Sparse sampling approach to efficient *ab initio* and many-body calculations at finite temperature

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ics of Conductive Multipole

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Green's function in physics



FIG. 1. (Color online) Left: Sketched Na_xCoO_2 diagram of dominating correlations and stable phases, based on Ref. 2. Right: *M* point ordering (top) and *K* point ordering (bottom) on the triangular lattice separating the lattice into a triangular sublattice (squares) and a kagome, honeycomb sublattice (thick lines).

Quantum embedding theoriesAccurate treatment of strong correlation in active spaceEmbedding in spaceEmbedding in energy



Dynamical mean-field theory, dynamical vertex approximation *etc*.

Constrained RPA etc.

Material calculations require efficient numerical treatment of one- and two-particle response functions.

Quantum embedding for quantum chemistry



T. N. Lan, A. A. Kananenka and D. Zgid, J. Chem. Theory Comput. 12, 4856 (2016)

Matsubara Green's function





Low-T scaling of data size

 $|\delta G(\tau=0)| < 10^{-8}$



Motivation

Storage
"IR" basis based on physical ground



Computation

Sparse sampling approach to solving diagrammatic equations (*GW*, Dyson, RPA, FLEX, ...)



Outline

Single-particle Green's function

- IR basis and sparse sampling
- Applications to *ab initio* calculations

Extension to two-particle Green's function

Matsubara Green's function and spectral function



The kernel knows typical structures appearing in $G(i\omega_n)$.

Intermediate representation (IR)

HS, J. Otsuki, M. Ohzeki, K. Yoshimi, PRB **96**, 035147 (2017) J. Otsuki, M. Ohzeki, HS, K. Yoshimi, PRE **95**, 061302(R) (2017)



Singular values

 $\Lambda \equiv \beta \omega_{\max}$



Compact representation of Green's function $G(i\omega) = \int_{-\omega_{\text{max}}}^{\omega_{\text{max}}} d\omega \ K^{\alpha}(i\omega, \omega)\rho(\omega) \quad \triangleleft \qquad G_l = -S_l^{\alpha}\rho_l$ $\equiv \sum_{i\omega} U_l^{\alpha}(i\omega)^* G(i\omega) \qquad \equiv \int_{-\omega_{\text{max}}}^{\omega_{\text{max}}} d\omega \rho(\omega) V_l^{\alpha}(\omega)$

 G_l decay as fast as $S_l!$



IR basis functions

 $\Lambda \equiv \beta \omega_{\max}$

 $U_l^{\rm F}(\tau) \equiv \sum U_l^{\rm F}(\mathrm{i}\omega_n) e^{-\mathrm{i}\omega_n \tau}$ $V_l^{\rm F}(\omega)$ l = 0l = 016 n14 $\Lambda = \beta \omega_{\text{max}}$ 1.0 $\Lambda = 100$ Legendre basis ($\Lambda \rightarrow 0$) 12 0.8 10 $\Lambda = 50$ 0.6 $\Lambda = 10$ 8 6 0.4 4 $\Lambda = 50$ 2 10 0.2 _ 0 l = 1l = 11.0 6 4 0.52 0.0 0 -0.5-1.0-1.5-60 -1 -1 () $\omega/\omega_{\rm max}$ $2\tau/\beta - 1$ Fine resolution at $\tau = 0, \beta$

Problems to be addressed in Part I

Storage "IR" basis based on physical ground $O(\log \beta)$



Computation
Sparse sampling approach to
diagrammatic equations (*GW*, Dyson, RPA, FLEX, DMFT...)



How to solve diagrammatic equation efficiently? Dyson equation $G(i\omega_n) = \frac{1}{i\omega_n - H - \Sigma(i\omega_n)}$

- Solve linear equation for $G(i\omega)$
 - $A(i\omega)G(i\omega_n) = 1 \quad \text{diagonal in } i\omega$ $A_{i\omega} \equiv i\omega_n H \Sigma(i\omega_n)$

Computational complexity: $O(\beta)$

• Solve linear equation for G_l

$$\sum_{l'=0}^{N} A_{ll'} G_l = 1_l$$

 $N \times N$ dense matrix $(N \propto \log \beta)$

Computational complexity: $O(\log^3 \beta)$

Asymptotically faster but not super efficient in practical calculation

Trick: sparse sampling $G_l = \sum_{l=1}^{\infty} U_l^{\alpha}(i\omega)^* G(i\omega)$ $i\omega = -i\infty$ What we want to compute Do we really need to know $G(i\omega)$ on all frequencies? eventually No. We do not have to compute this equation because frequency dependence of $G(i\omega)$ has only N degrees of freedom: $G(\mathrm{i}\omega) = \sum_{l=1}^{N} G_{l} U_{l}^{\alpha}(\mathrm{i}\omega) + \epsilon(N) \,.$ l=0 $\epsilon(N)$: exponentially small error Thus, we can determine G_l from $G(i\omega_n)$ only on appropriately selected *N* sampling frequencies.

Similar idea of sampling frequencies: T. Ozaki, PRB **75**, 035123 (2007), M. Kaltak and G. Kresse, PRB **101**, 205145 (2020) and so on.





