

第12回学習物理領域セミナー + 第64回DLAP

Fermi Machine

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Acknowledgements

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科研費
KAKENHI

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何が問題か？

量子多体問題の基底状態： $H|\Psi_0\rangle = E_0|\Psi_0\rangle$

$|\Psi_0\rangle = \sum_n c_n |\varphi_n\rangle$ N 粒子(N サイト)系のヒルベルト空間 $e^{\alpha N}$ ；

NPハード問題

系統的に精度を向上させられるアルゴリズムで
多項式時間(メモリ) $\propto N^\beta$ の範囲で、
如何にうまく基底状態を近似し、新しい物理を明らかにするか？

波及効果

物理学（高エネルギー、原子核、物性）、化学、生物学、
種々の工学的課題

既存の量子系の数値計算手法のあれこれ

量子モンテカルロ法

Feynman path integral

$$Z = \text{Tr} \exp[-\beta H] = \sum_x \langle x | \exp[-\beta H] | x \rangle$$

負符号問題

変分波動関数法 — 基底状態や励起状態波動関数を求める

密度行列繰込み群DMRG

テンソルネットワーク

変分モンテカルロ(VMC)

などなど

最近, ボルツマンマシン,
CNN, transformer などの
人工NNの開発が急速

変分不等式

$$E_g \leq \frac{\langle \Phi_\alpha | H | \Phi_\alpha \rangle}{\langle \Phi_\alpha | \Phi_\alpha \rangle}$$

強い多体相関（強相関）や量子もつれを伴う
チャレンジ課題をいかにうまく扱うか？
機械学習はどんな貢献ができるか？

機械学習（人工NN）を用いた全く新しい量子多体ソルバーの開発

1. 既存のBoltzmann machine, CNN, transformerなどでは
隠れ変数としてほとんど、イジング変数等の古典変数を採用している。
その代わりに隠れ変数としてフェルミオンを採用し,
「フェルミマシン」を構築する。
2. フェルミマシンのアルゴリズムに、銅酸化物等で実験検証が進む
電子の分数化の概念と理解を実体化することで,
強相関多体フェルミオンの示す創発構造を予め具現化して利用し,
量子もつれを効率的に取り込むことを狙う。

Introduction

既存のNN変分波動関数の例: RBMによる波動関数の表現

Carleo, Troyer, Science 355, 602 (2017)



Example; spin-1/2 Heisenberg model

$$H = \sum_{i,j} J_{i,j} \mathbf{S}_i \cdot \mathbf{S}_j$$

$$H_{\text{RBM}} = \sum_i a_i S_i^z + \sum_{i,j} W_{i,j} S_i^z h_j + \sum_j b_j h_j$$

h_j : 擬イジングスピノン (binary)

$Z_S = \text{Tr}_h \exp[-\beta_{\text{RBM}} H_{\text{RBM}}]$ ボルツマン重み

部分対角和を取った分配関数 → 基底状態波動関数の重み

$$Z_S = \exp[-a_i \beta_{\text{RBM}} S_i^z] \prod_k 2 \cosh \left[\beta_{\text{RBM}} (b_k + \sum_i W_{ik} S_i^z) \right] \rightarrow \mathcal{N}$$

$|\Psi\rangle = \mathcal{N} \sum_y |y\rangle$, 量子もつれのない単純な等分配状態
(温度 ∞ 状態)に作用させる

$\alpha = \{a, b, W\}$; 変分パラメタ

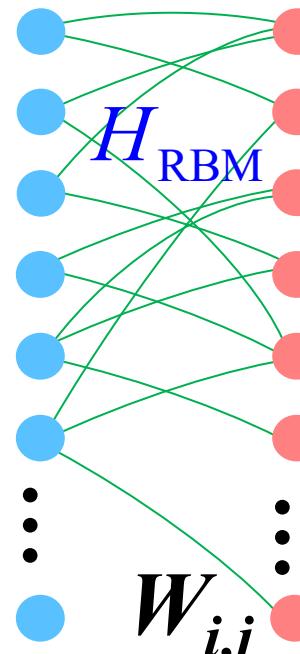
H の期待値 $\langle H \rangle = \langle \Psi | H \sum |x\rangle \langle x | \Psi \rangle / \langle \Psi | \Psi \rangle$

を下げるよう a, b, W を訓練する
→ 甘利による自然勾配法を用いる。

physical variables

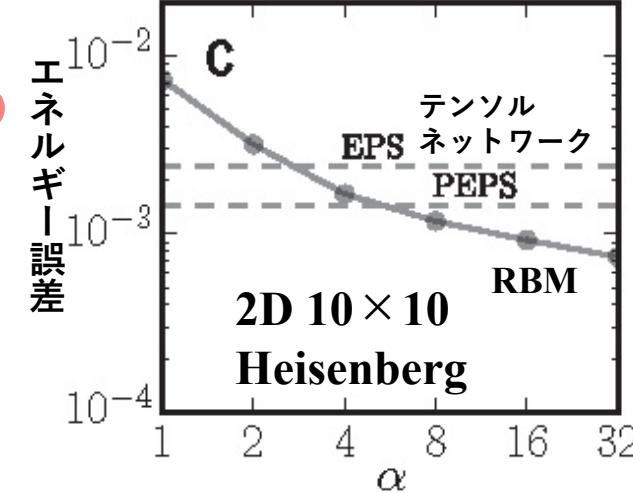
$$\mathbf{S}_i^z$$

⋮



hidden variables
 h_j

$$\mathbf{b}$$



$$\alpha = \frac{\text{隠れ変数の数}}{\text{物理変数の数}}$$

MC sampling

$$\mathbf{x} = \{S_1^z, S_2^z, \dots, S_N^z\}$$

チャレンジである遍歴電子系や
フラストレーションのあるスピン系
では精度に限界

従来の変分波動関数による基底状態表現の例, VMC

Gross, Sorella, Tahara & Imada

$$|\psi\rangle = \mathcal{L}_L \mathcal{P}_J \mathcal{P}_{\text{d-h}}^{\text{ex.}} \mathcal{P}_G \mathcal{L}^{S=0} |\phi_{\text{pair}}\rangle$$

$$|\phi_{\text{pair-product}}\rangle = \left[\sum_{ij} f_{ij} c_{i\sigma}^\dagger c_{j\sigma'}^\dagger \right]^{N/2} |0\rangle$$

pair-product (ペア積)
geminal ; 量子化学
HFB ; 原子核

Gutzwiller factor

$$\mathcal{P}_G = \exp \left[-g \sum_i n_{i\uparrow} n_{i\downarrow} \right]$$

long-ranged Jastrow factor

$$\mathcal{P}_J = \exp \left[-\frac{1}{2} \sum_{i \neq j} v_{ij} n_i n_j \right]$$

**quantum number
projection**

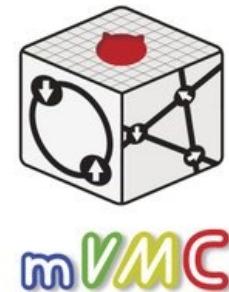
$$\mathcal{L}^S = \frac{2S+1}{8\pi^2} \int d\Omega P_S(\cos \beta) \hat{R}(\Omega)$$

spin, momentum, point group....

cf. 原子核計算

Open source: mVMC

Comput. Phys. Commun. 235, (2019) 447



M. IMADA

Combined VMC and Neural Network

Nomura, Darmawan, Yamaji, Imada
PRB 96, 205152 (2017)



