Change 2 kinks
  - T2 kink

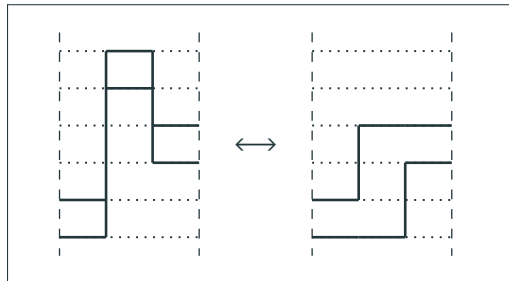


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- Add/Remove 2 kinks
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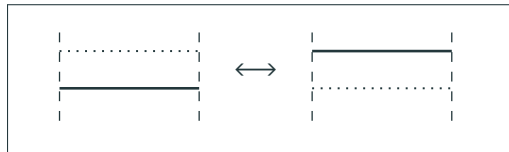


Detailed Balance:

$$\begin{aligned}
 & p_{\text{ck}} \frac{1}{2KN_0N_K} |W(C)| v(C \rightarrow C') \\
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 \end{aligned}$$

# CPIMC Algorithm

- Add/Remove 2 kinks
  - T2 kink • T4 kink
- Add/Remove 1 kink
  - T2 kink • T4 kink
- Change 2 kinks
  - T2 kink • T4 kink
- Excite stationary orbital



Detailed Balance:

$$p_{\text{eo}} \frac{1}{\tilde{N}_p \tilde{N}_q} |W(C)| v(C \rightarrow C')$$

$$= p_{\text{eo}} \frac{1}{\tilde{N}_p \tilde{N}_q} |W(C')| v(C' \rightarrow C)$$

# CPIMC Algorithm

- Add/Remove 2 kinks
  - T2 kink • T4 kink



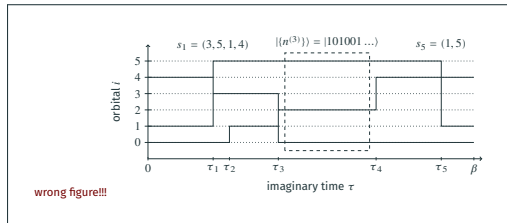
- Add/Remove 1 kink
  - T2 kink • T4 kink



- Change 2 kinks
  - T2 kink • T4 kink



- Excite stationary orbital



# CPIMC for the HEG — Hamiltonian

- $N$  electrons located in a finite volume  $V = L^3$  of cubic shape
- ions  $\leftrightarrow$  *uniform* background (*jellium*)  $\Rightarrow$  overall charge-neutrality
- single-particle CONS: plane waves  $\langle \mathbf{r} | \mathbf{k} \sigma \rangle = \frac{1}{\sqrt{V}} e^{i\mathbf{k} \cdot \mathbf{r}}$
- periodic boundary conditions  $\Rightarrow \mathbf{k} = \frac{2\pi}{L} (n_1, n_2, n_3)$  with  $n_1, n_2, n_3 \in \mathbb{Z}$
- Hamiltonian:

• in thermodynamic limit:

$$\hat{H} = \sum_{i=1}^N \frac{\hat{\mathbf{p}}_i^2}{2} + \frac{1}{2L^3} \sum_{i,j=1}^N \sum_{\mathbf{q} \neq 0} v_{\mathbf{q}} \left[ e^{i\mathbf{q}(\hat{\mathbf{r}}_i - \hat{\mathbf{r}}_j)} - \hat{N} \right]$$

• for finite systems:

$$\hat{H} = - \sum_{i=1}^N \frac{\hat{\mathbf{p}}_i^2}{2} + \frac{1}{2} \sum_{i \neq j=1}^N \underbrace{U(\hat{\mathbf{r}}_i, \hat{\mathbf{r}}_j)}_{\text{Ewald pair potential}} + \underbrace{E_{\text{M}}}_{\text{PBC self-interaction}}$$

• in ONV representation:

$$\hat{H} = \sum_{i,j} \varepsilon_{ij} \hat{a}_i^\dagger \hat{a}_j + \sum_{\substack{i < j, k < l \\ i \neq k, j \neq l}} \tilde{w}_{ijkl} \hat{a}_i^\dagger \hat{a}_j^\dagger \hat{a}_l \hat{a}_k, \quad \tilde{w} = w_{ijkl} - w_{ijlk}$$

$$\varepsilon_{ij} = \frac{k_i^2}{2} \delta_{ij}, \quad w_{ijkl} = \frac{4\pi}{v(\mathbf{k}_i - \mathbf{k}_k)} \delta_{(\mathbf{k}_i + \mathbf{k}_j)(\mathbf{k}_k + \mathbf{k}_l)} \delta_{\sigma_i \sigma_k} \delta_{\sigma_j \sigma_l}$$

