# CP2K Developers Meeting

March 8th, 2024 14:00-16:00 CET (see also <a href="https://www.cp2k.org/dev:meetings">https://www.cp2k.org/dev:meetings</a>)



# **CP2K Developers Meeting**

- 1. Current Development Efforts (all)
- 2. On-the-fly potentials (Martin Brehm)
- 3. GW and BSE for excitation energies (Max Graml)
- 4. Updates (Jan Wilhelm)
- 5. ...
- 6. CP2K-HFX FPGA Update (PC2)
- 7. CP2K@CASUS (Frederick Stein)
- 8. Current Issues when running CP2K (all)
- 9. Feature Deprecation (all)
- 10. CP2K Release (all)
- 11. Open CP2K-Related Positions (all)
- 12. CP2K-related Events (all)



# On-the-fly MLFF (Martin Brehm)

What are you currently working on or planning to work on?

- Jürg Hutter: there is overlap with a planned project
- Thomas Kühne: i-pi protocol might be an alternative integration option
- Ole: pytorch interface



### **Machine-Learning Force Fields**

- Have become very popular recently (e.g., Behler's NNPs)
- Towards the vision "ab initio accuracy at force field speed"

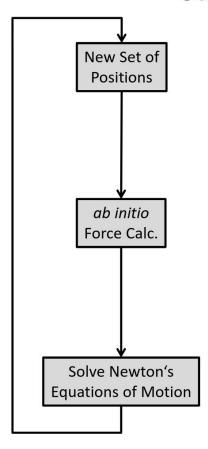
#### **However:**

- Training process is not straight-forward (coverage of full configuration space?) → Requires an expert
- Often, configurations outside of the high-confidence space are encountered during production run...
- → Stop production run, re-training, resume production run

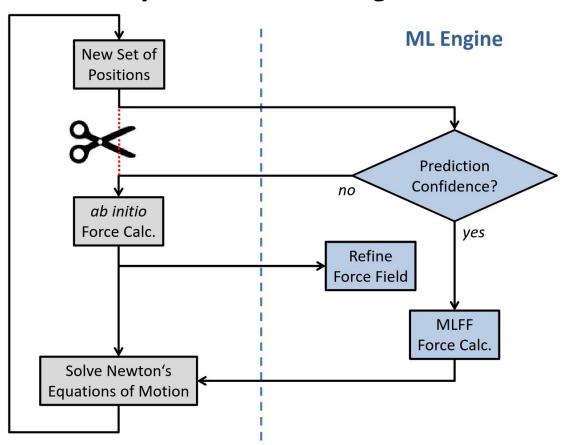
#### Conclusion: If ...

- ... you are a "standard AIMD user" without ML background,
- ... you want to run a single long trajectory,
- ... there is no trained MLFF available for your system,
- ..., today's MLFFs are not an option for you :-/

### **Standard AIMD**



### **On-the-Fly Machine-Learning Force Field**



### Implementation in VASP

**2019**: Ryosuke Jinnouchi, while Postdoc in Wien, implements such an approach in VASP:



"On-the-fly machine learning force field generation: Application to melting points", *Phys. Rev. B* **2019**, *100*, 014105, DOI 10.1103/PhysRevB.100.014105

- It is **not** based on neural networks uses Bayesian Inference ML
- It is a **true black box** (can be applied to any AIMD, no parameters need to be tuned, no expert needed)
- After a few hundred AIMD steps, already 90% of steps via MLFF.
   → can easily save a factor of > 10 in computer time
- This is **not** an empty marketing promise (a few colleagues of mine already use it for production)

### I find that's a really big thing.

I personally don't like VASP so much (commercial), see it as a part of my mission to convince scientists to use **free software**...

### **Introducing Prokyon**

I am planning to develop a C++ library "Prokyon" for on-the-fly MLFF applications (will be either GPL or L-GPL license).

- Will contain several models; the first one will be the Bayesian Inference ML as implemented in VASP
- Can be interfaced to any AIMD code in the future, but the prime target will be CP2k

Two planned modes of operation:

- a) CP2k drives the AIMD. Prokyon is invoked via MULTIPLE FORCE ENV as a second FORCE ENV
- **b)** Prokyon drives the AIMD. CP2k is invoked via libcp2k (as Gromacs QM/MM does)
- Interfacing Fortran ↔ C ↔ C++ works well (experience from Libvori)
- Technical details are currently being discussed (parallelization)
- Expect an Alpha version (hopefully) in summer...

# GW and BSE for excitation energies (Max Graml)

What are you currently working on or planning to work on?