ペア積波動関数+RBM

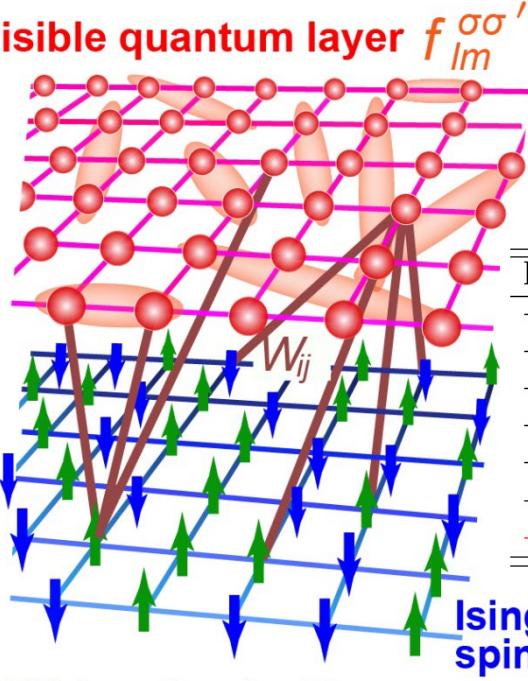
$$|\Psi\rangle = \mathcal{P} \mathcal{L}^{K=0} \mathcal{L}^{C_4} \mathcal{N} \sum |x\rangle \langle x| \mathcal{L}^{S=0} |\phi_{\text{pair-product}}\rangle \quad \text{RBM+PP}$$

RBM+PP architecture

$$|\phi_{\text{pair-product}}\rangle = \left[\sum_{ij}^x f_{ij} c_{i\sigma}^\dagger c_{j\sigma'}^\dagger \right]^{N/2} |0\rangle \quad \begin{array}{l} \text{量子もつれの取り込み} \\ \text{node構造の最適化} \\ \text{系統的改良の可能性} \end{array}$$

$$\mathcal{N} = \prod_k 2 \cosh(\beta(b_k + \sum_i W_{ik} S_i^z))$$

$$\mathcal{N} = \prod_{k,\sigma} 2 \cosh(\beta(b_k + \sum_i W_{i\sigma k} (2n_{i\sigma} - 1)))$$

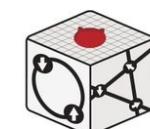


| Energy per site | Wave function | Reference |
|---------------------|----------------------|------------------------------|
| -0.494757(12) | Neural quantum state | Szabo <i>et al.</i> (2020) |
| -0.49516(1) | CNN | Choo <i>et al.</i> (2019) |
| -0.49521(1) | VMC($p=0$) | Hu <i>et al.</i> (2013) |
| -0.49530 | DMRG | Gong <i>et al.</i> (2014) |
| -0.49575(3) | RBM-fermionic w.f. | Ferrari <i>et al.</i> (2019) |
| -0.497549(2) | VMC($p=2$) | Hu <i>et al.</i> (2013) |
| -0.497629(1) | RBM+PP | present study |

NNが強結合超伝導やスピニ液体などの挑戦課題の解明に
使える実用最高レベルの精度に

RBM+PPを実装したVMCオープンソースコード

<https://github.com/issp-center-dev/mVMC/releases/tag/v1.3.0>



LADA

フラストレーションの大きい
正方格子 J_1 - J_2 ハイゼンベルク
模型 ($J_2/J_1 = 0.5$)
のベンチマーク

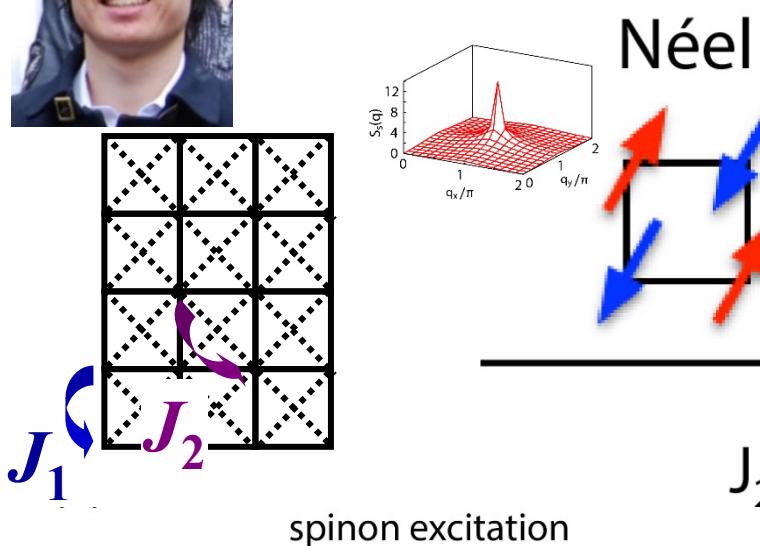
$b, W, f;$
variational parameter
 S, n ; **physical variable**

Fractionalization in J_1 - J_2 model

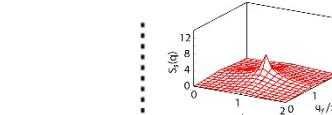
Y. Nomura, M. Imada, PRX, 11, 031034 (2021)



$$H = \sum_{i,j} J_{i,j} \mathbf{S}_i \cdot \mathbf{S}_j$$

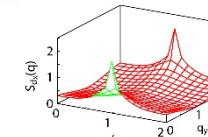


continuous



Néel

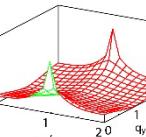
continuous



Spin
Liquid

長距離もつれ

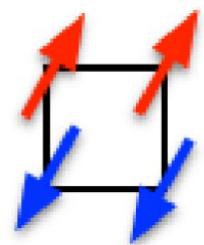
1st order



VBS

valence bond solid

Stripe



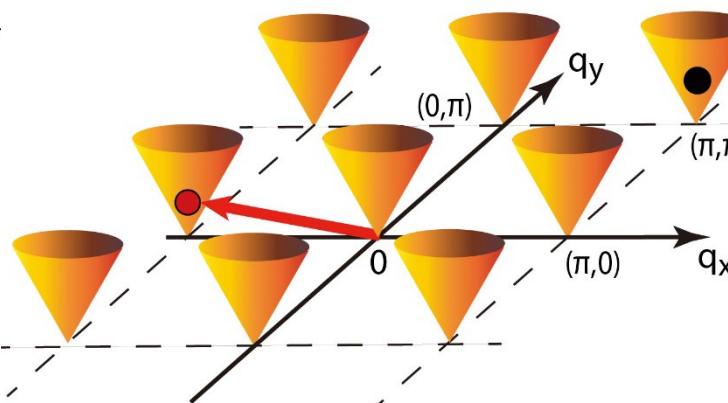
$$J_2^{\text{Néel}} \approx 0.49$$

$$J_2^{\text{VBS}} \approx 0.54$$

$$J_2^{\text{V-S}} \approx 0.61$$

spinon excitation

spin excitation



基本励起がスピン
ではなく
スピンが分裂した
スピノンで表される。

M. IMADA

Neural network for quantum many-body solver

However,....

hidden variables in RBM are so far classical (Ising spins).

If hidden variables are quantized and replaced by fermions,
they may represent fermionic entanglement and
node structure more efficiently.

To implement the correlation effects in the Fermi machine,
a key concept is the fermion fractionalization,
where strong correlation splinters a fermion (electron)
into entangled emergent multi-fermions.

$$H_{\text{TCFM}} = \sum_k \left[\varepsilon_c(k) c_{k\sigma}^\dagger c_{k\sigma} + \varepsilon_d(k) d_{k\sigma}^\dagger d_{k\sigma} \right] + \sum_{k,\sigma} \Lambda(k) (c_{k\sigma}^\dagger d_{k\sigma} + \text{H.c.})$$

This fractionalization has experimental & theoretical supports
indeed in the cuprate superconductors.

Hidden variables representing the entanglement/correlation
should be related to the fermionic self-energy representation.

カギとなるコンセプトはどんな分数化か？

分数化のいろいろ

1次元系のスピン電荷分離 Tomonaga, Luttinger

ハドロン \leftrightarrow クォーク Gell-Mann, Zweig, Ne`eman

ポリアセチレンソリトン SSH

分数量子ホール効果 Laughlin, Jain....

スレーブボソン、スレーブフェルミオン

Kotliar, Ruckenstein, Sachdev.....