How to solve Dyson equation

J. Li, M. Wallerberger, C.-N. Yeh, N. Chikano, E. Gull, <u>HS</u>, PRB 101, 035144 (2020)

1. Solve Dyson equation on sampling frequencies

$$G(\mathrm{i}\bar{\varpi}_k^\mathrm{F}) = \frac{1}{\mathrm{i}\bar{\varpi}_k^\mathrm{F} - H - \Sigma(\mathrm{i}\bar{\varpi}_k^\mathrm{F})} \qquad k = 1, \cdots, N$$

Computational complexity: $O(N_{orb}^3 \log \beta)$

2. Compute G_l from sampled values

$$\{G(\mathrm{i}\bar{\omega}_k^{\mathrm{F}})\} \to \{G_l\}$$

Fitting by applying pseudo inverse matrix

Computational complexity: $O(N_{orb}^2 \log^2 \beta)$

3. Evaluate $G(i\omega_n)/G(\tau)$ from G_l

Self-consistent GF2 & GW calculations

J. Li, M. Wallerberger, C.-N. Yeh, N. Chikano, E. Gull, <u>HS</u>, PRB 101, 035144 (2020)



The data always stay compact! (N<200)

Example: Molecules

J. Li, M. Wallerberger, C.-N. Yeh, N. Chikano, E. Gull, <u>HS</u>, PRB 101, 035144 (2020)

10 hydrogen atoms placed on a straight line with equal spacing *a*₀ cf. M. Motta *et al.*, PRX 7, 031059 (2017)

 $\beta = 1000 \ E_{\rm h}^{-1} \ (T \simeq 315.8 \ {\rm K})$





Reference data: $N \rightarrow +\infty$

Example: noble gas atoms with deep core states

J. Li, M. Wallerberger, C.-N. Yeh, N. Chikano, E. Gull, <u>HS</u>, PRB **101**, 035144 (2020) Chebyshev is no more doable. $\beta = 1000 E_{h}^{-1} (T \simeq 315.8 \text{ K})$



Example: GW of silicon crystal

J. Li, M. Wallerberger, C.-N. Yeh, N. Chikano, E. Gull, <u>HS</u>, PRB 101, 035144 (2020)



 $4 \times 4 \times 4$ mesh, $\beta = 1000 E_{h}^{-1} (T \simeq 315.8 \text{ K})$ GTH-DZVP basis of Gaussian orbitals (13 orbitals per Si atom)

Example: Migdal-Eliashberg theory

T. Wang, T. Nomoto, Y. Nomura, HS, J. Otsuki, T. Koretsune, and R. Arita, PRB 102, 134503 (2020)

Numerically demanding to estimate T_c of O(10) K considering the retardation effect from first principles

Linearized gap equation

$$\begin{split} \tilde{\lambda} \Delta_m(\boldsymbol{k}, i\omega_n) &= -\frac{T}{N_{\boldsymbol{k}}} \sum_{m'} \sum_{\boldsymbol{k}', i\omega_{n'}} \mathcal{K}_{mm'}(\boldsymbol{k} - \boldsymbol{k}', \underline{i\omega_n - i\omega_{n'}}) \\ &\times |G_{m'}(\boldsymbol{k}', i\omega_{n'})|^2 \Delta_{m'}(\boldsymbol{k}', i\omega_{n'}) \end{split}$$

Convolution can be performed efficiently using a sparse grid in τ . • Pairing interaction kernel $\mathcal{K}_{mm'} = \mathcal{K}_{mm'}^{el-ph} + \mathcal{K}_{mm'}^{C}$

• Renormalized electron Green's function

$$G_m(\boldsymbol{k}, i\omega_n) = \frac{1}{i\omega_n - \varepsilon_{m\boldsymbol{k}} - \Sigma_m(\boldsymbol{k}, i\omega_n)}$$

$$\Sigma_m(\boldsymbol{k}, i\omega_n) = -\frac{T}{N_{\boldsymbol{k}}} \sum_{m'} \sum_{\boldsymbol{k}', i\omega_{n'}} \mathcal{K}_{mm'}^{\text{el-ph}}(\boldsymbol{k} - \boldsymbol{k}', i\omega_n - i\omega_{n'}) G_{m'}(\boldsymbol{k}', i\omega_{n'})$$

Example: Migdal-Eliashberg equation

T. Wang, T. Nomoto, Y. Nomura, <u>HS</u>, J. Otsuki, T. Koretsune, and R. Arita, PRB **102**, 134503 (2020)



Converged results even for $T_c = O(10)$ K!

