



Görlitz – Jülich collaboration on MALA – JuDFT

February 11, 2025 | **Johannes Wasmer**¹ Lenz Fiedler² Philipp Rüßmann¹ Attila Cangi² Stefan Blügel¹ | ¹FZ Jülich PGI-1 ²HZDR CASUS MLMD

Talk held at HZDR-CASUS **MLMD** group meeting for ongoing collaboration project via HIDA Trainee Network.

Latest version of slides are [here](#).

My research organization

Helmholtz Association



- Largest of the four big non-university research organizations in Germany, 45k employees
- Basic & applied research
- Large facilities, accelerators, supercomputing, etc.

helmholtz.de/en

Julich Research Centre



- One of the largest in Europe, 7.5k employees
- Helmholtz Research Fields: ► **Information**, Energy, Earth & Environment, Matter.
- Site of first exascale computer in Europe (2025)

fzj.de/en

My funding

This research visit

HIDA Trainee Network



My PhD

HIDA Grad School HDSLEE &
EU Joint Virtual Lab



AI Data Analytics and Scalable Simulations



Join via
cost-daemon.eu



European network for data-driven
materials science

- ▶ **WG1:** Community standards: data, workflows and codes for materials design.
- ▶ **WG2:** Representations and algorithms for materials design for “single-modality” use.
- WG3:** Multi-modal machine learning methods for advanced materials design.
- WG4:** Process-structure-property relationships in materials. Novel insights and applications.
- WG5:** Training, Dissemination, Exploitation, Outreach

My research group

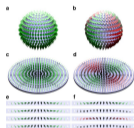
Division “Quantum Theory of Materials”



Stefan Blügel



PhD advisor



Expertise DFT, Quantum magnetism, Spintronics

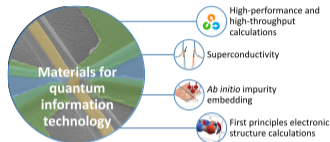
► Group “Materials for Quantum Information Technology”



Philipp Rüßmann



Co-advisor



Expertise DFT, Superconductivity, Topological materials

Code development



judft.de

//

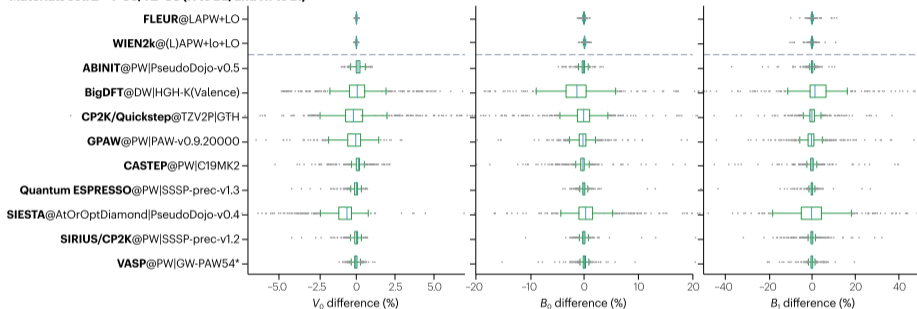


JuDFTteam

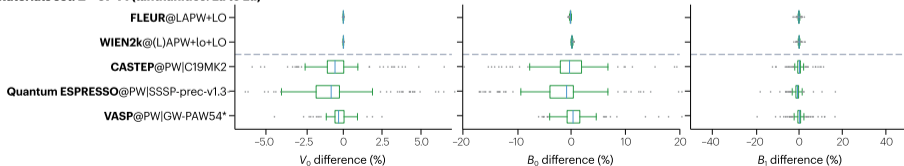
JuDFT codes accuracy

FLEUR and JuKKR are all-electron, full-potential open-source codes.

Materials set: Z = 1–56, 72–83 (H to Ba, and Hf to Bi)



Materials set: Z = 57–71 (lanthanides: La to Lu)



¹ Bosoni et al. 2024.



Automated Interactive Infrastructure and Database for Computational Science



⚙️ Workflows

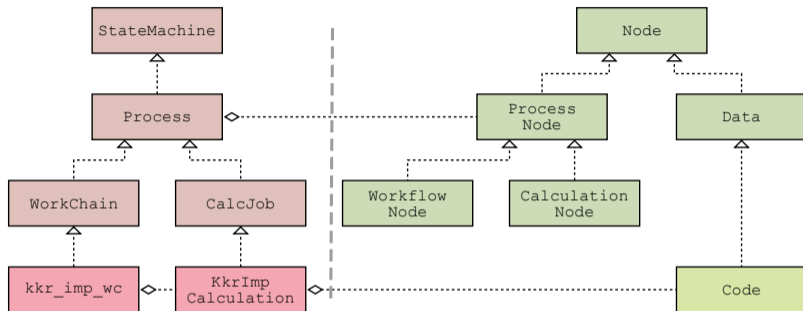
🔗 Provenance

🔌 Plugin Framework

🖥️ HPC Interface

📖 Open Science

🔄 Open source



⚙️ Engine

🗄️ Database

The Korringa-Kohn-Rostoker Green Function (KKR) method¹

¹Slides by Prof. Dr. Phivos Mavropoulos, National and Kapodistrian University of Athens, from his guest lecture, given January 19, 2021, in the graduate course “Density Functional Theory and Electronic Structure”, held by Prof. Dr. Stefan Blügel, at RWTH Aachen University, winter semester 2020 / 2021.

Green function: Definition

Time evolution operator

$$\psi(t) = e^{-iH(t-t')}\psi(t') \iff$$

Retarded and advanced Green function

$$\psi(t) = iG^R(t-t')\psi(t') \quad t' < t$$

$$\psi(t) = -iG^A(t-t')\psi(t') \quad t < t'$$

Fourier transform \rightarrow energy-dependent Green function:

$$G(E) = \int_{-\infty}^{\infty} G(t) e^{i(E+i\varepsilon)t} dt = (E + i\varepsilon - H)^{-1}$$

Spectral representation via eigenfunctions-eigenvalues:

$$G(E) = \sum_i \frac{|\psi_i\rangle\langle\psi_i|}{E - \epsilon_i}$$

$G(E)$ can be defined for complex E
(analytical continuation)

Properties of the Green function I:

Analytical properties

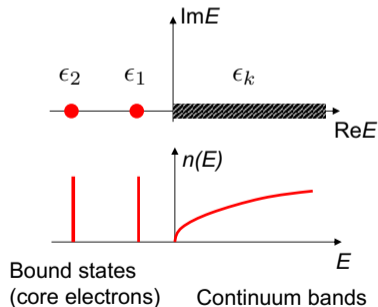
Analyticity in the physical sheet $\text{Im}\sqrt{E} > 0$

Poles at the eigenvalues ϵ_i and branch cut at the continuum ϵ_k

Find eigenvalues of $H \Leftrightarrow$
Locate singularities of $G(E)$

Density of states:

$$n(E) = -\frac{1}{\pi} \text{Im Tr } G(E)$$



Properties of the Green function II:

Connection to physical properties

Definition: $(E - H) G(E) = 1 \implies$

Real-space representation: $(E - H) G(\vec{r}, \vec{r}'; E) = \delta(\vec{r} - \vec{r}')$

Expectation values of observables:

$$\begin{aligned}\langle \mathcal{O} \rangle &= -\frac{1}{\pi} \text{Im Tr} \int^{E_F} dE \mathcal{O} G(E) && \text{(General expression with Tr)} \\ &= -\frac{1}{\pi} \text{Im} \int^{E_F} dE \int d^3r d^3r' \mathcal{O}(\vec{r}, \vec{r}') G(\vec{r}', \vec{r}; E) && \text{(Real-space representation)}\end{aligned}$$

Charge density: $\rho(\vec{r}) = -\frac{1}{\pi} \text{Im} \int^{E_F} dE G(\vec{r}, \vec{r}; E)$

Properties of the Green function III:

Dyson equation

Reference system Hamiltonian \rightarrow New system Hamiltonian

$$H_0 \longrightarrow H = H_0 + \Delta V$$

\downarrow \downarrow

$$G_0(E) = (E - H_0)^{-1} \longrightarrow \boxed{\begin{aligned} G(E) &= (E - H_0 - \Delta V)^{-1} \\ &= [1 - G_0(E) \Delta V]^{-1} G_0(E) \end{aligned}}$$

Dyson equation

Example:

Free electrons \rightarrow Electrons in a crystal

$$H_0 = -\nabla^2 \quad H_{\text{cryst}} = H_0 + V_{\text{cryst}}(\vec{r})$$

Electrons in a crystal \rightarrow Electrons in a crystal + impurity

$$H_0 = H_{\text{cryst}} \quad H = H_{\text{cryst}} + \Delta V_{\text{imp}}$$

Concept of KKR

Tessellation of space in atomic cells

Local scattering solution of Schrödinger's equation
in each cell (spherical wave representation)

Multiple scattering of waves among atoms
→ Coupling of local solutions

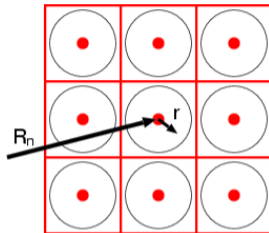
Starting point:

Free electron wave function

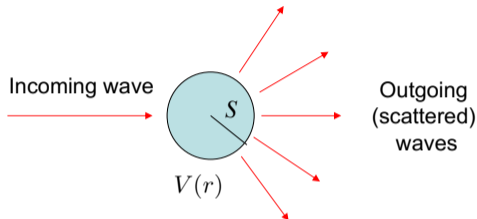
$$\psi_{\vec{k}}(\vec{r}) = e^{i\vec{k} \cdot \vec{r}} = 4\pi \sum_L i^l j_l(kr) Y_L(\hat{k}) Y_L(\hat{r}) \quad L \equiv (l, m)$$

Free electron Green function

$$g(\vec{r}, \vec{r}'; E) = -\frac{1}{4\pi} \frac{e^{ik|\vec{r}-\vec{r}'|}}{|\vec{r}-\vec{r}'|} = -i\sqrt{E} \sum_L j_l(r_{<}; E) h_l(r_{>}; E) Y_L(\vec{r}) Y_L(\vec{r}')$$



Scattering from a single potential



Green
function:

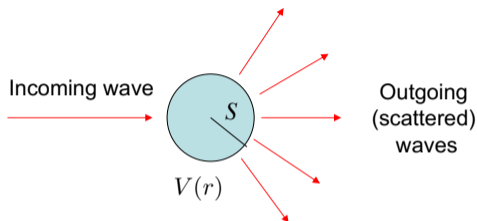
$$G(\vec{r}, \vec{r}'; E) = -i\sqrt{E} \sum_L \underbrace{R_l(r_{<}; E)}_{\substack{\text{Regular} \\ \text{solution} \\ \sim r^l}} \underbrace{H_l(r_{>}; E)}_{\substack{\text{Irregular} \\ \text{solution} \\ \sim 1/r^{l+1}}} Y_L(\vec{r}) Y_L(\vec{r}')$$

$L := (l, m)$

Free-space
Green
function:

$$g(\vec{r}, \vec{r}'; E) = -i\sqrt{E} \sum_L \underbrace{j_l(r_{<}; E)}_{\substack{\text{Bessel} \\ \text{function}}} \underbrace{h_l(r_{>}; E)}_{\substack{\text{Hankel} \\ \text{function}}} Y_L(\vec{r}) Y_L(\vec{r}')$$

Scattering from a single potential



Green function:

$$G(\vec{r}, \vec{r}'; E) = -i\sqrt{E} \sum_L \underbrace{R_l(r_{<}; E)}_{\text{Regular solution}} \underbrace{H_l(r_{>}; E)}_{\text{Irregular solution}} Y_L(\vec{r}) Y_L(\vec{r}')$$

$$t_l(E) = \int_{\text{cell}} d^3r j_l(\sqrt{E}r) V(r) R_l(r; E)$$

$L := (l, m)$

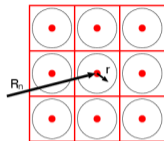
Free-space
Green
function:

$$g(\vec{r}, \vec{r}'; E) = -i\sqrt{E} \sum_L \underbrace{j_l(r_{<}; E)}_{\text{Bessel function}} \underbrace{h_l(r_{>}; E)}_{\text{Hankel function}} Y_L(\vec{r}) Y_L(\vec{r}')$$

KKR representation of GF

Crystal electron Green function:

$$(-\nabla^2 + V^n(\vec{r}) - E) G(\vec{r} + \vec{R}^n, \vec{r}' + \vec{R}^{n'}; E) = -\delta_{nn'} \delta(\vec{r} - \vec{r}')$$



Expanded as:

$$G(\vec{r} + \vec{R}^n, \vec{r}' + \vec{R}^{n'}; E) =$$

$$-i\sqrt{E} \sum_L R_L^n(\vec{r}_{<}; E) H_L^n(\vec{r}_{>}; E) \delta_{nn'} + \sum_{LL'} R_L^n(\vec{r}; E) \overset{\text{Structural Green functions}}{G_{LL'}^{nn'}(E)} R_L^{n'}(\vec{r}'; E)$$

“Single-site”
(atom in constant potential)

“Back-scattering”
(contribution of all other atoms)

Free electrons:

$$g = -i\sqrt{E} \sum_L j_L^n(\vec{r}_{<}; E) h_L^n(\vec{r}_{>}; E) \delta_{nn'} + \sum_{LL'} j_L^n(\vec{r}; E) g_{LL'}^{nn'}(E) j_L^{n'}(\vec{r}'; E)$$