どれとも異なる新奇な分数化：

1つのフェルミオンから複数のフェルミオンが創発

Signature of Fractionalization: Structure of Self-Energy in SC State

1電子グリーン関数

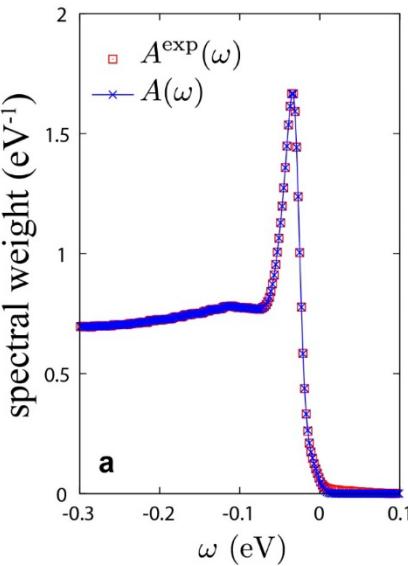
$$G(k, \omega) = \frac{1}{\omega - \varepsilon_0(k) - \Sigma(k, \omega)}$$

Σ ; 自己エネルギー

Interaction effect

$$A(k, \omega) = -\frac{1}{\pi} \text{Im} G(k, \omega) \quad \text{スペクトル関数 : 観測可能}$$

超伝導状態



ARPES data
角度分解光電子分光
Kondo *et al.* Nat. Phys.
7, 21 (2011)

$$G_{11}^{\text{nor}}(\mathbf{k}, \omega) = \left[\omega + \mu - \varepsilon_{\mathbf{k}} - (\Sigma^{\text{nor}}(\mathbf{k}, \omega) + W(\mathbf{k}, \omega)) \right]^{-1}$$

$$W(\mathbf{k}, \omega) = \frac{\Sigma^{\text{ano}}(\mathbf{k}, \omega)^2}{\omega - \mu + \varepsilon_{\mathbf{k}} + \Sigma^{\text{nor}}(\mathbf{k}, -\omega)^*}$$

W ; interaction effect from superconducting part

$A(k, \omega) \rightarrow \Sigma^{\text{nor}}, \Sigma^{\text{ano}}$ を別々に抽出 ; 逆問題

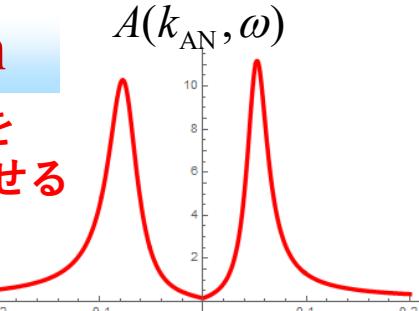
M. IMADA

Prediction by fractionalization: Self-energy peak cancellation

$$H_{\text{TCFM}} = \sum_k \left[\varepsilon_c(k) c_{k\sigma}^\dagger c_{k\sigma} + \varepsilon_d(k) d_{k\sigma}^\dagger d_{k\sigma} \right] + \sum_{k,\sigma} \Lambda(k) (c_{k\sigma}^\dagger d_{k\sigma} + \text{H.c.})$$

*Gのゼロを
容易に表せる*

$$G_c = \frac{1}{(\omega - \varepsilon_c) - \Sigma^{\text{nor}}(\mathbf{k}, \omega)}$$



$$A(k, \omega) = -\frac{1}{\pi} \text{Im } G^{\text{nor}}(k, \omega)$$

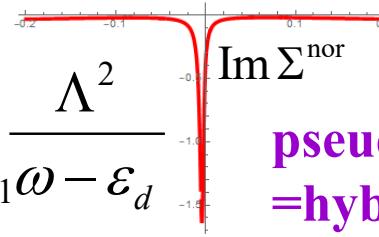
How about SC? Sakai, Civelli, Imada
PRL 116, 057003 (2016)

Extension of TCFM to SC phase

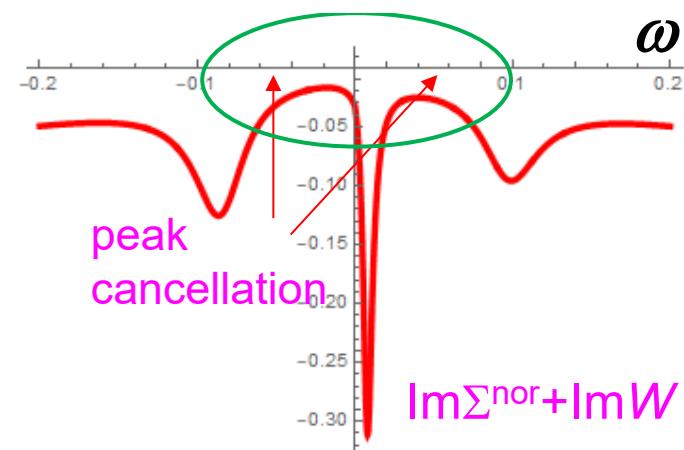
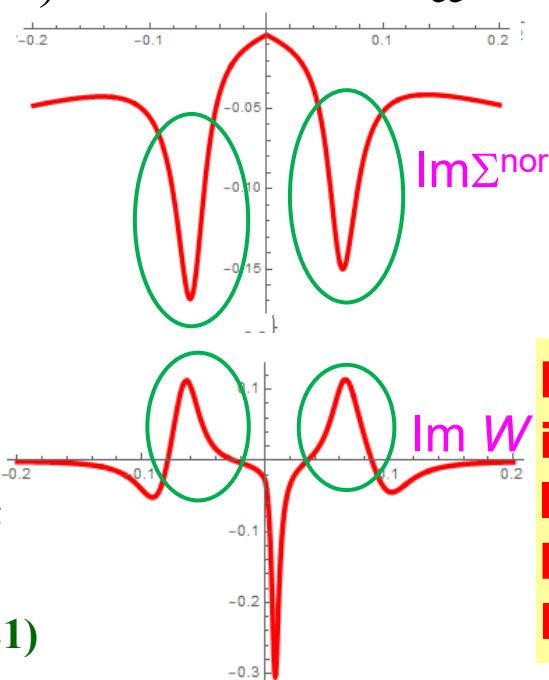
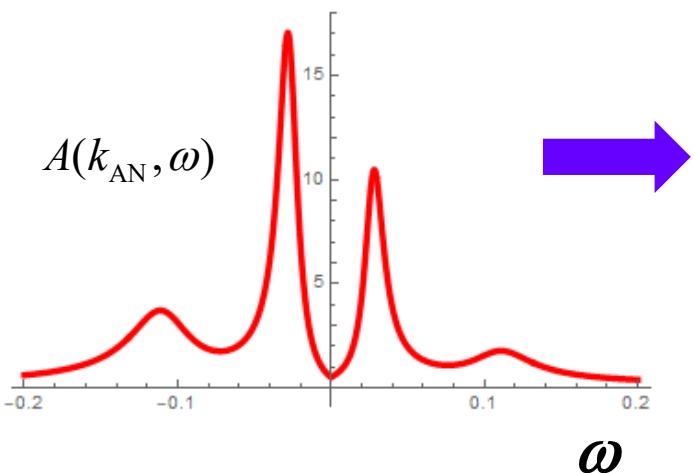
$$G_c^{\text{nor}}(\mathbf{k}, \omega) = \left[\omega + \mu - \varepsilon_{\mathbf{k}} - \left(\Sigma^{\text{nor}}(\mathbf{k}, \omega) + W(\mathbf{k}, \omega) \right) \right]^{-1}$$

$$W(\mathbf{k}, \omega) = \frac{\Sigma^{\text{ano}}(\mathbf{k}, \omega)^2}{\omega - \mu + \varepsilon_{\mathbf{k}} + \Sigma^{\text{nor}}(\mathbf{k}, -\omega)^*}$$

$$\Sigma^{\text{nor}}(\mathbf{k}, \omega) = \frac{\Lambda^2}{\omega - \varepsilon_d}$$



pseudogap
=hybridization gap
→銅酸化物擬ギャップ



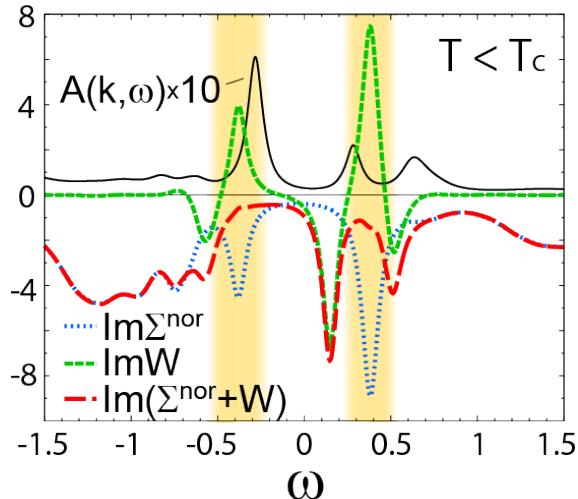
Prominent peaks emerge
in $\text{Im}\Sigma^{\text{nor}}$ and $\text{Im}\Sigma^{\text{ano}}$ and
 $\text{Im}\Sigma^{\text{ano}}$ peak generates SC.
but they cancel in $A(k, \omega)$:
Fingerprint of fractionalization