- Memory consumption 1/30
- Computational time 1/20

$$\Lambda = 10^5$$
 v.s. $N_{\omega} = 4096$

Open source software: irbasis

https://github.com/SpM-lab/irbasis

N. Chikano, K. Yoshimi, J. Otsuki, H. Shinaoka (2018) + M. Wallerberger (2019)

- Python and C++
- Step-by-step tutorial





N. Chikano, J. Otsuki, H. Shinaoka, PRB 98, 035104 (2018)

Outline

Single-particle Green's function

- IR basis and sparse sampling
- Applications to *ab initio* calculations

Extension to two-particle Green's function

LEWIN BOEHNKE AND FRANK LECHERMANN



FIG. 1. (Color online)^{Bethe: Salpeter equation} CoO₂ diagram of dominating correlations and stable phases, based on Ref. 2. Right: M point ordering (top) and K point ordering (bottom) on the triangular lattice separating the lattice into a triangular sublattice (squares) and a kagome, honeycomb sublattice (thick lines).

x = 0.30





x = 0.67

x = 0.30





Rich frequency structure



- Large size: $O(\beta^3)$
- More indices for spin, orbital, wave vector...



- IR approach frequency dependence
- Sparse sampling and tensor network representation

IR approach

HS, J. Otsuki, M. Ohzeki, K. Yoshimi, K. Haule, M. Wallerberger, E. Gull, PRB 97, 205111 (2018)



1. Sparse sampling

<u>HS</u>, D. Geffroy, M. Wallerberger, J. Otsuki, K. Yoshimi, E. Gull, J. Kuneš, SciPost Phys. 8, 012 (2020) m = -20

$$\Lambda = 10^3 \qquad X_{ijkl}^{\text{loc}}(i\omega_n, i\omega_{n'}; i\nu_m)$$





n

 1.5×10^5 sampling points $\ll 3 \times 10^9$ points in box

n

2. Dimensionality reduction by tensor network

HS, D. Geffroy, M. Wallerberger, J. Otsuki, K. Yoshimi, E. Gull, J. Kuneš, SciPost Phys. **8**, 012 (2020)

$$X_{ijkl}^{\text{loc}}(r, l_1, l_2, l_3) \simeq \sum_{d=1}^{D} x_{dr}^{(1)} x_{dl_1}^{(2)} x_{dl_2}^{(3)} x_{dl_3}^{(4)} x_{dijkl}^{(5)}$$

$$\int_{16}^{16} < 30 \qquad \text{Assumption: } D \text{ is small.}$$

• Further compactification of IR tensor





第一原理理 uting dynamic susceptibility in DMFT

Review: G. Kotliar et al., Rev. Mod. Phys. 78, 865 (2006)



Comparison with neutron scattering etc.

ーティング

More realistic systems



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D. Geffroy, J. Kaufmann, A. Hariki, P. Gunacker, A. Hausoel and J. Kuneš, PRL **122**, 127601 (2019)



• Worm sampling in Legendre basis by ALPS/CT-HYB 40 hours with 840 processes



Sparse QMC sampling



Solving Bethe-Salpeter equation

HS, D. Geffroy, M. Wallerberger, J. Otsuki, K. Yoshimi, E. Gull, J. Kuneš, SciPost Phys. 8, 012 (2020)

Interpolating the local vertex, one can solve Bethe-Salpeter equation using a dense large Matsubara mesh.



Goldstone mode in the condensed state

Remaining issue: How to solve diagrammatic equations at the twoparticle level in IR?

M. Wallerberger*, <u>HS</u>*, A. Kauch, arXiv:2012.05557

Future perspectives: Quantum embedding theories Embedding in space Embedding in energy



Dynamical mean-field theory, dynamical vertex approximation *etc*.

Constrained RPA etc.

Dynamical susceptibility calculations in DMFT, non-local diagrammatic calculations of quantum critical and unconventional superconductivity...

Summary

Intermediate representation +

Sparse sampling +

Tensor networks

Compact representation of Green's functions

M Diagrammatic equations at single-particle level

Sparse QMC measurement

Computation at two-particle level (ongoing)

 \Box Unconventional superconductivity *etc*.

One-particle theory

- HS, J. Otsuki, M. Ohzeki, K. Yoshimi, PRB 96, 035147 (2017)
- 「固体物理」に解説記事を執筆中! • J. Otsuki, M. Ohzeki, <u>HS</u>, K. Yoshimi, PRE 95, 061302(R) (2017)
- N. Chikano, J. Otsuki, <u>HS</u>, PRB **98**, 035104 (2018)
- N. Chikano, K. Yoshimi, J. Otsuki, <u>HS</u>, Compt. Phys. Commun. 240, 181 (2019)
- J. Li, M. Wallerberger, C.-N. Yeh, N. Chikano, E. Gull, <u>HS</u>, PRB **101**, 035144 (2020)
- T. Wang, T. Nomoto, Y. Nomura, <u>HS</u>, J. Otsuki, T. Koretsune, and R. Arita, PRB **102**, 134503 (2020)

Two-particle theory

- HS, J. Otsuki, M. Ohzeki, K. Yoshimi, K. Haule, M. Wallerberger, E. Gull, PRB 97, 205111 (2018)
- HS, D. Geffroy, M. Wallerberger, J. Otsuki, K. Yoshimi, E. Gull, J. Kuneš, SciPost Phys. 8, 012 (2020)
- M. Wallerberger*, <u>HS</u>*, A. Kauch, arXiv:2012.05557