Algebraic Dyson equation

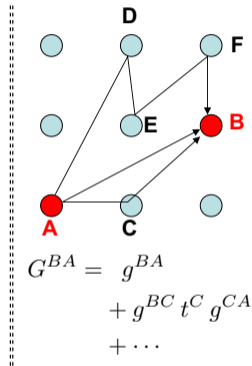
Dyson eq.: Green functions

$$G(E) = g(E) + g(E) V_{\text{cryst}} G(E)$$



Algebraic Dyson eq.: structural Green functions

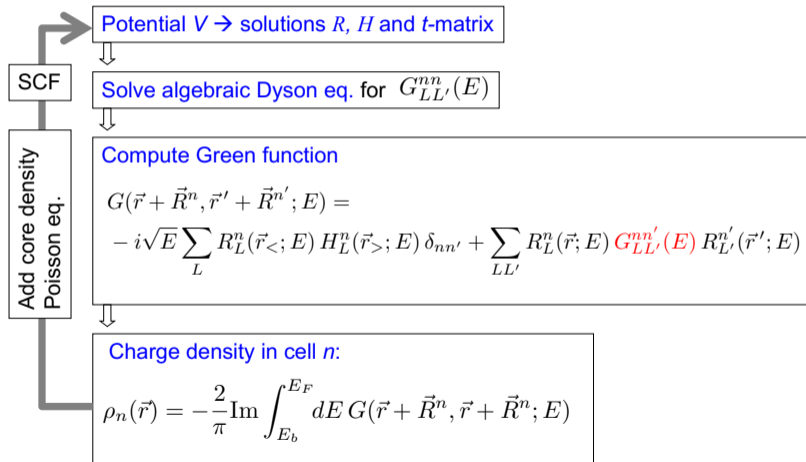
$$G_{LL'}^{nn'}(E) = g_{LL'}^{nn'}(E) + \sum_{n'', L''} g_{LL''}^{nn''}(E) t_{l''}^{n''}(E) G_{L''L'}^{n''n'}(E)$$



Interpretation: propagation over all scattering paths

$$G_{LL'}^{nn'} = g_{LL'}^{nn'} + \sum_{n'', L''} g_{LL''}^{nn''} t_{l''}^{n''} g_{L''L'}^{n''n'} + \sum_{n'', L''} \sum_{n''', L'''} g_{LL''}^{nn''} t_{l''}^{n''} g_{L''L'''}^{n''n'''} t_{l'''}^{n'''} g_{L'''L'}^{n'''n'} + \dots$$

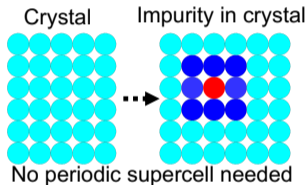
Summary of algorithm for valence-electron charge density



Impurity atoms in crystals

Dyson eq.: Reference system \rightarrow New system

$$\begin{aligned} G(E) &= (E - H_0 - \Delta V)^{-1} \\ &= [1 - G_0(E) \Delta V]^{-1} G_0(E) \end{aligned}$$



Computational effort depends on the number of *perturbed atomic potentials* N : $O(N^3)$

$$\begin{aligned} G &= G_0 + G_0 \Delta V G \\ &= G_0 + G_0 \Delta V G_0 + G_0 \Delta V G_0 \Delta V G_0 + G_0 \Delta V G_0 \Delta V G_0 \Delta V G_0 + \dots \\ &= G_0 + G_0 [\Delta V + \Delta V G \Delta V] G_0 \\ &:= G_0 + G_0 T G_0 \end{aligned}$$

T is confined in the perturbed region

Impurity atoms in crystals

Green function and Dyson equation for the impurity system

$$G(\vec{r} + \vec{R}^n, \vec{r}' + \vec{R}^{n'}; E) = \\ -i\sqrt{E} \sum_L R_L^n(\vec{r}_{<}; E) H_L^n(\vec{r}_{>}; E) \delta_{nn'} + \sum_{LL'} R_L^n(\vec{r}; E) \mathbf{G}_{LL'}^{nn'}(E) R_{L'}^{n'}(\vec{r}'; E)$$

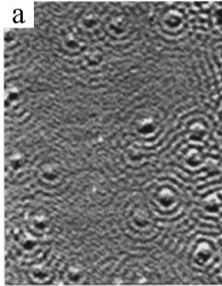
Summation only for the sites where t-matrix differs (impurities & surrounding atoms)

$$G_{LL'}^{nn'}(E) = G_{0;LL'}^{nn'}(E) + \sum_{n'', L''} G_{0;LL''}^{nn''}(E) [t_{l''}^{n''}(E) - t_{0;l''}^{n''}(E)] G_{L''L'}^{n''n'}(E)$$

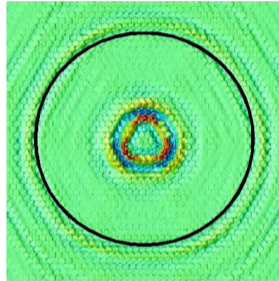
Size of linear system to solve: $\text{DIMENSION} = N_{\text{at}} \times (2l + 1)^2$

$$G_{LL'}^{nn'}(E) = [\mathbf{1} - \mathbf{G}_0(\mathbf{t} - \mathbf{t}_0)]^{-1} \mathbf{G}_0]_{LL'}^{nn'}$$

Application: Density oscillations



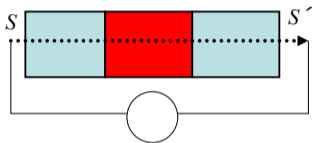
Exp.: Cu (111) surface
Petersen et al., PRB **57**,
R6858 (1998)



KKR calculation (Co in Cu surface)
Samir Lounis, PhD thesis (RWTH Aachen)

Other applications

Conductance in a junction:



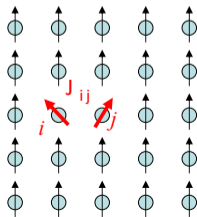
$$J = \Gamma \Delta U$$

$$\Gamma = -\frac{e^2 \hbar^3}{8\pi m^2} \int_S dS \int_{S'} dS' G(\mathbf{r}, \mathbf{r}'; E_F) \nabla_z \nabla'_z G^*(\mathbf{r}, \mathbf{r}'; E_F)$$

Exchange constants in ferromagnets

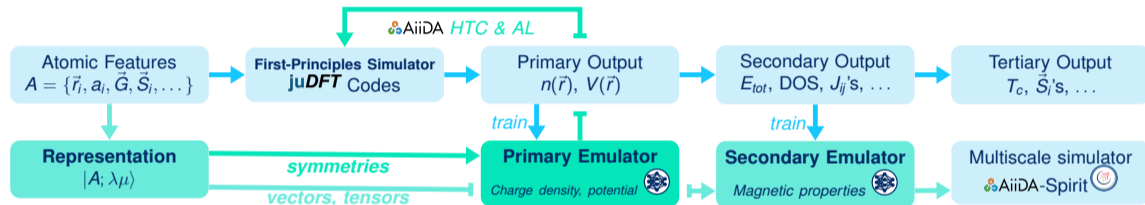
Heisenberg model:
$$E = - \sum_{i,j} J_{ij} \hat{e}_i \cdot \hat{e}_j$$

$$J_{ij} = -\frac{1}{4\pi} \text{Im} \int^{E_F} dE \text{Tr}_L \Delta t_{(\uparrow-\downarrow)}^i G_{\uparrow}^{ij}(E_F) \Delta t_{(\uparrow-\downarrow)}^j G_{\downarrow}^{ji}(E_F)$$