Imada, Suzuki, JPSJ 88, 024701 (2019)
Imada J. Phys. Soc. Jpn. 90, 074702 (2021)

Indications of electron fractionalization in theories and experiments

1. cluster DMFT for SC Hubbard: Cancellation of normal and anomalous contributions to spectral function $A(k,\omega)$;

分数化に特有の性質

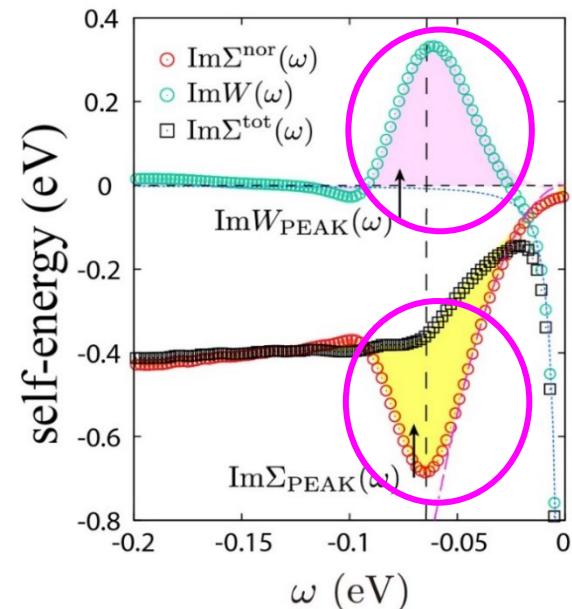


↔ Sakai *et al.* (2016)



Yamaji *et al.* (2021) →

隠れたフェルミオンの
存在の示唆



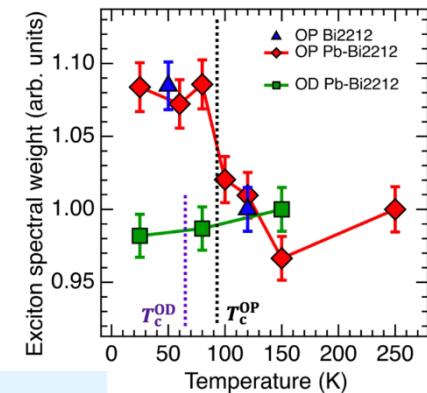
2. Same cancellation inferred from machine learning of cuprate ARPES

3. RIXS data supporting fractionalization

Imada (2021)

Singh *et al.* (2022) →

Two component fermion model (TCFM)



M. IMADA

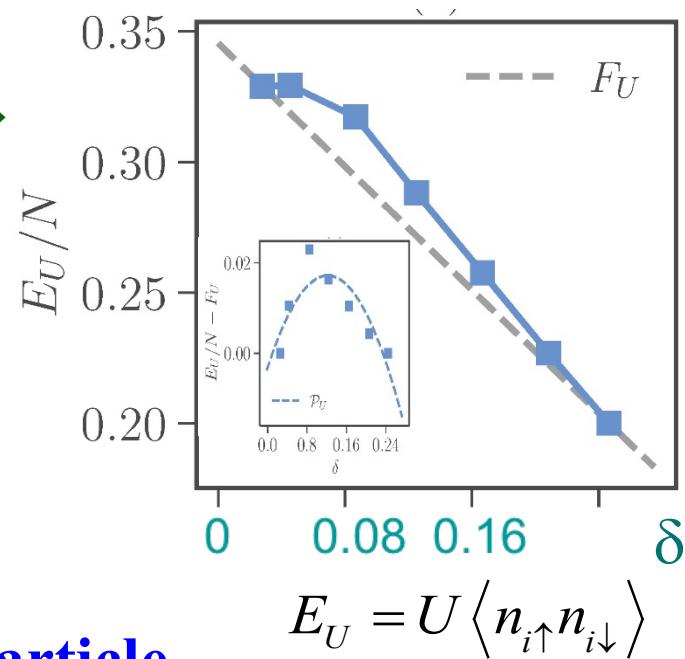
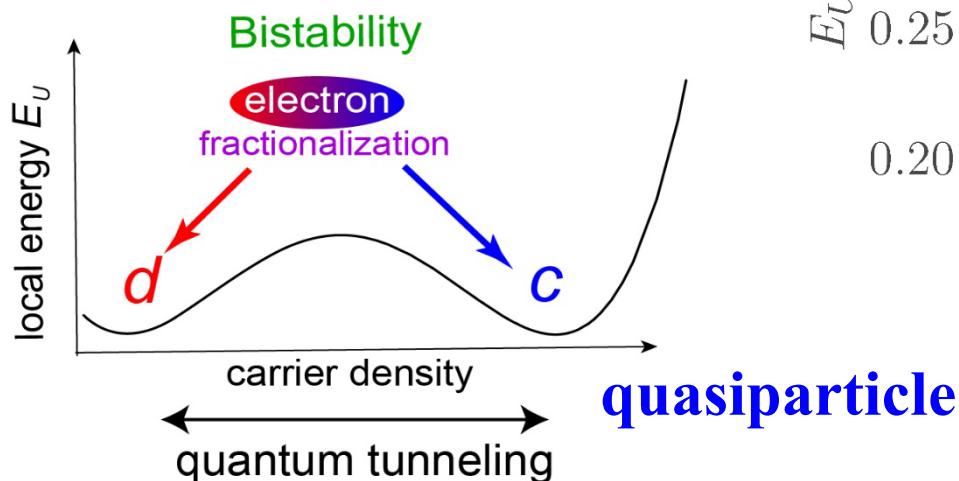
$$H = \sum_k [\varepsilon_c(k) c_{k\sigma}^\dagger c_{k\sigma} + \Lambda_k (c_{k\sigma}^\dagger d_{k\sigma} + \text{H.c.}) + \varepsilon_d(k) d_{k\sigma}^\dagger d_{k\sigma}]$$

Origin of electron fractionalization

Ab initio studies of cuprates: Schmid *et al.* (2023) →

局所斥力によって分数化と局所有効引力が創発する

dark fermion d
= detected as
incoherent part
of electron



1成分強相関電子の c と d への分数化

$$\text{TCFM} \quad H = \sum_k [\varepsilon_c(k) c_{k\sigma}^\dagger c_{k\sigma} + \Lambda_k (c_{k\sigma}^\dagger d_{k\sigma} + \text{H.c.}) + \varepsilon_d(k) d_{k\sigma}^\dagger d_{k\sigma}]$$

New fractionalization picture

Fractionalization of a fermion into two fermions associated with trend to 1st-order transition; phase separation replaced by quantum resonant, entangled and spatially uniform state stabilized by quantum tunneling

cf. slave boson, slave fermion; Kotliar-Ruckenstein, Lee-Nagaosa, Read-Sachdev

What is fractionalization? How to materialize?

