



Simulation Science Laboratory 2018

An Analysis Tool for Materials Design

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Abstract — In this project, the raw data of the DFT simulations from Fleur are read, pre-processed and stored in suitable variables, then extracted various features from the data from the DFT simulations such as Fermi velocity, Effective mass. Then this data is visualized using suitable color pots for a physicist to understand the plots, material characteristics using the plots. This API is converted into a frontend GUI for easier use in future. The overall project helps physicists to solve their problem of reading the DFT simulations data.

Keywords Fleur, DFT Simulations, visualization, Front end GUI, Feature extraction.

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Acknowledgments

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List of Abbreviations

Chapter 1

Introduction

Problem Statement

Solid State physics deals with the study of large scale properties of solid materials resulting from the atomic scale properties. A solid state physicist can get to know the atomic scale properties from the experiments conducted on the material. Another way to elucidate physical properties is numerical simulation by means of Density Function Theory (DFT). This method is computational Quantum mechanical modeling method to investigate the electronic structure of many body systems of an atom or molecule. Electron density, Intermolecular forces, charge transfer excitations, calculation of band gap among others are properties that are made amenable by DFT simulations.

Fleur is one such DFT simulation which is developed by physicists at Juelich Forschungszentrum and most importantly its an open source and anyone can access it and use it. Like any other simulation, DFT simulation also outputs a lot of data for a solid state physicist to understand and determine the properties of a cell structure of a molecule through which physical properties of the material can be determined. Fleur is run specifying the cell structure of a molecule and so can be used on any molecule's cell structure and obtain the characteristics of same.

Generally it is used to find/simulate properties of solids with impurities in cell structure. Fleur outputs the data of DFT simulation. It gives the data of ground state and excited state properties of solids. These data which are raw are generally not accessible directly and have to be processed through steps for any solid state physicist to understand the raw data extracted from Fleur DFT simulation. This becomes the major problem and which is dealt in this project. The goal of this project was to implement a complete data analysis pipeline for this application. The steps include preprocessing followed by data exploration and visualization.

Motivation and Requirements

An API which solves the physicist's problem of understanding the simulation data which transforms raw data to useful data such that it is process and visualizes. This API has to process the output files from fleur directly so that it is easier for physicists to get the data processed without external effort. Having said this, modularization and easy maintainability of code also matters since the format and structure of simulation results keeps varying over the time with the development of simulation code and more data may be collected from simulation.

This API shouldn't only be solving the physicist's problem but also should be fast enough in terms of computation. As known, using of computer code to get outputs and may cause confusions in using it as computer code and there is always chance of altering the code while using it. So a front end GUI is needed such that a physicist can use the GUI to input the parameters and to output the processed data. As this is used for research purpose and the features such as plots, images and other should be of high quality and resolution such that there is no uncertainty of the results.

Steps

As a project, step by step procedure has completed to complete the task of processing and visualization.

- Understanding the problem: Firstly the theory of DFT simulation, Fleur code, band plots, density plots, file format of data and other necessary theory needed are learnt and problem of the project to process the data in different steps is understood.
- Pre-Processing: Once problem is clearly understood, first step of project comes to pre-processing the data. Reading the data, sorting the data and storing it sorted which can be used in further stages of implementation.
- Exploring the data: From the raw data, not just band plots but many features can be extracted. Finding out the features through exploring the data and trying to understand and extract those features is done.
- Visualization: The data which is pre-processed and extracted need to be visualized into plots such that any user with solid state physics background can understand it with an observation.
- Front End: A GUI is developed so that its easier in future for any physicist to just run the GUI and get the plots and other visualization instead going through hassle of code and letting a chance of code being disturbed unintentionally
- Results and lookup: Getting to know the features extracted, studying it, reporting the same.

Chapter 2

Theoretical Background

...some introduction sentence depending on into chapter...

Fleur computes the electronic structure in crystals using the Density Functional Theory approach (DFT), which is the state of the art method for this problem. The many-body Schrödinger equation, that can be used to describe electrons in solids, is almost impossible to solve directly, because the storage of the wavefunctions of each of the N electrons in the system at each spatial coordinate exceeds the memory of any currently available computer for even quite small N . This motivates the DFT approach, that uses two fundamental theorems to reduce the computational complexity of the many-body problem significantly: The Hohenberg-Kohn theorem allows to use the electron density instead of the N -electron wavefunctions to uniquely characterize the ground state of a system. With the Kohn-Sham equations, the interacting Hamiltonian of the system can be replaced by non-interacting equations with an effective potential. This approach reduces the dimensionality of the problem from N^3 to 3 and trades the interacting Hamiltonian for a set of non-interacting equations that have to be solved self-consistently. In general, the DFT approach is not just limited to computations of electrons in crystals, but is also for example used in chemistry to compute nonperiodic molecules. The output of a DFT calculation includes various physical quantities (for example the 3-dimensional spatial electron density, which is the central quantity in DFT calculations), but for this project we focus on two quantities that are easy to interpret, already reduced in dimensionality and frequently used in both experimental and theoretical physics.