Vision: Electronic structure learning

as integrated, high-level multiscale workflows

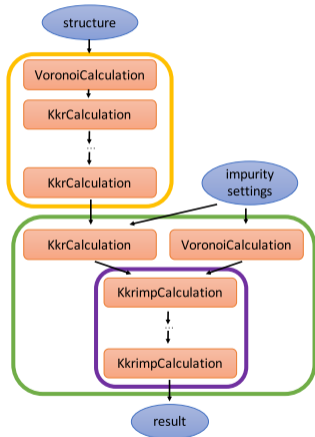


Better “initial guess”
for fast SCF convergence

Magnetic property prediction
(ML-Exc) for spin dynamics

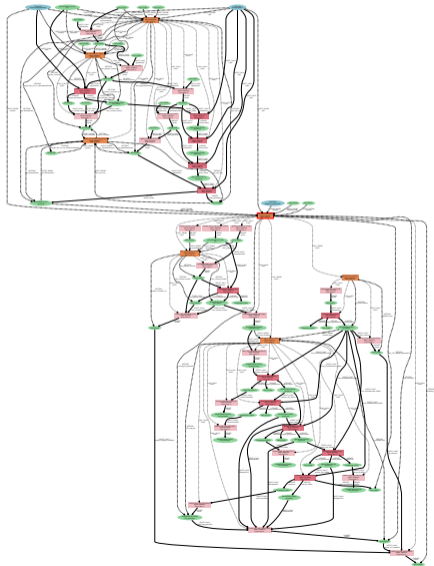
Project “Better initial guess”

AiiDA-KKR workflows



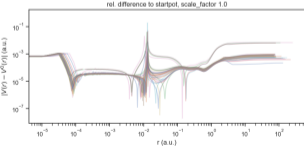
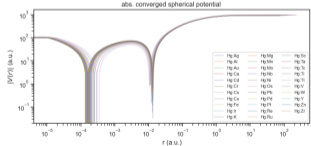
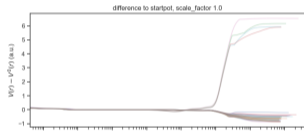
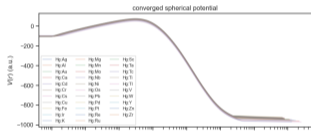
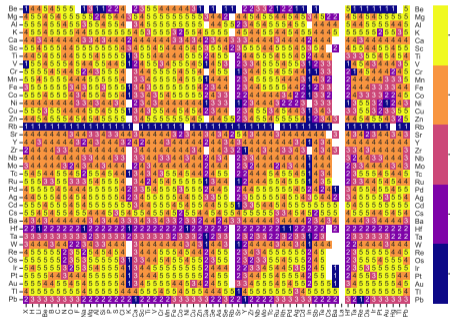
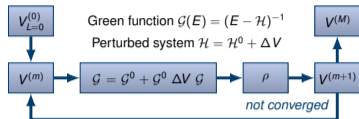
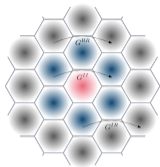
a

^aRüßmann, Bertoldo, and Blügel 2021.



Data generation

Target Electron potential difference $\Delta V_{imp}(\vec{r})$



Spherical impurity potentials in first Voronoi cell of Hg:X embeddings,
left upper to bottom right: V , $V - V^0$, $|V|$, $|V - V^0|$.

The Jacobi-Legendre framework¹

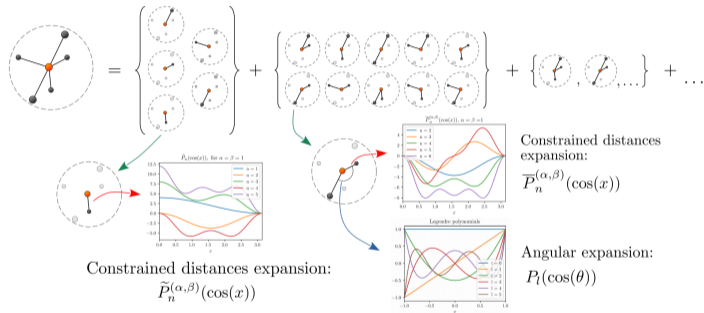
for electronic structure representation

Body order:

2B

3B

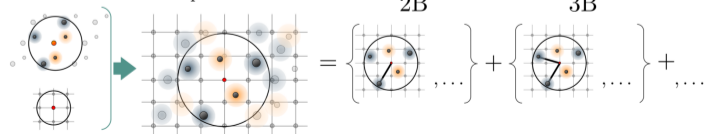
4B



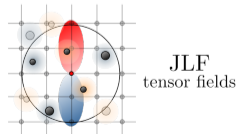
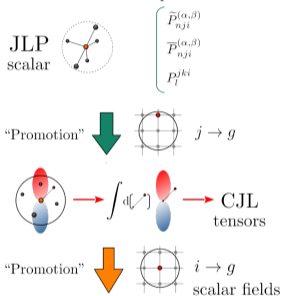
Grid centered representation

2B

3B



Expansion basis



¹Michelangelo Domina 2024.

Atom-based JLCDM

- The full KKR potential is expanded in real-space Voronoi (Wigner-Seitz) cells around each atom and convoluted with shape functions

$$V(\vec{r}) = \sum_L V_L(r) Y_L(\vec{r}) \quad \text{with} \quad V_{LL'}(r) = \sum_{L''} C_{LL'L''} V_{L''}(r)$$

$$V^n(\vec{r}) = V(\vec{r} + \vec{R}^n) \Theta(\vec{r}) \quad \text{with} \quad \Theta(\vec{r}) = \sum_L \Theta_L^n(r) Y_L(\vec{r})$$

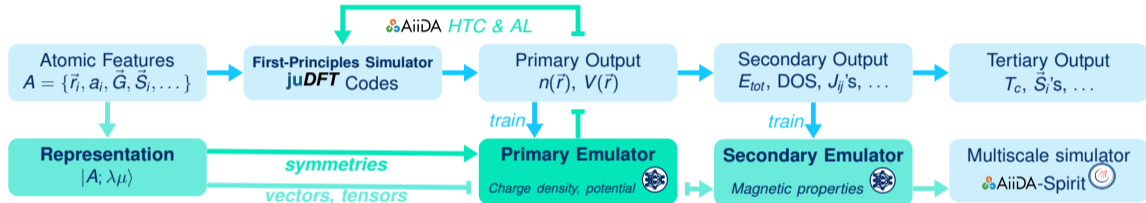
- Idea: Expand the Jacobi-Legendre grid description at each atom, instead of each grid point.