Zhu, Zhu PRB 87, 085120 (2013), Sakai *et al.* PRB 94, 115130 (2016)

$$\mathcal{H} = Un_{\sigma}n_{-\sigma}$$

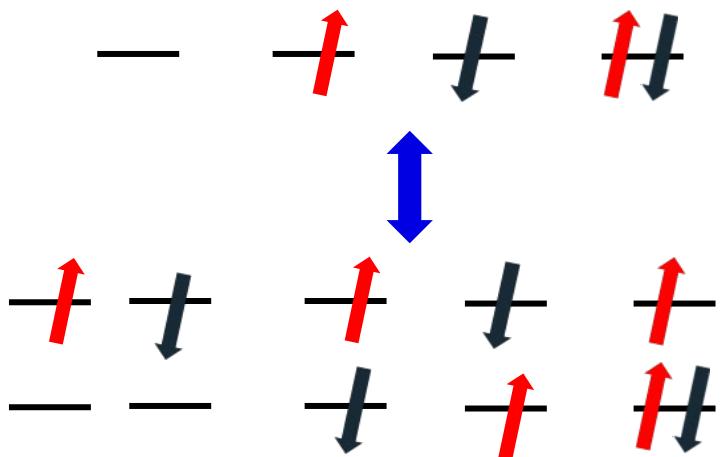
Hubbard in the atomic limit (1-site Hubbard)

$$n_{\sigma} = c_{\sigma}^{\dagger}c_{\sigma}$$

$$\tilde{c}_{\sigma} \doteq c_{\sigma}$$

$$\tilde{d}_{\sigma} \doteq c_{\sigma}(1 - 2n_{c-\sigma})$$

$$\begin{aligned} -\tilde{c}_{\sigma}^{\dagger}\tilde{d}_{\sigma} &= -c_{\sigma}^{\dagger}c_{\sigma}(1 - 2n_{-\sigma}) \\ &= -n_{\sigma} + 2n_{\sigma}n_{-\sigma} \end{aligned}$$



Two-component fermion model (TCFM)

$$H = \sum_{\sigma} [\mu_{\tilde{c}} \tilde{c}_{\sigma}^{\dagger} \tilde{c}_{\sigma} + \mu_{\tilde{d}} \tilde{d}_{\sigma}^{\dagger} \tilde{d}_{\sigma} + \Lambda (\tilde{c}_{\sigma}^{\dagger} \tilde{d}_{\sigma} + h.c.)]$$

$$\mu_{\tilde{c}} = \mu_{\tilde{d}} = -\Lambda = \frac{U}{2}$$

\tilde{d}_{σ} is identified as the dark fermion

\tilde{c}, \tilde{d} : fermion operator,

orthogonal as an average at half filling $\langle n_{\sigma} \rangle = 1$:

exact fermion anticommutation of \tilde{c} and \tilde{d}

$$\left\langle \left[\tilde{c}, \tilde{d}^{\dagger} \right]_{+} \right\rangle = 0, \quad \left\langle \left[\tilde{d}, \tilde{d}^{\dagger} \right]_{+} \right\rangle = 1, \dots$$

Restrict Hilbert space
within $n=1,2,3$ for TCFM.

Mott gap interpreted by TCFM

Imada, Suzuki, JPSJ 88, 024701 (2019)

diagonalization of H

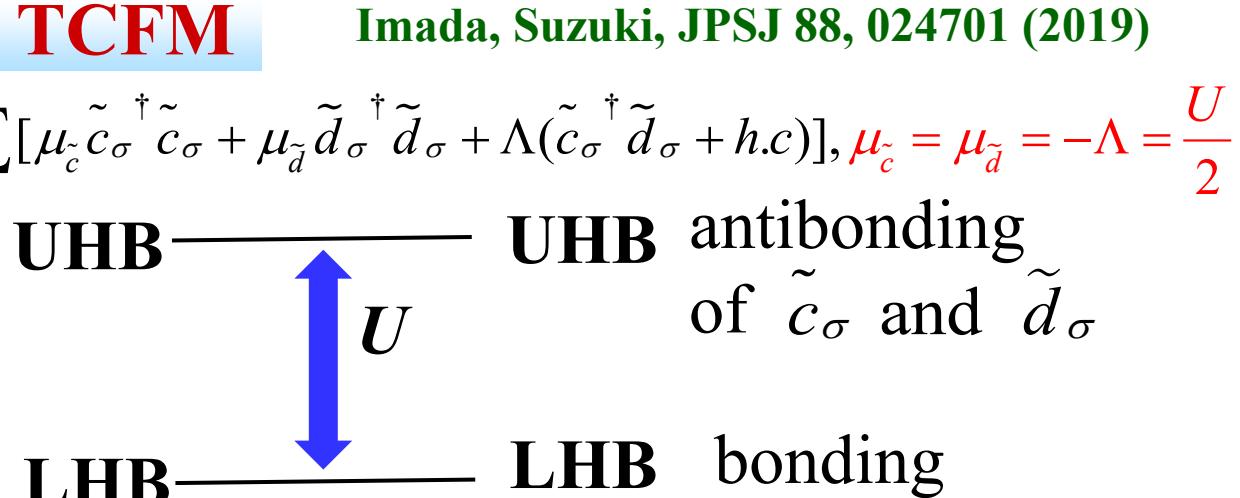
$$\Rightarrow \frac{1}{2}(\tilde{c}_\sigma^\dagger - \tilde{d}_\sigma^\dagger) = c_\sigma^\dagger n_{-\sigma}$$

$$\frac{1}{2}(\tilde{c}_\sigma^\dagger + \tilde{d}_\sigma^\dagger) = c_\sigma^\dagger (1 - n_{-\sigma})$$

$$\tilde{b}_\sigma^\dagger = (\tilde{c}_\sigma^\dagger + \tilde{d}_\sigma^\dagger)/\sqrt{2}, \quad \tilde{a}_\sigma^\dagger = (\tilde{c}_\sigma^\dagger - \tilde{d}_\sigma^\dagger)/\sqrt{2}$$

$$H = \sum_\sigma [E_{\tilde{b}} \tilde{b}_\sigma^\dagger \tilde{b}_\sigma + E_{\tilde{a}} \tilde{a}_\sigma^\dagger \tilde{a}_\sigma], \quad E_{\tilde{b}} = 0, \quad E_{\tilde{a}} = U$$

electron “fractionalization”



$$\text{UHB} \quad \text{LHB}$$

$$c_\sigma = c_\sigma(n_{-\sigma} + (1 - n_{-\sigma}))$$

Mott gap is a “hybridization gap”

gap (mass) generation without SSB
c.f. CO, AF, Nambu-Jona Lasinio mechanism

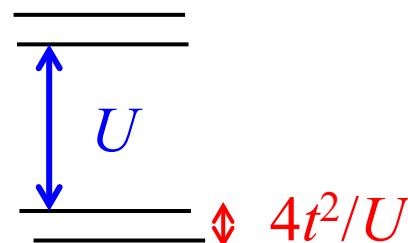
This mapping also works for nonzero t

2 sites

$$t_{\tilde{c}} = t, \quad t_{\tilde{d}} = -t$$

$$E_0 = \frac{U}{2}Q, \quad E_1 = 0, \quad E_2 = U, \quad E_3 = \frac{U}{2}P$$

$$P = 1 + \sqrt{1 + R^2}, \quad Q = 1 - \sqrt{1 + R^2}, \quad R = 4t/U$$



Correct Mott gap U and superexchange energy $4t^2/U$

W. IMADA

Underlying Concept of Fractionalization

Strong coupling superconductivity and quantum spin liquid
に見られる共通の機構

強い局所斥力が電荷ギャップを生む(Mottness)

↔ 2成分への分数化と成分間の混成が創発

自発的対称性の破れを伴わないギャップ（質量）形成

分数化と連動した創発局所引力の発生

(attraction by repulsion)