The band structure $E(\mathbf{k})$ represents the eigenenergies of the eigenfunctions of the Hamiltonian for each crystal momentum \mathbf{k} . It is the dispersion relation of electrons in the crystal and relates allowed momenta and energies. In general, $E(\mathbf{k})$ is defined at any point inside the Brillouin zone, which is a special choice of the unit cell in the reciprocal lattice of the crystal. Both, the real-space lattice and the reciprocal lattice are shown for a face centered cubic (fcc) crystal in figure 2.1. For larger unit cells with fewer symmetries, the Brillouin zone can be much more complicated than in the shown example. In order to reduce the dimensionality of $E(\mathbf{k})$ with

$k \in \mathbb{R}^3$, the dispersion relation is only sampled along a discrete one-dimensional path between high symmetry points in the Brillouin zone. This path still contains most of the relevant physical features.

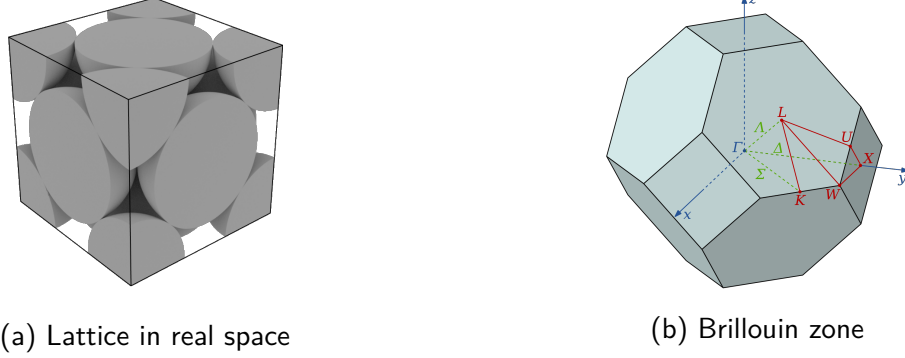


Figure 2.1: Brillouin zone of a fcc lattice. The red curve in the reciprocal lattice represents a possible sampling path of $E(\mathbf{k})$ in the reciprocal lattice.

The second central quantity in this project is the Density of States (DOS) $D(E)$, which describes the number of states per energy interval that is independent from the crystal momentum. In this sense, the DOS is also derived from $E(\mathbf{k})$, but instead of just selecting a subset, the \mathbf{k} dependency is summed out. Both complementary quantities together are a common choice for comprehensive visualizations of electronic structure data while still capturing most of the important physics.

For applications, it is useful to investigate where the contributions to $E(\mathbf{k})$ and $D(E)$ come from. Therefore, the data contains the weights of each basis function of the DFT calculation belonging to the individual atom groups and the orbitals. This means, spatial information about the system can be restored by considering only contributions from certain atom groups. (An atom group contains all atoms that are equivalent with respect to symmetries of the real space lattice.) On the other hand, the projection on the hydrogen orbitals s , p , d , f encode information about the shape of the wavefunction at each atom. These contributions are stored in the form of relative weights that can be summed to include contributions for multiple groups and orbitals. In case of distinct spins in the crystal, $E(\mathbf{k})$, $D(E)$ and the weights can be different for both spins and are therefore stored individually.

Chapter 3

Implementation

As per the requirements expounded upon in the introduction, the deliverable of the project should be a finished software product. The software is written in Python so as to integrate easily with the research group's ongoing software projects around the Fleur code [Blü+18]. These are chiefly the group's materials science tool collection `masci-tools` [RBR18], where also this project's code is hosted, and the 'Automated Interactive Infrastructure and Database for Computational Science' (AiiDA) [Piz+16]. The product stakeholders split into frontend users and code developers. In order to accommodate this, the product is organized into three unidirectionally dependent subpackages or -modules, see Figure 3.1a.

An important design consideration was to account for unknown use cases. This has been realized in each submodule by the decoupling of **interface** and **implementation**. The interfaces do not rely on any specific input file format, visualization method or library, unlike the implementations for a specific task or **application**. In this chapter, the word 'application' denotes the band structure and density of states visualization, and for these applications, implementations are provided.

This design choice was also one reason why the product does not reuse any of the `masci-tools` routines which partly solve quite similar problems, but seemed to be too specialized in a cursory code review. For these developers, one added value of the project product could be to inspire the hopefully easy integration into a common interface, where the current abstraction level could serve as a starting point.