$$V(\vec{r}) = \sum_i \sum_n a_n \tilde{P}_n^{ig} + \sum_{ij} \sum_{n_1, n_2, L} a_{n_1, n_2, i} \bar{P}_{n_1}^{ig} \bar{P}_{n_2}^{ij} P_L^{ijg}$$

Angular expansion $\int P_L(\hat{r}_{gi} \cdot \hat{r}_{gj}) Y(\hat{r}_{gi}) d\hat{r} \longrightarrow \int P_L(\hat{r}_{ij} \cdot \hat{r}_{ig}) Y(\hat{r}_{ig}) d\hat{r}$

Vision: Electronic structure learning

as integrated, high-level multiscale workflows

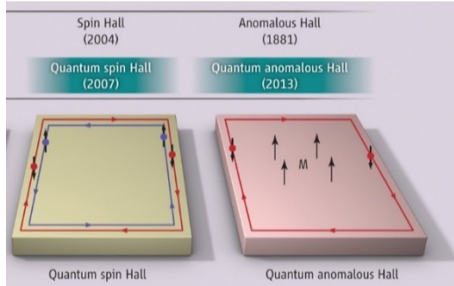


Better “initial guess”
for fast SCF convergence

Magnetic property prediction
(ML-Exc) for spin dynamics

Topological insulators and magnetic impurities

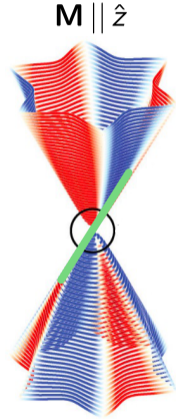
- Magnetic doping of topological insulators (**TIs**) can induce a topological phase transition
 - Ferromagnetic ordering
 - Out-of-plane anisotropy



(QSHE)
→ Topological insulator
*Two counter propagating
edge states*

(QAHE)
One single edge state

S. Oh, Science 340, 153 (2013)



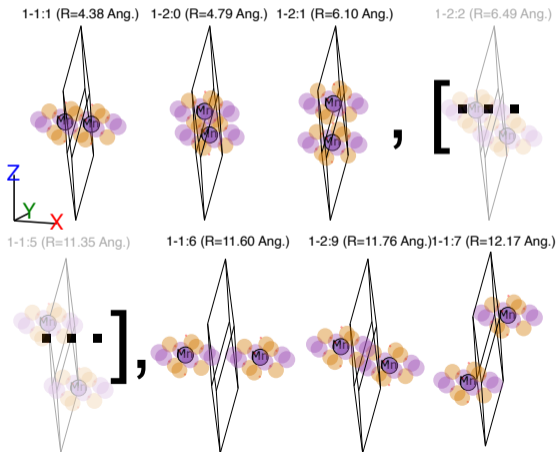
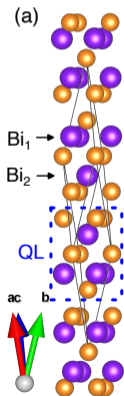
Henk et al., PRL 109, 076801 (2012)

Project “ML-Exc”

Magnetic co-doping of topological insulators

Bi_2Te_3

Dimer clusters of 3d, 4d transition metal defects



Single-impurity database, N=2'000.

go.fzj.de/judit

Dimer database, N=2'000^a.

Co-doping can help to control

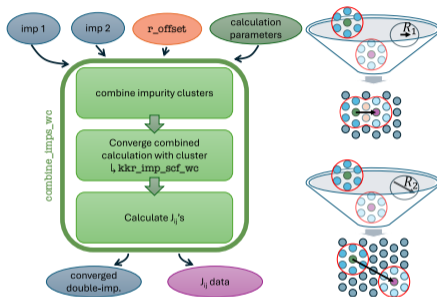
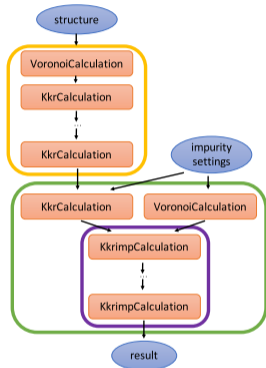
- critical T_c of QAHE
- exchange splitting Δ_{xc}
- long-range magnetic ordering

for applications in spintronics and fault-tolerant quantum computing.

^aMozumder et al. 2024.

Project “ML-Exc”

AiiDA-KKR workflows²

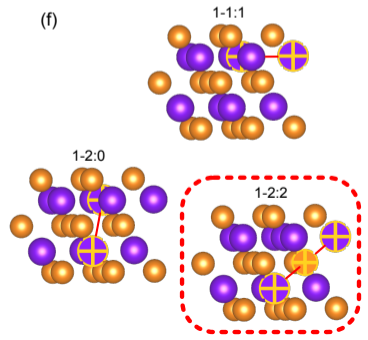
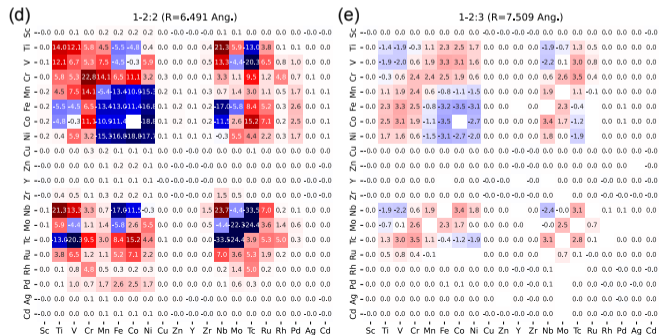
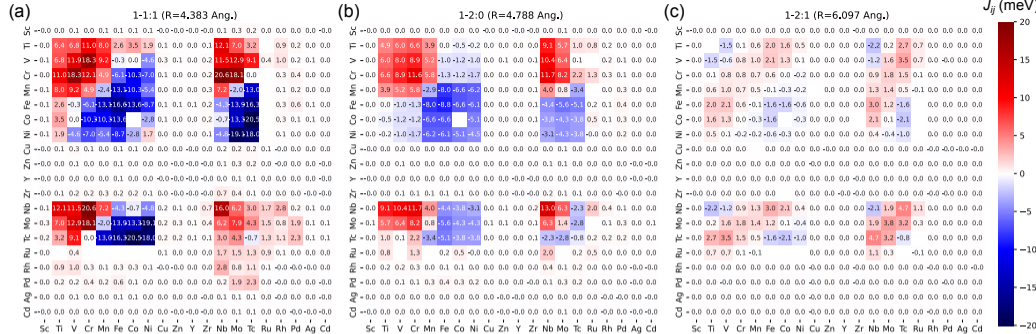


Extended Heisenberg Hamiltonian. $H = -\frac{1}{2} \sum_{i,j} J_{ij} \vec{S}_i \cdot \vec{S}_j - \frac{1}{2} \sum_{i,j} \vec{D}_{ij} \cdot (\vec{S}_i \times \vec{S}_j)$

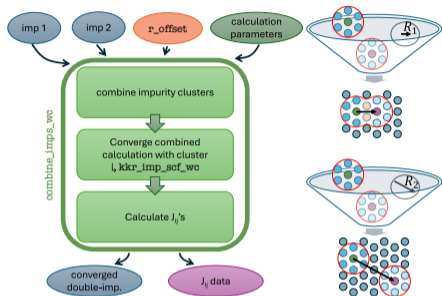
Exchange constants from method of infinitesimal rotations¹. $\mathcal{J}_{ij} = -\frac{1}{\pi} \text{Im} \int_{-\infty}^{E_F} dE \text{Tr}[\delta t_i G_{ij} \delta t_j G_{ji}]$

¹ Liechtenstein et al. 1987.