→ 強結合超伝導のクーパー対の起源

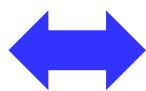
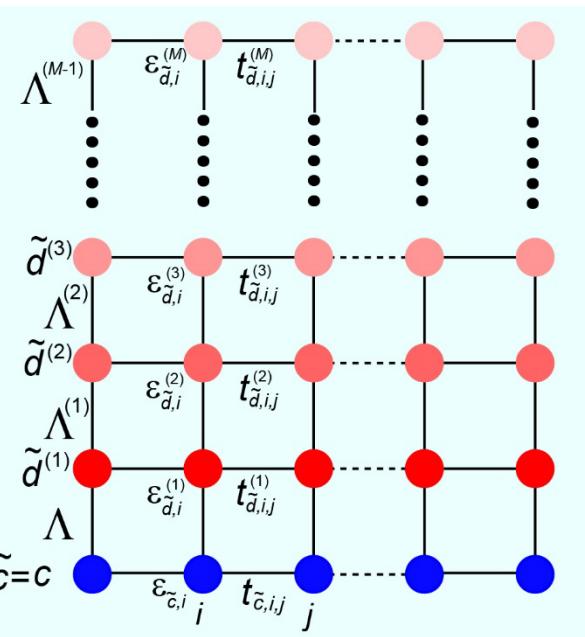
⇒ 量子スピン液体のスピン分数化の起源

Fermi machineのアルゴリズムと性質

Neural network algorithm based on fractionalization

$$H = \sum_k [\mu_{\tilde{c}}(k) \tilde{c}_{k\sigma}^\dagger \tilde{c}_{k\sigma} + \Lambda_k (\tilde{c}_{k\sigma}^\dagger \tilde{d}_{k\sigma}^{(1)} + \text{H.c.})] \\ + \sum_m [\mu_{\tilde{d}}^{(m)}(k) \tilde{d}_{k\sigma}^{(m)\dagger} \tilde{d}_{k\sigma}^{(m)} + \Lambda_k (\tilde{d}_{k\sigma}^{(m)\dagger} \tilde{d}_{k\sigma}^{(m+1)} + \text{H.c.})]$$

MCFM (multi-component fermion model)



$$H = \sum_{k\sigma} \mu_c(k) c_{k\sigma}^\dagger c_{k\sigma} + U \sum_i n_{i,\sigma} n_{i,-\sigma}$$

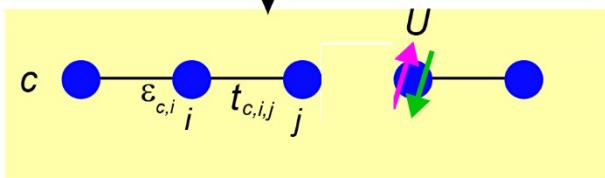
Hubbard

**novel neural network/machine learning
algorithm**

**Correspondence between interacting fermions
and noninteracting MCFM**

**RBM; イジング相互作用
FM; もつれを生む、混成**

Hubbard
model



M. IMADA

Procedure of Fermi machine

1. For given MCFM parameters, calculate the ground state

w.f. of the MCFM $|\Psi_0^{\text{MCFM}}\rangle$

2. Map $|\Psi_0^{\text{MCFM}}\rangle$ to the w.f. of the Hubbard $|\Psi_0^{\text{Hub}}\rangle$ by the mapping rule

$$\tilde{d}_{\ell\sigma}^{(m)} \doteq \tilde{d}_{\ell\sigma}^{(m-1)} (1 - 2n_{\tilde{d}_{\ell,-\sigma}^{(m-1)}}), \quad \dots$$

$$\tilde{d}_{\ell\sigma}^{(1)} \doteq c_{\ell\sigma} (1 - 2n_{c_{\ell,-\sigma}})$$

$$|\Psi_{0\sigma}^{\text{Hub}}\rangle = \prod_q^{k_F} \left[\sum_{m=0}^M \alpha_{m,q} \sum_{\ell} \frac{e^{iq\ell}}{\sqrt{N_s}} c_{\ell,\sigma}^\dagger (1 - 2n_{\ell,-\sigma})^m \right] |0\rangle$$

3. Calculate $\langle x | \Psi_0^{\text{Hub}} \rangle$ for a given MC sample

4. Perform MC sampling of $\langle x |$ and calculate

$$\Rightarrow \langle E_0 \rangle = \frac{\sum_x \langle \Psi_0^{\text{Hub}} | H | x \rangle \langle x | \Psi_0^{\text{Hub}} \rangle^{-1}}{N_{\text{sample}}}, \quad W(x) = |\langle x | \Psi_0^{\text{Hub}} \rangle|^2$$

$$\langle E_0 \rangle = \frac{\langle \Psi_0^{\text{Hub}} | H | \Psi_0^{\text{Hub}} \rangle}{\langle \Psi_0^{\text{Hub}} | \Psi_0^{\text{Hub}} \rangle}$$

Monte Carlo estimate

5. Optimize the MCFM parameters to lower $\langle E_0 \rangle$

Benchmarks

atomic limit $\mu_{\tilde{c}} = \mu_{\tilde{d}} = -\Lambda = \frac{U}{2}$

2 sites $t_{\tilde{c}} = t, \quad t_{\tilde{d}} = -t$

4 sites

half filling

| | exact E_0 | E_0 by Fermi machine | $t_{\tilde{d}}$ | Λ | $\mu_{\tilde{c}}$ | $\mu_{\tilde{d}}$ |
|---------|-------------|------------------------|-----------------|-----------|-------------------|-------------------|
| $U = 4$ | -2.10275 | -2.10275 | 1.0 | -1.002 | 1.002 | 4.644 |
| $U = 8$ | -1.32023 | -1.32023 | 1.0 | -2.238 | 2.238 | 7.229 |
| | -1.32023 | -1.32023 | 2.0 | -3.330 | 3.330 | 10.731 |

hole doped

| | exact E_0 | E_0 by Fermi machine | Λ | $\mu_{\tilde{c}}$ | $\mu_{\tilde{d}}$ | $\lambda_{\uparrow,\uparrow}^{(M)} = \lambda_{\uparrow,\uparrow}^{(M)}$ | $\lambda_{\uparrow,\downarrow}^{(M)} = \lambda_{\downarrow,\uparrow}^{(M)}$ |
|---------|-------------|------------------------|-----------|-------------------|-------------------|---|---|
| $U = 8$ | -3.20775 | -3.20775 | -2.71408 | 2.71408 | 4.73336 | -0.35912 | -0.71824 |

Exact mapping

$$\Lambda_{\sigma\sigma'}^{(M)}[k] = \lambda_{\sigma\sigma'}^{(M)} \cos[k]$$

Relation to Self-Energy Expansion

Nakajima (1958) – Zwanzig – Mori(1965)

Roth (1969), Mancini, Matsumoto (1996)

Onoda,Imada (2001,2003)

$$\hat{\omega}A = [A, H]$$

$$\mathcal{P}_1 X = \frac{\langle [X, A^\dagger]_+ \rangle}{\langle [A, A^\dagger]_+ \rangle} A$$

$$i \frac{d}{dt} G(t) = \varepsilon^{(11)} G + \delta\hat{\omega}G$$

$$i \frac{d}{dt} A(t) \equiv \hat{\omega}A = \mathcal{P}_1 \hat{\omega}A + (1 - \mathcal{P}_1) \hat{\omega}A$$

$$= \tilde{\varepsilon}^{(11)} A + \delta\hat{\omega}A$$

$$i \frac{d}{dt} \delta\hat{\omega}A(t) = \tilde{\varepsilon}^{(21)} A + \tilde{\varepsilon}^{(22)} \delta\hat{\omega}A + \delta\hat{\omega}\delta\hat{\omega}A$$

$$i \frac{d}{dt} \delta\hat{\omega}\delta\hat{\omega}A(t) = \tilde{\varepsilon}^{(31)} A + \tilde{\varepsilon}^{(32)} \delta\hat{\omega}A + \tilde{\varepsilon}^{(33)} \delta\hat{\omega}\delta\hat{\omega}A + \dots$$

$$G(\omega) = \frac{1}{\omega - \varepsilon^{(11)} - \Sigma_1(\omega)}$$

$$\Sigma_1(\omega) = \frac{\varepsilon^{(21)}}{\omega - \varepsilon^{(22)} - \Sigma_2(\omega)}$$

.....