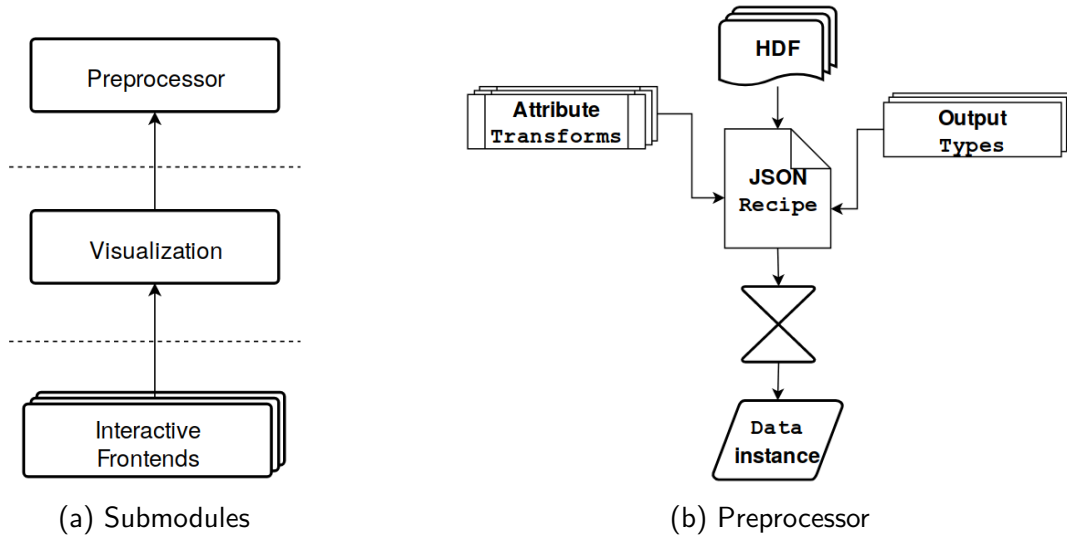


Figure 3.1: Module Design.

HDF Preprocessor Module

Interface

This is the 'backend' of the tool. It is basically a file reader for the input data, for example a Fleur simulation output. Supported formats are the Hierarchical Data Format (HDF) [Kor11] for the band structure, and a simple Fleur-specific comma-separated values (CSV) format for the density of states (DOS).

The HDF format is a flexible binary container for all kinds of common binary and text file formats. Each file which constitutes a Dataset inside the HDF file. The format supports metadata annotation and high-throughput input/output (I/O). As a consequence, it is considered by some developers in some application domains that rely on numerical simulation codes, to be one possible base for the establishment of common domain-specific rich data exchange standards in order to increase code interoperability. These developers are in the process of extending their codes' I/O capabilities towards that end. However, HDF's flexibility comes at the price of a relatively complex Application Programming Interface (API) as the keyhole for all operations.

The preprocessor module tries to hide that complexity by introducing the concept of Recipes, see Figure 3.1b. A specific application Recipe is a dictionary that aims to describe a complete [Extract-Transform-Load](#) (ETL) pipeline for one specific application of the original data. The 'extract' is the reading of a dataset from HDF, the 'transform' a sequence of once-through functions applied to the the dataset, and the 'load' the aggregation of all transformed datasets into one runtime object which has all the methods for operations on the data that are going to

be used later on in the intended application.

The ‘transform’ and ‘output’ type methods are defined in hierarchical Transform and Output_Type classes. Multiple inheritance is used to sort them from general to application-specific applicability. This structure is built using Python’s AbstractBaseClass (ABC). The advantages of the ‘recipes approach’ are:

- All ETL processes for one application are collected in one simple list (the recipe), not locked in different code locations with conflicting contexts. In this list, entries can be sorted in any manner, e.g. alphabetical for perusal. Thus a recipe also serves as a concise documentation of how an application-domain HDF format should be handled.
- Recipes are de/serializable (can be read from and saved to disk) and thus be machine-created and manipulated, for instance in a workflow pipeline.
- The ETL processes, when declared in this interface, can be easily reused across applications. A recipe can combine different output types into a new type.

The feature that enables this flexibility is **type introspection**: the preprocessor processes the datasets listed in the recipe in the order of their mutual dependencies as found in the introspected listed transform and output methods. When all transformed datasets have been added to the object, all specified output types are searched and all their methods and attributes added. Thus the output object’s type is defined at runtime, when the preprocessing is finished.

Implementation for Band Structure Visualization

The frontend has to draw three kinds of plots: a 3D atom plot of the unit cell or supercell, and a combined band structure and DOS plot sharing the same vertical energy axis. If no DOS data is present, the DOS part of the plot shall be omitted. All three plots are controlled by one set of graphical frontend control elements (widgets) for varying the parameters. In the current implementation, the data for the first two plots come from a HDF file, while the data for the DOS plot come from CSV files.