² Rüßmann, Bertoldo, and Blügel 2021.



Spin dynamics with AiiDA-spirit¹

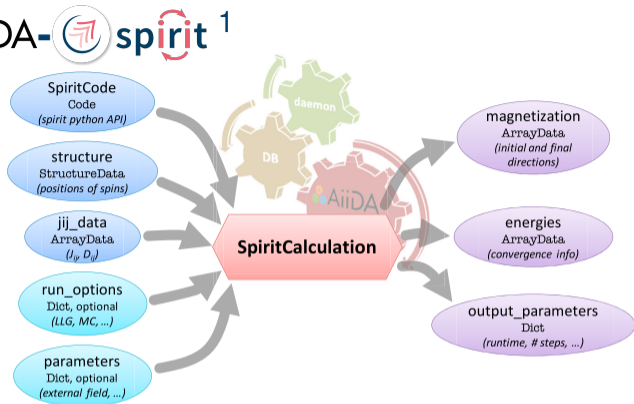


Liechtenstein formula

$$\mathcal{J}_{ij} = -\frac{1}{\pi} \text{Im} \int_{-\infty}^{E_F} dE \text{Tr}[\delta t_i G_{ij} \delta t_j G_{ji}]$$

ML-IAP approach.

$$E_k = \sum_k E_k \longrightarrow J_{ij} = \sum_k (J_{ij})_k$$



Landau-Lifshitz-Gilbert equation

$$\frac{\partial \vec{S}_i}{\partial t} = -\gamma' \vec{S}_i \times \vec{B}_i^{\text{eff}} - \lambda \vec{S}_i \times (\vec{S}_i \times \vec{B}_i^{\text{eff}})$$

juspin.de

¹ Rüßmann, Ribas Sobreviela, et al. 2022.

Project “ML-Exc”

Model selection

Model	Citation	Type	CSSP	NN	Defects	Spin	Symmetry
AutoML	Conrad et al. 2022	Composition	X			-	-
Coulomb matrix	Rupp et al. 2012	Descriptor				-	explicit
SOAP	Bartók, Kondor, and Csányi 2013	Descriptor				-	explicit
Alchemical SOAP	Lopaničsyna et al. 2023	Potential	X			-	explicit
Disordered SOAP	Sommer et al. 2023	Descriptor			X	-	explicit
GAP	Bartók, Payne, et al. 2010	Potential	X			?	explicit
ACE	Bochkarev et al. 2022	Potential				-	explicit
MEGNetSparse	Kazeev et al. 2023	Potential	X	X	X	-	explicit
MACE	Batatia, Kovács, et al. 2022	Potential	X	X		-	explicit
MACE-MP	Batatia, Benner, et al. 2023	UIP	X	X		-	explicit
Magnetic ACE	Rinaldi et al. 2024	Potential				noco	explicit
SpinGNN++	Yu et al. 2024	Potential	X	X		noco	explicit
CHGNet	Deng et al. 2023	UIP	X	X		coll	explicit
SNRep	M. Domina, Cobelli, and Sanvito 2022	Potential				coll	explicit
PET	Pozdnyakov and Ceriotti 2024	Potential	?	X		coll	approximate

Project “ML-Exc”

Tensorial interaction

Heisenberg Hamiltonian in tensor form.

$$\mathcal{H}_H = - \sum_{j>i} \vec{m}_i \cdot \mathcal{J}_{ij} \vec{m}_j$$

Tensor components: isotropic, anti-symmetric (DMI) and anisotropic or traceless symmetric part (neglected so far).

$$\mathcal{J}_{ij} = J_{ij} \mathbb{1} + \mathcal{J}_{ij}^A + \mathcal{J}_{ij}^S$$

with $J_{ij}^{xx} = J_{ij}^{yy} = J_{ij}^{zz} = \frac{1}{3} J_{ij}$ and

$$\mathcal{J}_{ij}^A = \begin{bmatrix} 0 & J_{ij}^{xy} & J_{ij}^{xz} \\ J_{ij}^{yx} & 0 & J_{ij}^{yz} \\ J_{ij}^{zx} & J_{ij}^{zy} & 0 \end{bmatrix} = \begin{bmatrix} 0 & J_{ij}^{xy} & -J_{ij}^{zx} \\ -J_{ij}^{xy} & 0 & J_{ij}^{yz} \\ J_{ij}^{zx} & -J_{ij}^{yz} & 0 \end{bmatrix} = \begin{bmatrix} 0 & -D_{ij}^z & D_{ij}^y \\ D_{ij}^z & 0 & -D_{ij}^x \\ -D_{ij}^y & D_{ij}^x & 0 \end{bmatrix}$$

HIGHER-ORDER SPIN INTERACTION

✓ $H_1 = \sum'_{ij} J_{ij} \hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_j$ (The bilinear term)

W. Heisenberg, Z. Physik **49**, 619 (1928); P. W. Anderson, Phys. Rev. **79**, 350 (1950)

✓ $H_2 = \sum'_{ij} B_{ij} (\hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_j)^2$ (The biquadratic term)

P. W. Anderson, in *Magnetism* Vol. 1 (Academic Press, 1963); C. Kittel, Phys. Rev. **120**, 335 (1960).

✓ $H_3 = \sum'_{ijkl} K_{ijkl} \left[(\hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_j) (\hat{\mathbf{S}}_k \cdot \hat{\mathbf{S}}_l) + (\hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_l) (\hat{\mathbf{S}}_j \cdot \hat{\mathbf{S}}_k) - (\hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_k) (\hat{\mathbf{S}}_j \cdot \hat{\mathbf{S}}_l) \right]$ (The ring exchange interaction term)

D. J. Thouless, Proc. Phys. Soc. **86**, 893 (1965)

M. Takahashi, J. Phys. C: Solid State Phys. **10**, 1298 (1977).

✓ $H_4 = \sum'_{ijk} Y_{ijk} (\hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_j) (\hat{\mathbf{S}}_j \cdot \hat{\mathbf{S}}_k)$ (The 4-spin—3-site interaction term)