Continued fraction expansion

WADA

Heierarchy in TCFM and MCFM

$$H = \sum_k [\varepsilon_c(k) c_k^\dagger c_k + \Lambda(k) (c_k^\dagger d_k + h.c.) + \varepsilon_d(k) d_k^\dagger d_k]$$

$$G = (\omega - H)^{-1} = \begin{pmatrix} \omega - \varepsilon_c & -\nu \\ -\nu & \omega - \varepsilon_d \end{pmatrix}^{-1} = \frac{1}{(\omega - \varepsilon_c)(\omega - \varepsilon_d) - \nu^2} \begin{pmatrix} \omega - \varepsilon_d & \nu \\ \nu & \omega - \varepsilon_c \end{pmatrix}$$

$$G_{cc} = \frac{\omega - \varepsilon_d}{(\omega - \varepsilon_c)(\omega - \varepsilon_d) - \Lambda^2} = \frac{1}{(\omega - \varepsilon_c) - \frac{\Lambda^2}{\omega - \varepsilon_d}}$$

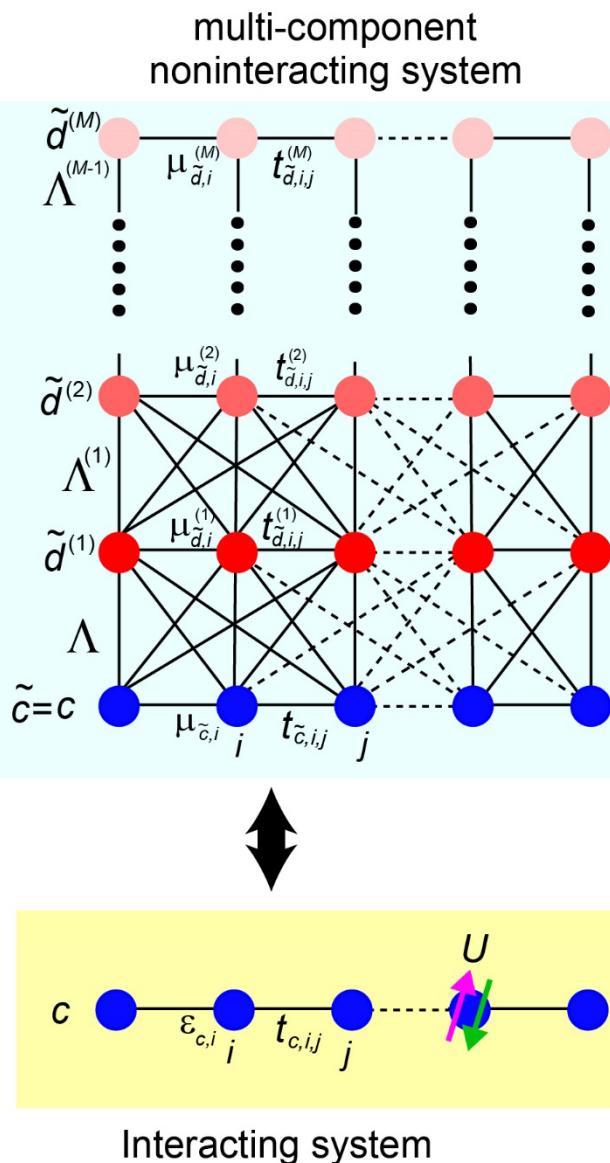
$$G = \frac{1}{\omega - \varepsilon_c - \Sigma_1} \quad \Sigma_1 = \frac{\Lambda^2}{\omega - \varepsilon_d} \rightarrow \frac{\Lambda^2}{\omega - \varepsilon_d - \Sigma_2} \quad \textbf{MCFM}$$

射影演算子法 (EOM)の連分数展開と同型

Sakai *et al.* (2016)

M. IWADA

Extension to nonlocal coupling



$$\tilde{d}^{(m)}{}_{\ell\sigma} \doteq \tilde{d}^{(m-1)}{}_{\ell\sigma} (1 - 2n_{\tilde{d}^{(m-1)}{}_{\ell,-\sigma}}), \quad \dots$$

$$\begin{aligned} \tilde{d}_{\ell\sigma}^{(m)\dagger} &\rightarrow \sum_{\ell'\sigma'} \frac{e^{iq(\ell'-\ell)}}{\sqrt{N_S}} \Theta_{\sigma',\sigma}^{(m)}(\ell' - \ell) \tilde{d}_{\ell\sigma}^{(m-1)\dagger} (-)^{n_{\tilde{d}^{(m-1)},\ell',\sigma'}} \\ &\rightarrow \Xi_{\sigma}^{(m)}(q, \ell) c_{\ell\sigma}^{\dagger} \\ \Xi_{\sigma}^{(m)}(q, \ell) &= \prod_{n=1}^m \left[\sum_{\ell_n, \sigma_n} \frac{e^{iq(\ell_n - \ell)}}{\sqrt{N_S}} \Theta_{\sigma_n, \sigma}^{(m-n+1)}(q, \ell_n - \ell) (-)^{mn_{\ell_n, \sigma_n}} \right], \\ \Xi_{\sigma}^{(0)}(q, \ell) &= 1 \end{aligned}$$

$$|\Psi_{0\sigma}^{\text{Hub}}\rangle = \prod_q^{k_F} \left[\sum_{m=0}^M \alpha_{mq} \sum_{\ell} \Xi_{\sigma}^{(m)}(q, \ell) c_{\ell\sigma}^{\dagger} \right] |0\rangle$$

Enhance the entanglement of distant electrons

運動方程式の方法(射影演算子やグラムシュミット法)
ではなく、非局所カップリングによる一般のNNと同じ戦略でrepresentabilityを高める。

M. IWADA

Physical Insight of Mapping

$$\tilde{c}_\sigma \doteq c_\sigma$$

$$\tilde{d}_\sigma \doteq c_\sigma(1 - 2n_{c-\sigma})$$

$$\tilde{d}_{i\sigma}^{(m)} \doteq \tilde{d}_{i\sigma}^{(m-1)}(1 - 2n_{\tilde{d}^{(m-1)} i, -\sigma})$$

Scattering and suppression of double occupation

$$\tilde{d}_{\ell\sigma}^{(m)\dagger} \rightarrow \sum_{\ell'\sigma'} \frac{e^{iq(\ell'-\ell)}}{\sqrt{N_S}} \Theta_{\sigma',\sigma}^{(m)}(\ell' - \ell) \tilde{d}_{\ell\sigma}^{(m-1)\dagger} (-)^{n_{\tilde{d}^{(m-1)}, \ell', \sigma'}}$$

**incorporate nonlocal correlation and entanglement
between electrons at (ℓ, σ) and (ℓ', σ')**

\Leftrightarrow nonlocal coupling in RBM, DBM

Jastrow factor in conventional VMC

M. TAKADA

Optimization of variational parameters α_k

Amari, Neural Comput. 10, 251 (1998).
Sorella PRB 64, (2001) 024512

Natural gradient (Stochastic reconfiguration) method Tahara, MI JPSJ 77 (2008) 114701

$$\left| \overline{\psi_\alpha} \right\rangle = \frac{1}{\sqrt{\langle \psi_\alpha | \psi_\alpha \rangle}} |\psi_\alpha\rangle$$