The band structure plot is a scatter plot. It plots discrete $E(\mathbf{k})$ data from the simulation. Firstly, the plotter needs the k -path (where $|\mathbf{k}| = k$) for the horizontal axis. The preprocessor, having received the recipe FleurBands, computes it from the k -points in the HDF in a transform. Secondly, the plotter needs the eigenenergies for every point on the k -path, labeled by the band index ν , and its associated l -like charge $n_{s,k,\nu,g,l}$. This dataset is fivedimensional, and represents the contribution of spin s , point k point on k -path, band index ν , atom group g , and character or orbital l (here: only s,p,d,f) to the specific eigenenergy. So plotting must involve a dimension reduction. The preprocessor resolves the processed data into the like-named output type FleurBands. This type has a data selection method. The respective BandPlot type calls this selector with a user selection of subsets of all (s, k, ν, g, l) . The method then computes the according **effective weight** shown in Equation 3.1. The plotter uses that for the dot size

of each $E(k)$ in the plot. Before rendering, the plotter normalizes the energies to the Fermi Level.

$$W_{s,k,\nu}^{\text{eff}} = \left(\frac{\sum_{\substack{g \in \text{groups} \\ l \in \text{characters}}} n_{s,k,\nu,g,l} N_g}{\sum_{\substack{g \in \text{all groups} \\ l \in \text{all characters}}} n_{s,k,\nu,g,l} N_g} \right) \left(W_{s,k,\nu}^{\text{unf}} \right)^\alpha \quad (3.1)$$

N_g denotes the number of atoms in a group. $W_{s,k,\nu}^{\text{unf}}$ is the unfolding weight and α its exponent. In the frontends, the latter is also a user control. The effect of unfolding is illustrated in Figure 3.2 for a toy example of a monatomic chain, with a two-atom supercell of size a' representing $\alpha = 0$ and a one-atom unit cell of size a representing $\alpha = 1$. A use-case is discussed in the Application Chapter 5.

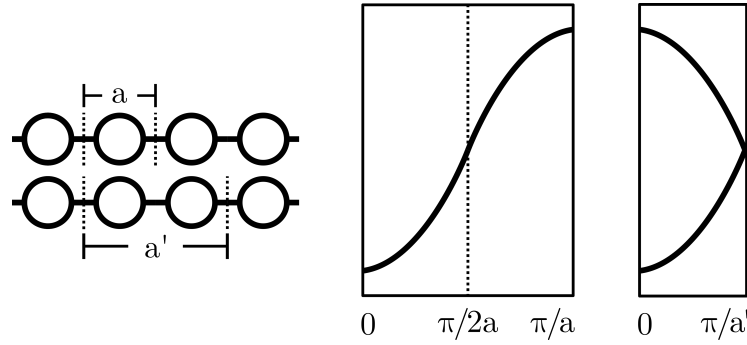


Figure 3.2: Band unfolding: example for a simple monatomic chain [Hof87].

Even for small band structures, the selector has to access on the order 10^7 individual data points for every selection change. Optimizations were introduced which included:

- a cutoff filter that skips effective weights too small to show up on the plot,
- use of optimized Numpy functions like `np.tensor` for the effective weight summation,
- buffering of unchanged data between two selections,
- array reshaping.

Together, these tweaks achieve a speedup of approximately 10^2 in plotting speed. Thanks to that, the tool remains usable even when the input HDF is in the 10^2 MB range.

Visualization Module & Interactive Graphical Frontends

Visualization Module

The Python visualization landscape abounds with a rapidly evolving plethora of plotting libraries for different application contexts and technology stacks [VR17]. Thus the project's visualization module's first design objective was to account for that fact by decoupling it from any specific library use, and modularizing it for intended applications. This structure again is built using Python's `AbstractBaseClass` (ABC) interface and multiple inheritance. Each application is represented by an abstract base class that contains the common plotting method signatures. Likewise, each plotting library is represented by an abstract base class that contains library-specifics. An *implementation* inherits both from one library base class and one or more application base classes. See Fig. 3.3 for an impression. Thus switching the library in a use context should require minimal adjustment, and a new application (base class) can be built from existing ones.

The second design objective was for the plotting methods to hide all interactions with the actual plotting library used under the hood. The application's base class' method arguments should only be tied to the data, not the plotting library. Thus different frontend implementations need no or minimal individual setup before they can call the same method for one specific plot and receive the identical visualization with identical interactive behavior.

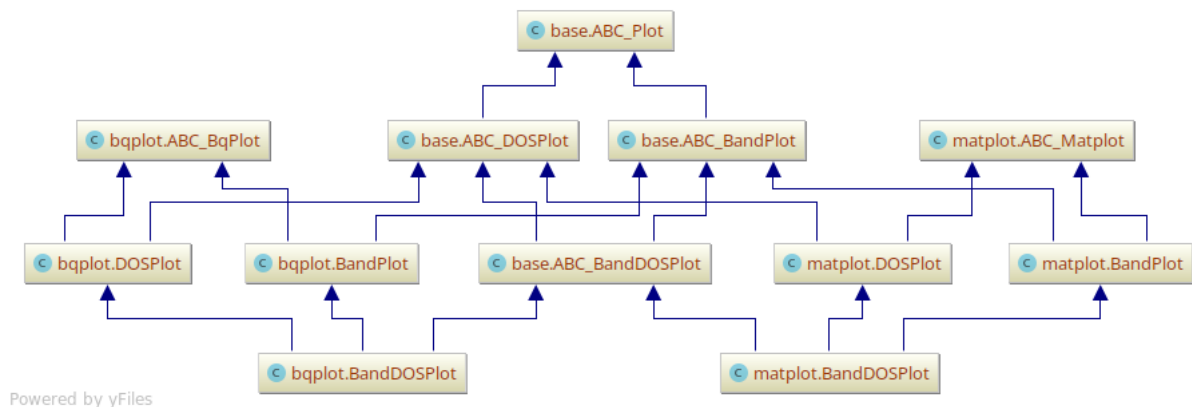


Figure 3.3: Visualization Module Design: Example Structure for two applications BandPlot and DOSPlot, and two plotting libraries matplotlib and bqplot.