### Bethe Salpeter equation for computing electronic excitations

#### **Bethe Salpeter equation**

$$\begin{pmatrix} A & B \\ B & A \end{pmatrix} \begin{pmatrix} \mathbf{X}^{(n)} \\ \mathbf{Y}^{(n)} \end{pmatrix} = \Omega^{(n)} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} \mathbf{X}^{(n)} \\ \mathbf{Y}^{(n)} \end{pmatrix}$$
$$A_{la,jb} = \left( \epsilon_a^{G_0 W_0} - \epsilon_t^{G_0 W_0} \right) \delta_{lj} \delta_{ab} + \alpha^{S/T} v_{la,jb} - W_{lj,ab} (\omega = 0)$$
$$B_{lajb} = \alpha^{S/T} v_{la,bj} - W_{lb,aj} (\omega = 0)$$

#### Hermitian equation

$$\begin{array}{c} C\, {\bf Z}^{(n)} = \Omega^{(n)^2}\, {\bf Z}^{(n)} \\ C = ({\bf A} - {\bf B})^{0.5}\, ({\bf A} + {\bf B})({\bf A} - {\bf B})^{0.5} \end{array}$$

#### Tamm Dancoff approximation

$$\mathbf{A}\,\mathbf{X}_{\mathrm{TDA}}^{(n)} = \Omega_{\mathrm{TDA}}^{(n)}\,\mathbf{X}_{\mathrm{TDA}}^{(n)}$$

#### Canonical computational cost: $O(N^6)$

#### Benchmark on Thiel's set

### Bethe Salpeter equation for computing electronic excitations

#### Bethe Salpeter equation

$$\begin{pmatrix} A & B \\ B & A \end{pmatrix} \begin{pmatrix} \mathbf{X}^{(n)} \\ \mathbf{Y}^{(n)} \end{pmatrix} = \Omega^{(n)} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} \mathbf{X}^{(n)} \\ \mathbf{Y}^{(n)} \end{pmatrix}$$

$$\mathbf{A}_{la,jb} = \left(\epsilon_a^{\mathsf{G}_0\mathsf{W}_0} - \epsilon_l^{\mathsf{G}_0\mathsf{W}_0}\right) \delta_{lj} \delta_{ab} + \alpha^{S/T} v_{la,jb} - \mathsf{W}_{lj,ab}(\omega = 0)$$

$$B_{lajb} = \alpha^{S/T} v_{la,bj} - W_{lb,aj}(\omega = 0)$$

#### Hermitian solution

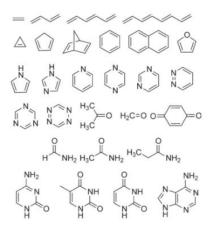
$$C \mathbf{Z}^{(n)} = \Omega^{(n)^2} \mathbf{Z}^{(n)}$$
  
 $C = (A - B)^{0.5} (A + B)(A - B)^{0.5}$ 

#### Tamm Dancoff approximation

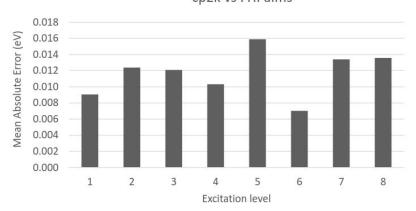
$$\mathbf{A} \mathbf{X}_{\mathrm{TDA}}^{(n)} = \Omega_{\mathrm{TDA}}^{(n)} \mathbf{X}_{\mathrm{TDA}}^{(n)}$$

#### Canonical computational cost: $O(N^6)$

Typical excitation energy: 3-5 eV



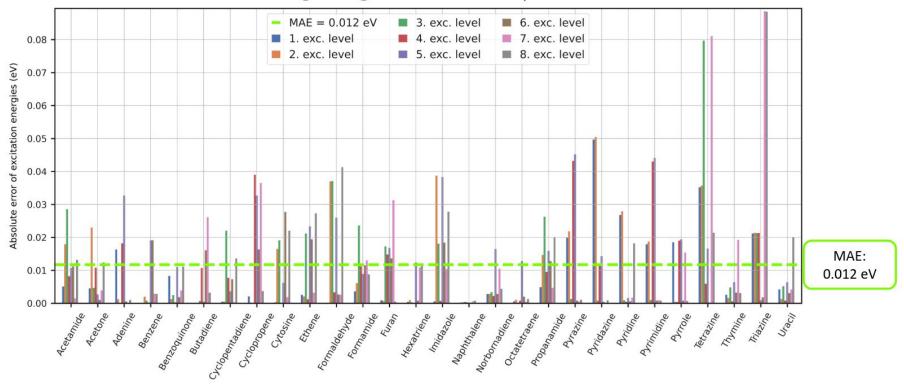
# BSE@G0W0@PBE0 on Thiel's set – cp2k vs FHI aims



### Bethe Salpeter equation for computing electronic excitations

Typical excitation energy: 3-5 eV



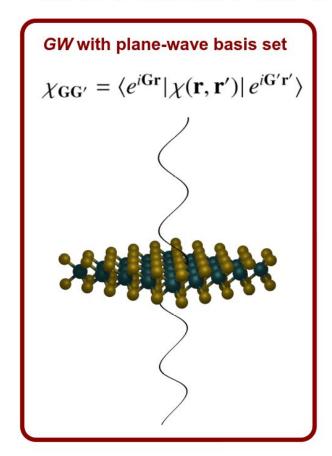


# Updates (Jan Wilhelm)

What are you currently working on or planning to work on?