Consider $\alpha \rightarrow \alpha + \gamma$

$$\left| \overline{\psi_{\alpha+\gamma}} \right\rangle = \left| \overline{\psi_\alpha} \right\rangle + \sum_{k=1}^p \gamma_k \left| \overline{\psi_{k\alpha}} \right\rangle + O(\gamma^2)$$

$$\left| \overline{\psi_{k\alpha}} \right\rangle \doteq \frac{\partial}{\partial \alpha_k} \left| \overline{\psi_\alpha} \right\rangle = \frac{1}{\sqrt{\langle \psi_\alpha | \psi_\alpha \rangle}} \left(\frac{\partial}{\partial \alpha_k} |\psi_\alpha\rangle - \frac{\langle \psi_\alpha | \left(\frac{\partial}{\partial \alpha_k} \right) | \psi_\alpha \rangle}{\langle \psi_\alpha | \psi_\alpha \rangle} |\psi_\alpha\rangle \right)$$

$$\Delta_{\text{norm}}^2 \equiv \left\| \left| \overline{\psi_{\alpha+\gamma}} \right\rangle - e^{-\Delta\tau H} \left| \overline{\psi_\alpha} \right\rangle \right\|^2 \approx \left\| \left| \overline{\psi_{\alpha+\gamma}} \right\rangle - (1 - \Delta\tau H) \left| \overline{\psi_\alpha} \right\rangle \right\|^2$$

$$= \Delta\tau^2 \left\langle \overline{\psi_\alpha} \right| H^2 \left| \overline{\psi_\alpha} \right\rangle + 2\Delta\tau \gamma_k g_l + \sum_{k,l=1}^p \gamma_k \gamma_l S_{kl}$$

$$S_{kl} = \frac{\partial^2}{\partial \alpha_k \partial \alpha_l} \left\langle \overline{\psi_\alpha} \right| \overline{\psi_\alpha} \right\rangle, \quad g_l = \frac{\partial E}{\partial \alpha_l}, \quad E = \left\langle \overline{\psi_\alpha} \right| H \left| \overline{\psi_\alpha} \right\rangle$$

$$\gamma_k = -\Delta\tau \sum_{l=1}^p S_{kl}^{-1} g_l$$

If $\gamma_k = -\Delta t g_k$, $g_k = \frac{\partial E}{\partial \alpha_k}$
⇒ steepest descent method

If $\gamma_k = -\Delta t \sum_{l=1}^p (h^{-1})_{kl} g_l$,
 $h_{kl} = \frac{\partial^2 E}{\partial \alpha_k \partial \alpha_l}$ **⇒ Newton method**

虚時間発展 $e^{-\tau H} \left| \overline{\psi_\alpha} \right\rangle$
 で基底状態へ漸近

**variation of
the wave function**

**S^{-1} : stabilize against singular change
of the wavefunction**

Optimization of variational parameters α

${}^T A$: transpose of A

$$|\Psi_{0\sigma}^{\text{Hub}}\rangle = \prod_q^{k_F} \left[\sum_{m=0}^M \alpha_{mq} \sum_{\ell} \Xi_{\sigma}^{(m)}(q, \ell) c_{\ell\sigma}^{\dagger} \right] |0\rangle$$

1次摂動の表式

$$\delta |\Psi_0^{\text{MCFM}}\rangle = \sum_{n \neq 0} \frac{\langle \Psi_n^{\text{MCFM}} | \delta H_{\text{MCFM}} | \Psi_0^{\text{MCFM}} \rangle}{E_n - E_0} |\Psi_n^{\text{MCFM}}\rangle$$

を使えば H_{MCFM} の変分パラメタを微小量変えた時の波動関数の変化が陽に計算できる。

一方、ハミルトニアンを対角化するユニタリ変換 U_{α} の微小な変化量 δU_{α} が自然勾配法から求められれば

$$H_{\text{MCFM}} = U_{\alpha} D_{\text{MCFM}} (U_{\alpha})^{-1}, \quad \delta U = U_{\alpha+\gamma} - U_{\alpha}, \quad \delta(U_{\alpha}^{-1}) = \delta({}^T U_{\alpha}) = {}^T (\delta U)$$

$$\delta H_{\text{MCFM}} = H_{\text{MCFM}}(\xi + \delta) - H_{\text{MCFM}}(\xi) = (\delta U) D^T U + U D (\delta^T U)$$

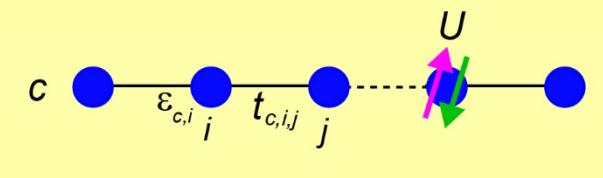
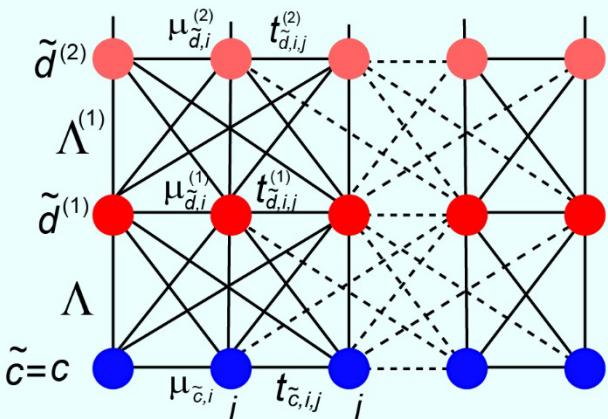
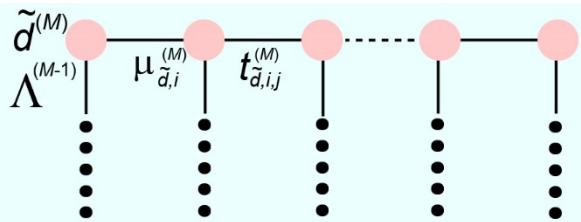
より、 H_{MCFM} の変分パラメタの変化量がわかる。

H_{MCFM} を対角化する
ユニタリ変換

$$U_{\alpha} = \begin{pmatrix} \alpha_{m=0, q=q_0}, \alpha_{m=0, q=q_1}, \alpha_{m=0, q=q_2}, \dots \\ \alpha_{m=1, q=q_0}, \alpha_{m=1, q=q_1}, \dots \\ \dots \end{pmatrix}$$

Representability

multi-component
noninteracting system



Interacting system

$$\mu_{\tilde{d},i}^{(m)} \rightarrow \infty \text{ for odd } m$$

偶数番目の層がすべて孤立

$$H = \sum_k [\mu_c(k) \tilde{c}_{k\sigma}^\dagger \tilde{c}_{k\sigma}] + \sum_n [\mu_{\tilde{d}}^{(n)}(k) \tilde{d}_{k\sigma}^{(n)\dagger} \tilde{d}_{k\sigma}^{(n)}] \quad n=2m$$

1. 偶数番目の各層のエネルギーが縮退するように各層のchemical potentialを選べば基底状態は各層の基底状態の任意の線形結合となる。
2. 任意の多体状態は1体状態の直積の線形結合で表せる。(1体状態の直積は多体状態の完全直交基底を張っている)

$$|\Psi_{0\sigma}^{\text{MCFM}}\rangle = \sum_{n=1}^{M/2} \alpha_n \prod_q^{k_F} |\Psi_{\sigma n}^{\text{MCFM}}(q)\rangle$$

$$|\Psi_{\sigma n}^{\text{MCFM}}(q)\rangle = \tilde{d}_{q\sigma}^{(n)\dagger} |0\rangle$$

十分大きな M を取れば、MCFMの基底状態は Hubbard (相互作用系) の多体基底状態を厳密に表わせる。(変分状態)

M. IMADA

Materialization of fractionalization by fermi machine

Interacting fermions

↔ Non-interacting multi-component fermions

cf. holographic correspondence in AdS-CFT,
hidden fermion \Leftrightarrow bath in DMFT, DMET
bulk-edge correspondence in topological matter
path int. d dim quantum \Leftrightarrow $d+1$ dim. classical

フェルミマシンの特長、優位性

- ★ ボルツマンマシンでイジング自由度を使って量子性を取り込むより効率的。
- ★ VMCで用いる波動関数で対称性を破らずにギャップ構造やグリーン関数のゼロを表わすのは厄介だが、フェルミマシンでは容易。

Outlook

- ★ Efficient optimization of MCFM variational parameters \rightarrow large systems
- ★ 隠れ層のホワイトボックス化; 強相関電子系の電子構造が隠れ量子変数の挙動からあぶりだされる可能性
- ★ Better neural network by mapping to HFB (geminal) states
by Pfaffian instead of SD?
- ★ Extension to non-periodic such as random systems

Thank you

M. MADA