Desktop Frontend

A Desktop Frontend is always a helping hand for the physicists to just execute it on the computer when one wants to go through the plots of the raw data that is already in the computer.

Since the reading of HDF files and preprocessing of the data is done using Python, it is decided that it would be better to use Python for frontend development. Considering various packages for frontend in python such as PyQt, Tkinter and other libraries available for Desktop GUIs, Tkinter is selected, as it is bundled with Python and thus meets the requirement of easy maintainability best. It is a package where every button can be designed and can be assigned to a function. GUI types like Label, Button, CheckButton, ListBox, Canvas for plots, Tab for viewing each plot in different tab have been used to make a simple Desktop front end. It is simple to use with limited options for what the end-user, the solid-state physicist, needs. It is also easy to convert the Desktop frontend code into a executable software and run in any system without any installation.

Web Frontend

As Web Frontends continue to replace traditional Desktop Frontends in many application domains [Dou+08], so Python-based Frontends and visualizations are increasingly moving towards the browser, too. There, GUIs with interactive visualizations are often called **Dashboards**. For this project, a survey was undertaken to find the most suitable technology stack for a Web Frontend. The full survey is documented in [Was19]. The requirements for the solution, on top of those stated in the introduction, were defined as follows: The solution...

1. **openness**: relies solely on Open-Source-Software (OSS) with licensing suitable for academic use, sports a stable release cycle, developer base and documentation,
2. **dashboarding**: features graphical control elements (widgets) that interact with InfoVis¹ plotting libraries,
3. **deployment**: ideally works like any web service, i.e. only a modern web browser is required to use it,
4. **maintenance**: requires only Python and no Web Development knowledge like e.g. Javascript, with respect to the product stakeholders.

The last point implies a client-server model where the dashboard app is hosted on a remote machine. This model requires a communication framework and protocol between the Python interpreter running on the server and the JavaScript interpreter running in the client browser. As per requirement no. 4, unlike a generic Python web framework like e.g. [Flask](#), the framework should take care of that communication by itself. Four major frameworks were identified which fulfill the first three requirements: [Project Jupyter](#), [PyViz](#), [Bokeh](#), and [Dash by Plotly](#). The last two only partially fulfilled the requirement no. 4, so they were discarded. PyViz is the newest contender among these four. Its expressed goal is to untangle the Python visualization

¹InfoVis libraries: visualizations of information in arbitrary spaces, not necessarily the three-dimensional physical world. Example: matplotlib. SciVis libraries: visualizing physically situated data. Example: VTK [Bed18].

jungle by providing one high-level API that ties together all major Python InfoVis libraries and data formats, including support for dashboarding. That ambitious goal comes at the price of sacrificing support for 3D plotting [aut18], which was needed in this project for the atoms plot. So PyViz had to be discarded.

That left Project Jupyter. By now, a wide variety of popular plotting libraries have made their tools capable of working with Jupyter's widget library `ipywidgets`. However, Jupyter only partially fulfills requirement no. 3 – a Jupyter notebook (app) cannot, by itself, be published (deployed) as a stand-alone website outside a live Jupyter environment [Bed18]:

[...] “However, despite their web-based interactivity, the `ipywidgets`-based libraries (`ipyleaflet`, `pythreejs`, `ipyvolume`, `bqplot`) are difficult to deploy as public-facing apps because the Jupyter protocol allows arbitrary code execution” [...].

To avoid requiring users to setup a working Jupyter environment on their machine, the go-to solution for this problem is to setup a [JupyterHub](#) multi-user server. This still requires users to register an account there, so it's not completely open. Fortunately though, the intended users are contributors to the AiiDA project, and so should have access to the JupyterHub-based [AiiDaLab](#) service where the app can be registered. Details on this procedure and alternative hosting solutions can be found in the developer section of the manual [4 on page 16](#).

Chapter 4

Manual

The project code and documentation is hosted on the `masci-tools` repository [RBR18] under the branch `studentproject18ws`. All of the project code resides in the folders `binder` (for the Web Frontend Demo) and `studentproject18w` (all code and documentation). The `README.md` serves as the manual. Therefore, the remaining part of this chapter is a T_EX-ified version of that `README.md`.

.....
SiScLab 2018 Student Project **Analysis Tool for Materials Design**. Written in Python3.