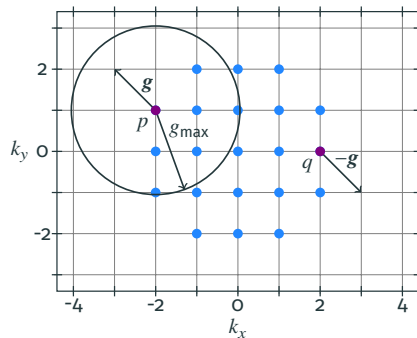
Momentum Conservation  $\Rightarrow$  no single-particle excitations (T2)

---

$$^1|g| < g_{\max}, \text{ e.g. } g_{\max} = 2$$

Momentum Conservation  $\Rightarrow$  no single-particle excitations (T2)

- MC-Steps 1,2: Add/Remove kink:  
choose 2 occupied  $n_i(\tau)$



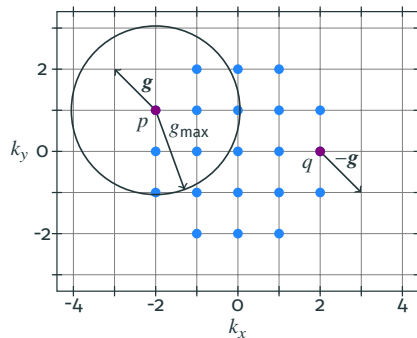
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<sup>1</sup> $|g| < g_{\max}$ , e.g.  $g_{\max} = 2$



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- MC-Steps 1,2: Add/Remove kink:
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  - $\rightarrow$  randomly choose *excitation vector*<sup>1</sup>  $\mathbf{g} \in \mathbb{Z}^3$

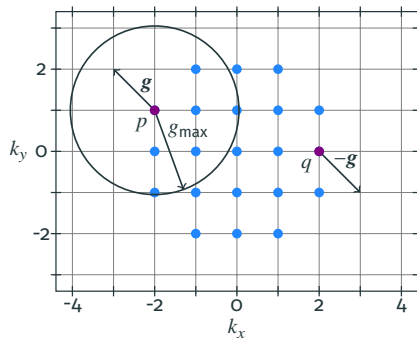


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  - $\rightarrow$  propose  $\mathbf{k}_{r/s} = \mathbf{k}_{p/q} \pm \mathbf{g}$

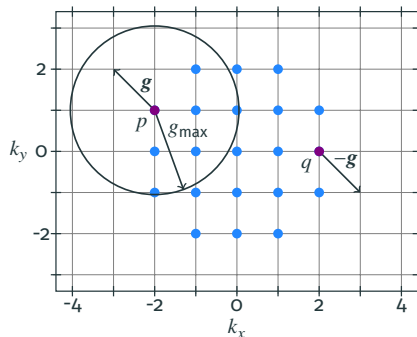


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- MC-Steps 1,2: Add/Remove kink:
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  - $\rightarrow$  propose  $\mathbf{k}_{r/s} = \mathbf{k}_{p/q} \pm \mathbf{g}$
- MC-Step 4: excite stationary orbital
  - randomly choose stationary orbital within shell  $i$  of energy  $\varepsilon_k \in [i\Delta_\varepsilon, i + 1\Delta_\varepsilon)$

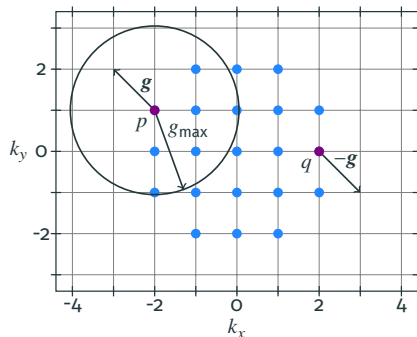


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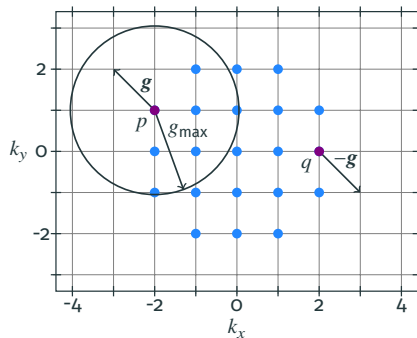


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Momentum Conservation  $\Rightarrow$  no single-particle excitations (T2)

- MC-Steps 1,2: Add/Remove kink:
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  - randomly choose stationary orbital within shell  $i$  of energy  $\varepsilon_k \in [i\Delta_\varepsilon, i + 1\Delta_\varepsilon)$
  - $\rightarrow$  propose target orbital from adjacent shells  $[i - 1, i, i + 1]$



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# CPIMC for the HEG — Sign Problem