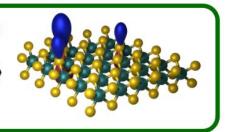


GW on 2D materials is factor 10 000 to 100 000 faster compared to plane-wave GW

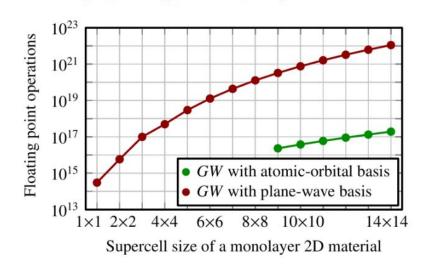


GW with atomic-orbital basis set

$$\chi_{PQ} = \langle \varphi_P(\mathbf{r}) | \chi(\mathbf{r}, \mathbf{r}') | \varphi_Q(\mathbf{r}') \rangle$$



https://doi.org/10.1021/acs.jctc.3c01230



Memory efficient iteration of 3-center integrals, extreme example: 11x11 cell of MoS<sub>2</sub>
 (363 atoms, 2D material), TZVP-MOLOPT, 1 node on Noctua (1024 GB RAM), 204 hours

$$\Sigma_{\lambda\sigma}(i\tau) = \sum_{\text{atom } A} \sum_{\text{atom } B} \sum_{\nu \text{ (at atom } A)} \sum_{Q \text{ (at atom } B)} \left[ \sum_{\mu} (\lambda\mu|Q) G_{\mu\nu}(i\tau) \right] \left[ \sum_{P} (\nu\sigma|P) W_{PQ}(i\tau) \right]$$

Comparison computation time:

### code in src/gw\_methods; input:

```
&PROPERTIES

&BANDSTRUCTURE

&GW

NUM_TIME_FREQ_POINTS 10

MEMORY_PER_PROC 15 ! Used 64 MPI processors -> memory per process: 1024 GB/64 = 16 GB

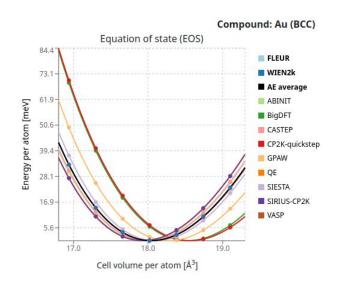
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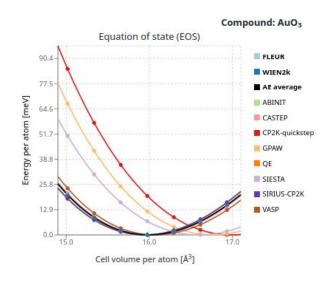
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• Optimization of CP2K on Supermuc-NG Phase 2 (Intel-GPU Ponte Vecchio, Hans Pabst)

Any plans for SCF with spinors and SOC from GTH pseudos?
 (KS-matrix will be complex with SOC and double in size, SCF with SOC will be important for calculation of forces with SOC)

# Outliers in verification paper, e.g. for gold.



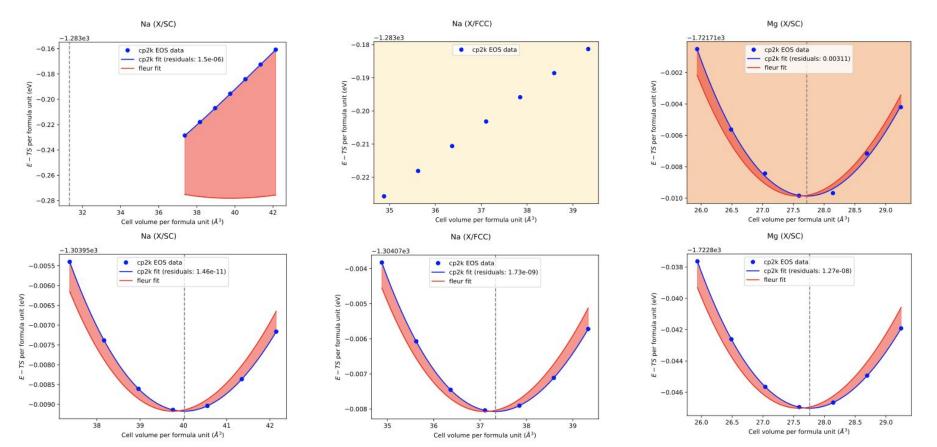


https://www.nature.com/articles/s42254-023-00655-3

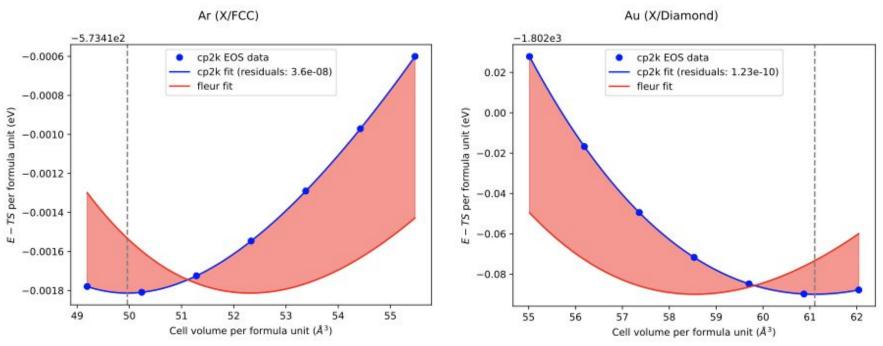
https://acwf-verification.materialscloud.org



# Outliers in verification paper: UZH Protocol



# Outliers in verification paper: UZH Protocol



Similarly: Ba, Kr, (Ne), Pd, Rh, Rn, Ru, (Xe)

# CP2K HFX with accelerators (PC2)

Electron repulsion integral engines:

- Intel FPGA (Xin Wu, Tobias Kenter)
- Xilinx FPGA (Zhenman Fang) and Xilinx Al Cores (Johannes Menzel)
- Nvidia GPUs (Marcello Puligheddu)
- AMD GPUs (just started)

Ongoing: integration into CP2K (rewrite of HFX and load balancing)



# **Current Development Efforts**

What are you currently working on or planning to work on?

- Importing a code for machine learning potential into CP2K (Alireza Ghasemi)
  - > Possible solutions?
    - Interface as a library?
    - Experience with previous case imported into CP2K?
  - Compilation?
  - ➤ Input files?



### FLAME: a library of atomistic modeling environments

#### Use cases:

- I. Machine learning interatomic potentials: using drivers/sampling methods available in CP2K (libcp2k?)
- II. Global optimization: using CP2K DFT energy/forces (libcp2k?)

### Dependencies:

- 1) spglib
- 2) futile from BigDFT

### Compilation: two independent options

- 1) Using autotools: in the case of futile via jhbuild
- 2) Using FPM, no autotools and no **jhbuild** even futile

### Input files:

If an external package is imported into CP2K, how about input files diffferences?

# CP2K@CASUS

- Job offering as CP2K developer at CASUS
  - deadline was in March 6
  - still pending interview
- FFTW3+MPI:
  - prepare PR for some refactoring
  - in progress: switch to FFTW3 blocking scheme
- Performance Engineering (Andreas Knüpfer)
- Finite Temperature RPA

# manual.cp2k.org

#### Input Reference

- Input descriptions support Latex-formulas, Markdown, and unicode incl. emoji.
- Special treatment for <u>XC SECTIONS</u>.
- New precommit check for missing spaces in multi-line descriptions.

#### **Methods Section**

- Curated structure, see e.g. the section on <u>X-Ray</u> or <u>HFX</u>.
- Live preview for pull requests on GitHub.
- Most howtos have been moved over from Wiki to Github.
- Still missing:
  - Older topics, e.g. Metadynamics
  - Overview pages that, e.g. <u>Optical Spectroscopy</u>



# Current Issues when Running CP2K



# CP2K-Release

# Open CP2K-Related Positions

- position in Jürg Hutters group
- position at CASUS: see slide CP2K@CASUS

### **CP2K-Related Events:**

#### Plans:

- Paderborn+CASUS/HZDR:
  - 3rd and 4th of April: "Post-DFT/HF methods for the condensed phase with CP2K": ADMM, RI, RPA,... (workshop with talks, virtual)
    - https://events.uni-paderborn.de/e/cp2kpostdft
  - ~Q3/24: Gromacs & CP2K on QM/MM (school with tutorial, 3-4 days in person)
- UK computer centre ARCHER2 is running an online workshop on use of CP2K on the 8th April.
  - If anyone would like to contribute a short presentation (20-30 mins) ideally linked to a tutorial or example page it would be really appreciated. RPA, LS, excited state contributions would be particularly great.
  - mwatkins@lincoln.ac.uk if interested.