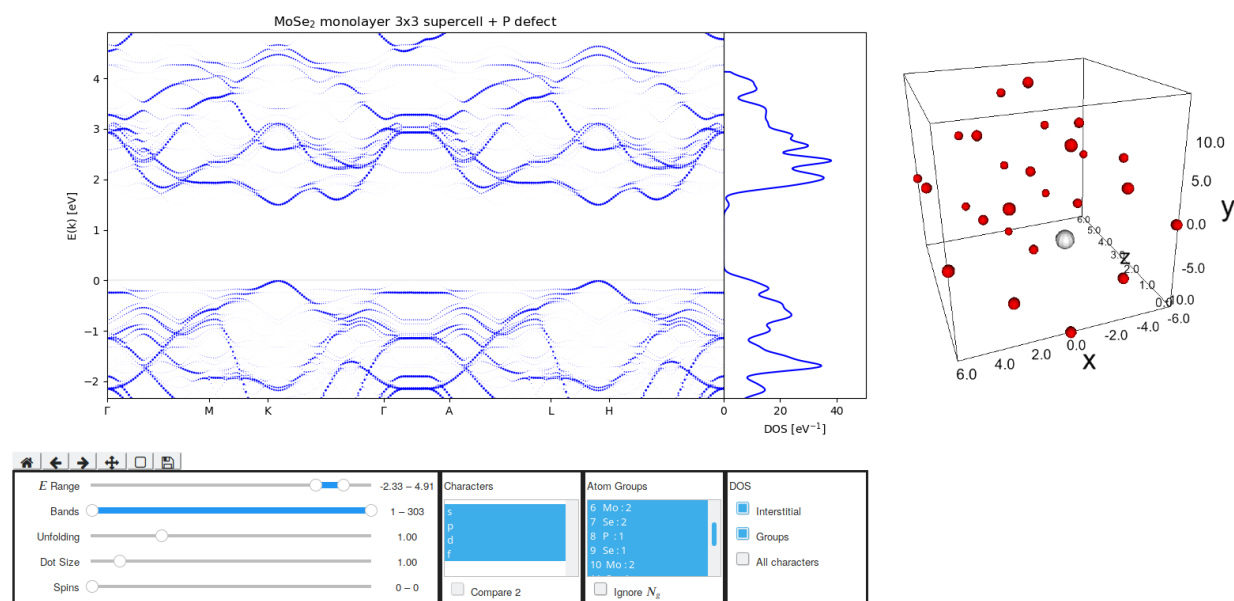
Authors: [Johannes Wasmer](#), [Christian Partmann](#), and [Praneeth Katta](#).

Overview

This subfolder `studentproject18ws` is currently a largely independent side-project accompanying the main module `masci-tools`. It was created in a student project, and consists of three submodules:

- `preprocessor`: a HDF reader interface, and one implementation for [Fleur](#) band structure simulation output
- `visualization`: a plotting interface, and one implementation for [Fleur](#) bandstructure+DOS plots
- `frontends`: a Desktop GUI and a Web Dashboard (Tk and Jupyter) for interactive [Fleur](#) bandDOS plots.

A more thorough description and example use cases can be found in the project [report](#) and [presentation](#).



For Frontend Users

General Remarks

These remarks apply to all frontends.

Though the Desktop and Web Frontend are functionally identical, there might be small differences in how the controls are used and how they are labeled.

File Input

The frontends currently expects band structure data in the HDF output format of [Fleur](#). The density of states data is expected to be in the CSV output format of [Fleur](#), one file per spin. If no density of states files are supplied, the frontend will just draw a band structure plot (BandPlot) and omit the adjoined density of states plot (DOSPlot). Thus, in the following BandDOSPlot stands for both kinds of plot. The Web Frontend will only show controls for data that is present in the input (e.g., DOS and spin controls).

Desktop Frontend

Installation

A windows executable file (.exe) is made by packing all the required packages into the file. Any modern PC running on windows can run the frontend without any installation process and there

is no prerequisite to execute this executable file. PC need not have python or other packages installed.

For executables for other operating systems, please contact the developers.

Usage

Desktop based front end GUI is easy to use. By just running the .exe file provided will open the software with the packages involved to run the software. There are three tabs/windows in the software. In the first tab, absolute paths to the input data files must be entered in this order: HDF and (optional) DOS file for spin '0' and '1'.

Controls for all plots:

- **Atom Groups:** draw the BandDOSPlot only for the selected symmetry groups.
- **Character:** select one or more band Characters (orbitals) 'S', 'P', 'D', 'F'.
- **Spin:** select any one spin or both spins.
- **Marker size:** Default marker size of 1.0 is selected. However, user have a choice to increase the marker size of the dots (eigenenergies) plotted in the BandPlot.
- **Ymin, Ymax:** This control is used to limit the range energy range of the BandDOSPlot.
- **BandMin, BandMax:** This control is used to limit the band range of the BandDOSPlot.
- **Update, SaveButton:** Update the BandDOSPlot to the newly selected data by user. Save the the plot as a PDF on disk.
- **Exponential weight:** The unfolding exponent for supercell calculations (see [report](#)). Value 0.0 means no unfolding. If the calculation is done with a unit cell, this control has no effect.
- **Compare 2Characters:** When a user wants to compare 2 characters, this button makes the BandPlot show the influence of each character to each eigenergy using a sequential (2) colormap. The control is disabled if other than two characters are selected.
- **Ignore Atom group:** This button allows an option to ignore the atom groups.