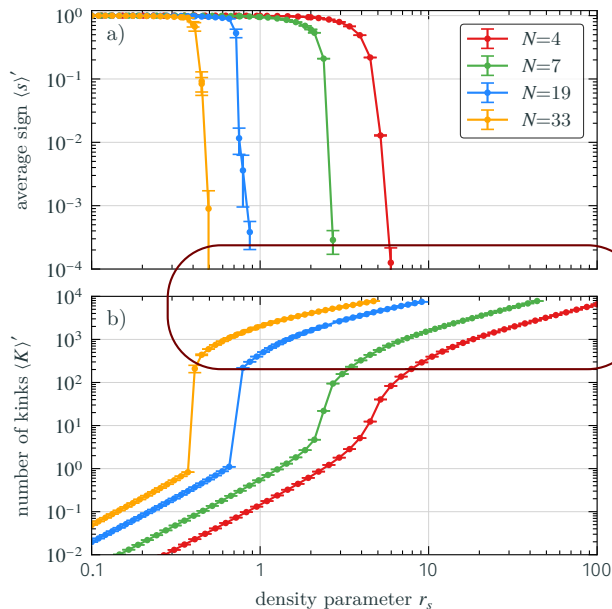
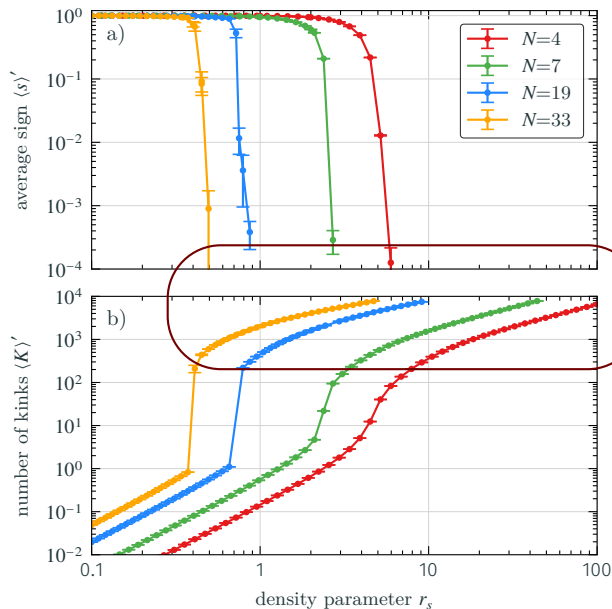


figure:  $\Theta = 0.125$ ,  
from: T. Schoof. PhD thesis, ITAP, CAU,  
2016. p. 115

# CPIMC for the HEG — Sign Problem



Kink-Explosion:  
physical or due to  
 $W(C) \rightarrow |W(C)|$   
???

Observation:  
Contributions from large  $K$   
cancel within errorbars.

figure:  $\Theta = 0.125$ ,  
from: T. Schoof. PhD thesis, ITAP, CAU,  
2016. p. 115

introduce factor  $V_K(\kappa, \delta)$  to partition sum:  $Z \rightarrow Z(\kappa, \delta) = \int C_K W(C_K) \cdot V_K(\kappa, \delta)$



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### Property:

- convergence parameter  $\kappa$ :

$$\lim_{\kappa \rightarrow \infty} V_K(\kappa, \delta) = 1$$

### Fermi-like function:

$$V_K(\kappa, \delta) = \left[ e^{-\delta(\kappa - K + \frac{1}{2})} + 1 \right]^{-1}$$

### Procedure:

- choose convenient  $\delta$  (e.g. = 1)
- simulate for various  $\kappa$
- extrapolate estimates to the limit  $\kappa \rightarrow \infty$

# CPIMC for the HEG — Variance Reduction: Kink Potential

introduce factor  $V_K(\kappa, \delta)$  to partition sum:  $Z \rightarrow Z(\kappa, \delta) = \int C_K W(C_K) \cdot V_K(\kappa, \delta)$

## Property:

- convergence parameter  $\kappa$ :

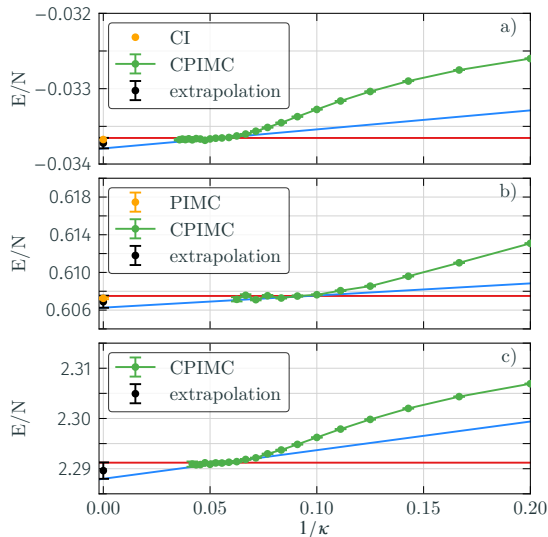
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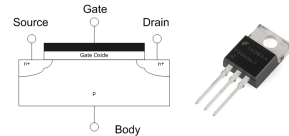
implemented and tested for qm. coupled harmonic oscillators

HEG solved at finite temperatures

exact data for WDM regime

x include relativistic effects for  $r_s \rightarrow 0$ ,  $\beta > 1$

x 2D Electron Gas



MOSFET realize 2DEG in inversion mode<sup>1</sup>

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<sup>1</sup>[https://en.wikipedia.org/wiki/Two-dimensional\\_electron\\_gas](https://en.wikipedia.org/wiki/Two-dimensional_electron_gas)

MUSE

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SIMULATION  
*Zero*