M. Hoffmann *et al.*, Phys. Rev. B **101**, 024418 (2020).

✓ $H^{\text{CC}} = -\frac{1}{2} \left[\hat{\mathbf{S}}_i \cdot (\hat{\mathbf{S}}_j \times \hat{\mathbf{S}}_k) \right] \tau_{ijk}^\dagger \kappa_{ii'}^{\text{CC}} \tau_{i'j'k'} \left[\hat{\mathbf{S}}_{i'} \cdot (\hat{\mathbf{S}}_{j'} \times \hat{\mathbf{S}}_{k'}) \right]$ (The chiral-chiral interaction term)

S. Grytsiuk *et al.*, Nat. Commun. **11**, 511 (2020).

✓ $\sum'_{i,j,k} (\hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_j) (\hat{\mathbf{S}}_j \cdot \hat{\mathbf{S}}_k) (\hat{\mathbf{S}}_k \cdot \hat{\mathbf{S}}_i)$ (6-spin—3-site interaction term)

RELATIVISTIC EXCHANGE INTERACTION

$$\hat{\mathcal{H}}_{\text{soc}} = -\lambda \sum_i \sum_m \sum_{\mu} \frac{\langle g_i(\mathbf{r}) | \hat{L}_i^{\mu} | m_i(\mathbf{r}) \rangle}{E_{i,m} - E_{i,g}} \left[\hat{S}_i^{\mu}, \hat{\mathcal{H}}^{(s)} \right]$$

Spin Hamiltonian

$$J_{12} \hat{\mathbf{S}}_1 \cdot \hat{\mathbf{S}}_2 \longrightarrow D_{12}^{\text{bl}} \cdot (\hat{\mathbf{S}}_1 \times \hat{\mathbf{S}}_2)$$

$$B_{12} (\hat{\mathbf{S}}_1 \cdot \hat{\mathbf{S}}_2)^2 \longrightarrow D_{12}^{\text{bq}} \cdot \left[(\hat{\mathbf{S}}_1 \cdot \hat{\mathbf{S}}_2) (\hat{\mathbf{S}}_1 \times \hat{\mathbf{S}}_2) + (\hat{\mathbf{S}}_1 \times \hat{\mathbf{S}}_2) (\hat{\mathbf{S}}_1 \cdot \hat{\mathbf{S}}_2) \right]$$

$$\begin{aligned} K_{1234} \left[(\hat{\mathbf{S}}_1 \cdot \hat{\mathbf{S}}_2) (\hat{\mathbf{S}}_3 \cdot \hat{\mathbf{S}}_4) \right. \\ \left. + (\hat{\mathbf{S}}_1 \cdot \hat{\mathbf{S}}_4) (\hat{\mathbf{S}}_2 \cdot \hat{\mathbf{S}}_3) - (\hat{\mathbf{S}}_1 \cdot \hat{\mathbf{S}}_3) (\hat{\mathbf{S}}_2 \cdot \hat{\mathbf{S}}_4) \right] \longrightarrow \\ D_{12}^{\text{ring}} \cdot (\hat{\mathbf{S}}_3 \cdot \hat{\mathbf{S}}_4) (\hat{\mathbf{S}}_1 \times \hat{\mathbf{S}}_2) + D_{34}^{\text{ring}} \cdot (\hat{\mathbf{S}}_1 \cdot \hat{\mathbf{S}}_2) (\hat{\mathbf{S}}_3 \times \hat{\mathbf{S}}_4) \\ + D_{14}^{\text{ring}} \cdot (\hat{\mathbf{S}}_2 \cdot \hat{\mathbf{S}}_3) (\hat{\mathbf{S}}_1 \times \hat{\mathbf{S}}_4) + D_{23}^{\text{ring}} \cdot (\hat{\mathbf{S}}_1 \cdot \hat{\mathbf{S}}_4) (\hat{\mathbf{S}}_2 \times \hat{\mathbf{S}}_3) \\ - \left[D_{13}^{\text{ring}} \cdot (\hat{\mathbf{S}}_2 \cdot \hat{\mathbf{S}}_4) (\hat{\mathbf{S}}_1 \times \hat{\mathbf{S}}_3) + D_{24}^{\text{ring}} \cdot (\hat{\mathbf{S}}_1 \cdot \hat{\mathbf{S}}_3) (\hat{\mathbf{S}}_2 \times \hat{\mathbf{S}}_4) \right] \end{aligned}$$

$$Y_{123} \sum'_{i,j,k} (\hat{\mathbf{S}}_i \cdot \hat{\mathbf{S}}_j) (\hat{\mathbf{S}}_j \cdot \hat{\mathbf{S}}_k)$$

$$\longrightarrow \sum_{i=1}^3 D_{\Delta}^{4-3} \cdot \left[\hat{\mathbf{S}}_i \hat{\chi}_{123} + \hat{\chi}_{123} \hat{\mathbf{S}}_i - 2|\hat{\mathbf{S}}_1|^2 (\hat{\mathbf{S}}_2 \times \hat{\mathbf{S}}_3) - 2|\hat{\mathbf{S}}_2|^2 (\hat{\mathbf{S}}_3 \times \hat{\mathbf{S}}_1) - 2|\hat{\mathbf{S}}_3|^2 (\hat{\mathbf{S}}_1 \times \hat{\mathbf{S}}_2) \right]$$

Community resources

Best of atomistic machine learning



Largest list of atomistic ML tools on the web
(400+), auto-ranked, regular updates^a

go.fzj.de/baml

^aWasmer et al. 2023.

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Discussion slides

Solution in k -space for periodic systems

$$G_{LL'}^{nn'}(E) = g_{LL'}^{nn'}(E) + \sum_{n'', L''} g_{LL''}^{nn''}(E) t_{l''}^{n''}(E) G_{L''L'}^{n''n'}(E)$$



Fourier transformation of G, g

$$G_{LL'}(\vec{k}; E) = g_{LL'}(\vec{k}; E) + \sum_{L''} g_{LL''}(\vec{k}; E) t_{l''}(E) G_{L''L'}(\vec{k}; E)$$



Matrix inversion

$$G_{LL'}(\vec{k}; E) = \left[\left(1 - \mathbf{g}(\vec{k}; E) \mathbf{t}(E) \right)^{-1} \mathbf{g}(\vec{k}; E) \right]_{LL'}$$



On-site term for charge density

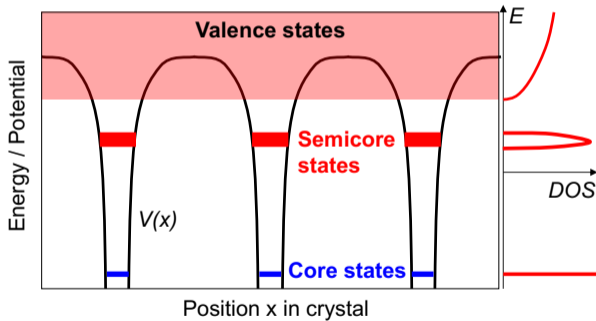
$$G_{LL'}^{nn}(E) = \frac{1}{V_{\text{BZ}}} \int_{\text{BZ}} d^3k G_{LL'}(\vec{k}; E)$$

$$[G] = [g] + [g] \cdot [t] \cdot [G]$$

Framework:

One-electron states in a crystal

Self-consistent field $V(x)$; density-functional theory



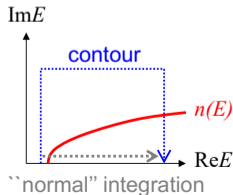
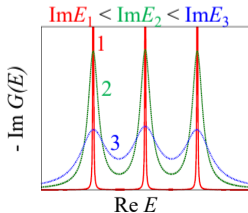
Complex energy contour

$$\rho_n(\vec{r}) = -\frac{2}{\pi} \text{Im} \int_{E_b}^{E_F} dE G(\vec{r} + \vec{R}^n, \vec{r} + \vec{R}^n; E)$$

Integration:
a lot of E -points

$$G(E) = \sum_{n\vec{k}} \frac{|\psi_{n\vec{k}}\rangle \langle \psi_{n\vec{k}}|}{\boxed{E - \epsilon_{n\vec{k}}}} \quad \text{Poles} \quad \rightarrow \quad n(E) = \sum_{n\vec{k}} \delta(E - \epsilon_{n\vec{k}})$$

$$G(E + i\varepsilon) = \sum_{n\vec{k}} \frac{|\psi_{n\vec{k}}\rangle \langle \psi_{n\vec{k}}|}{E + i\varepsilon - \epsilon_{n\vec{k}}} \quad \rightarrow \quad n(E + i\varepsilon) = \sum_{n\vec{k}} \frac{\varepsilon}{\varepsilon^2 + (E - \epsilon_{n\vec{k}})^2}$$



Lorentzian broadening
→ Less E -points for integration

$G(E)$ is analytical
→ Integral is the same.

Secular equation

Tessellation of space in atomic cells

Local solution of Schrödinger's equation in each cell
(spherical wave representation)

$$R_l(\vec{r}; E) Y_L(\hat{r})$$

In a sense: “optimal basis functions”

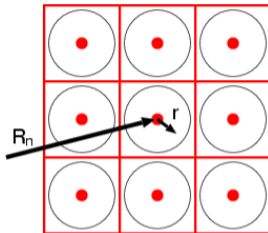
General solution:

$$\Psi_{\vec{k}}(\vec{r}) = \sum_L c_L(\vec{k}; E) R_l(\vec{r}; E) Y_L(\hat{r})$$

Insert into Schrödinger's eq. to obtain coefficients c_L

$$\sum_{L'} \left[\delta_{LL'} - t_l(E) g_{LL'}(\vec{k}; E) \right] c_{L'}(\vec{k}; E) = 0$$

↑
Fourier-transformed coefficients of free-space Green function



Multiple scattering

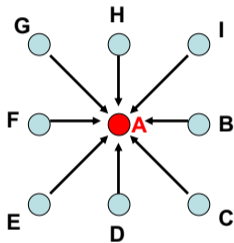
Bookkeeping:

$$1. \quad (\text{Incoming wave})_A = \sum_{n \in \{B, C, \dots\}} (\text{Outgoing waves})_n$$

$$2. \quad \psi_{\vec{k}}^{\text{sc}(n)}(\vec{r}) = \sum_L b_{\vec{k}L}^{\text{sc}(n)} h_L(\vec{r}; E)$$

$$\psi_{\vec{k}}^{\text{inc}(n')}(\vec{r}') = \sum_L b_{\vec{k}L}^{(n')} j_L(\vec{r}'; E)$$

related by
t-matrix



3. Transformation of Hankel functions:

$$h_L(\vec{r}' + \vec{R}^{n'} - \vec{R}^n; E) = \frac{i}{\sqrt{E}} \sum_{L'} g_{LL'}^{nn'}(E) j_{L'}(\vec{r}'; E)$$

4. Periodic system \rightarrow Bloch property $\rightarrow k$ -space

$$\sum_{L'} \left(\delta_{LL'} - g_{LL'}(\vec{k}; E) t_{L'}(E) \right) c_{\vec{k}L'}^{(n')} = 0 \quad \Psi_{\vec{k}}(\vec{r}) = \sum_L c_{\vec{k}L} R_L(\vec{r}; E)$$

$$\text{KKR secular equation} \rightarrow E = E(\vec{k})$$

Fermi surfaces, scattering off impurities

Lippmann-Schwinger equation

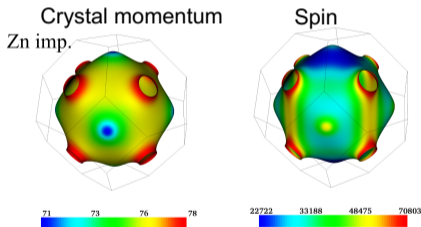
$$\Psi_{\vec{k}}^{\text{imp}}(\vec{r}) = \Psi_{\vec{k}}(\vec{r}) + \int d^3r' G_0(\vec{r}, \vec{r}'; E) [V^{\text{imp}}(\vec{r}') - V^{\text{cryst}}(\vec{r}')] \Psi_{\vec{k}}^{\text{imp}}(\vec{r}')$$

t -matrix (transition amplitude)

$$T_{\vec{k}\vec{k}'} = (\Psi_{\vec{k}}, V_{\text{imp}} \Psi_{\vec{k}'}^{\text{imp}})$$

Scattering rate

$$w_{\vec{k}\vec{k}'} = 2\pi |T_{\vec{k}\vec{k}'}|^2 \delta(E_{\vec{k}} - E_{\vec{k}'})$$



Finite lifetime (ps) of Fermi surf. states
in fcc Cu due to scattering off 1% Zn
impurities.

Summary

Green function method: Alternative to the Schrödinger equation

KKR Green function: Expansion in site-dependent scattering waves with single-site and multiple-scattering part

$$G(\vec{r} + \vec{R}^n, \vec{r}' + \vec{R}^{n'}; E) = \\ -i\sqrt{E} \sum_L R_L^n(\vec{r}_{<}; E) H_L^n(\vec{r}_{>}; E) \delta_{nn'} + \sum_{LL'} R_L^n(\vec{r}; E) G_{LL'}^{nn'}(E) R_{L'}^{n'}(\vec{r}'; E)$$





Impurity problem: Dyson equation

Only perturbed sites need recalculation




$$G(E) = (E - H_0 - \Delta V)^{-1} \\ = [1 - G_0(E) \Delta V]^{-1} G_0(E)$$

Large systems > 10000 atoms: Sparse matrices allow for O(N) calculations (KKR-nano method by R. Zeller et al.)




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