Controls for the DOSPlot only:

- **Select groups:** include selected atom groups in the DOS
- **Interstitial:** include the interstitial in the DOS
- **All characters:** include all characters in the DOS regardless of character selection. In the DOS CSV file, different input data is used (a summed column).

After the Update button is clicked, a BandPlot or BandDOS plot is produced in Tab 2. A 3D atomic plot is produced in Tab 3.

Troubleshooting

If the BandPlot is not visible:

- Click update two to three times.
- Check if the three input files (if any) are belonging to the same Fleur calculation and selected appropriately.
- Check if at least one Atom Group, one Character, one Spin is selected.
- Check if Ymin is less than Ymax and similarly BandMin is less than BandMax such that software is able to plot.

If the DOSPlot is not visible:

- Make sure either Select Groups or Interstitial is selected.

If the problem persists, try restarting the GUI. If that fails, please open an issue or contact the developers.

Web Frontend

Access

The Web Frontend is a Jupyter Dashboard. It is in experimental state (no fileupload yet). You can try it out [here on Binder](#). You can run it locally (see developer section). If you have an [AiiDaLab account](#): the dashboard is planned to be published as an app there.

Usage

Using the Dashboard should be self-explanatory to the domain user. Some tips:

- if the plot window is not on startup or gets stuck, reload/rerun once.
- unlike the Desktop frontend, plot updates are instantaneous.
- unlike the Desktop Frontend, empty selections are impossible.
- multi-selection boxes: use ctrl or shift to select multiple items.
- slider values can also be typed into the adjoining text box.
- try out the zoom and pan tools below the plot, they're useful.

For Developers

Installation

Create project virtual environment

With conda (recommended): - [Install Anaconda \(3 recommended\)](#) - Install the environment `masci-stupro` with the necessary and recommended dependencies:

```
1 conda create -f environment.yml
source activate masci-stupro
```

With virtualenv (untested):

```
virtualenv masci-stupro
source masci-stupro/bin/activate
3 pip install -r requirements_pip.txt # install requirements
```

Programmatic use

Though masci-tools is available via PyPI, there is currently no plan to integrate studentproject18ws. If you want to use it in your code, clone the repo, use it in an IDE, or append the path to your sys.path:

```
import sys
2 if path_repo not in sys.path:
    sys.path.append(path_repo)

# now import works
from studentproject18w.hdf.reader import Reader
7 # ...
```

In this example, a Fleur HDF file is preprocessed using the Recipe FleurBands. The resulting output data with the extracted and transformed HDF datasets and attached load methods (Extract-Transform-Load) is then passed to a plotter, alongside some DOS CSV files for a bandstructure plot using matplotlib as backend library.

```
from studentproject18w.hdf.reader import Reader
from studentproject18w.hdf.recipes import Recipes
3 from studentproject18w.plot.matplot import BandDOSPlot
import matplotlib.pyplot as plt

data = None
reader = Reader(filepath=filepath_hdf)
8 with reader as h5file:
    data = reader.read(recipe=Recipes.FleurBands)
    #
    # Note:
    # Inside the with statement (context manager),
    # all data attributes that are type h5py Dataset are available
13 # (on-disk access). When the statement is left, the HDF5 file
```

```

# gets closed and the datasets are closed.
#
# Use data outside the with-statement (in-memory access: all
# HDF5 datasets converted to numpy ndarrays):
18 data.move_datasets_to_memory()

plotter = BandDOSPlot(plt, data, filepaths_dos)
(fig, ax_bands, ax_dos) = plotter.setup_figure()
23
data_selection = some_selection_process()
plotter.plot_bandDOS(*data_selection)
plt.show()

```

Try out Web Frontend locally

The demo notebook with the Dashboard is `studentproject18w/frontend/jupyter/demo/demo.ipynb`.

If using Jupyter Notebook

If using Windows, omit keyword `source`.

```

source activate masci-stupro
cd mypath/masci-tools/studentproject18ws/
jupyter-notebook .
4 # if Home is not set to this dir, try this instead:
# /home/you/anaconda3/envs/myenv/bin/python /home/you/anaconda3/envs/myenv/bin/ ↵
↵ jupyter-notebook .

```

If using Jupyter Lab

Additional installation step needed:

```

source activate masci-stupro
jupyter labextension install @jupyter-widgets/jupyterlab-manager jupyter- ↵
↵ matplotlib ipyvolume
cd mypath/masci-tools/studentproject18ws/
jupyter-lab

```

Frontend Deployment

Desktop Frontend

To create executables for different operating systems, use [PyInstaller](#). The target file is `frontend/tkinter/gui.py`.

Web Frontend

The Web Frontend is currently a single Jupyter Notebook. In order to publish it as a usable standalone app, additional work has to be done.

- (recommended: create `frontend/jupyter/Dashboard.py` widget and put code of [demo_back.ipynb](#) notebook inside it. Use [aiidalab-widgets-base > StructureUploadWidget](#) as a template. Create `frontend/jupyter/Dashboard.ipynb` notebook. Use [StructureUploadWidget Demo Notebook](#) as a template.)
- Add [fileupload](#) to widget (again, like in `StructureUploadWidget`. See [binder_fileupload_test.ipynb](#) notebook for a demo that works with binder.)
- Now the Web Frontend should work on Binder.
- For publishing the app on AiiDA Lab, the app has to be registered in the [aiidalab-registry](#).
 - The project code is in Python3, but `aiidalab` requires Python2. So the code has to first be backported by hand using the `future` package. If this takes too long, maybe try the tool [3to2](#).
 - Use the simplest app in the registry, [aiidalab-units](#) as a template. Adapt code.
 - Try it out first in the [Quantum Mobile Virtual Machine](#), which has `aiidalab` installed and configured. Else try it in a virtual environment with `aiidalab` installed from PyPI.
 - Register the app.

Note: other publishing options besides Binder and AiiDALab are listed [here](#). For instance, [Google Colaboratory](#) is a free Notebook hosting service that allows file upload.

Extending the code

Use Case: HDF with DOS data included

The Fleur output HDF format is expected to change and incorporate more data. In turn, this project's code has to be extended as well. The procedure is outlined for an example use case: the incorporation of DOS data into the band structure HDF (thus eliminating the need for separate DOS CSV files). The instructions show how to extend the preprocessor, the visualization and frontend submodules to that scenario.

- Add a new output type to `hdf/output_types`, say `FleurBandDOS`. Let it inherit from

output type `FleurBands`. If you want an output type just for the DOS as well, add a type `FleurDOS` and let `FleurBandDOS` inherit it.

- Add a new recipe to `hdf/recipes` e.g. `FleurBandDOS`. Copy unchanged things from recipe `FleurBands`.
- If needed, add new transforms to `hdf/input_transforms`. Adhere to the transform function standard there. If there are mutual dependencies, add them to the list in the top of the file.
- Add a DOS data selection method to the output type `FleurBands`. The `DOSPlot` in `plot/base` types will need those to plot the DOS plot. Simply adapt from the function in `dos/reader` for the DOS CSV files, adopt the identical signature.
- In the `DOSPlot` types in submodule `plot`, add a switch to the constructor that can distinguish the three cases (bands, bands+CSV DOS, bands+HDF DOS). Use the switch in the `plotDOS` methods, and for the case bands+HDF DOS, call your new `FleurBandDOS` function.

Extending the Visualization (Plots)

- In addition to the inheritance scheme based on Python `AbstractBaseClass` (ABC) detailed in the [report](#), the Plot types in `plot` have an additional facility that helps to keep the appearance of different Frontends synchronized: each type has an attribute `icdv` of type `InteractiveControlDisplayValues`. This is an ABC with the same inheritance as the application Plot types. For every plot control argument that an application type's Plot type exposes in its methods' signatures, this attribute describes the parameters of the accompanying control widget in the Frontend text label, default values, value ranges, and so on. In the current code, only the Web Frontend uses this facility, so the labels in the Desktop Frontend differ slightly.
- It is worth pointing out that unlike other languages, Python does not enforce implemented abstract methods to have the same method signature. However, when a new implementation for a different plotting library/backend is added, it is recommended to adopt the `abstractmethod` signature. That way, changing the backend in a use case only requires to change the import.

Open Issues

- The `FleurBands` data selection method could be optimized even more by replacing *all* numerical operations with `numpy` routines.
- Running the Frontends in a debugger or with a counter reveals: on a plot selection changes, the plot seems to be redrawn not once but several times. The cause could not

be found so far.

- In the Desktop Frontend, the Update Button has to be clicked several times.
- In the Web Frontend, on startup, the plot is only visible after two loads/cell runs.

.....

Chapter 5

Applications

To illustrate the use of the graphical user interface, two different physical applications are shown in the desktop and the web frontend, respectively. From the physics point of view, the example in the desktop version focuses more on the density of states and the visualization of spin contributions, while the dataset in web frontend focuses on the band structure $E(\mathbf{k})$ and the visualization of defect states in supercells.

Web Frontend: MoSe₂ Crystal

Figure 5.2 shows the visualization of a band structure calculation of a 3 dimensional Molybdenum diselenide (MoSe₂ bulk) crystal using the default settings of the GUI. Even with the default settings the band structure plots clearly indicate, that MoSe₂ is a semiconductor, since there are no states at the Fermi level. Because the minimum of the conduction band is located at an other k as the maximum of the valence band, the plot shows that MoSe₂ has an indirect band gap. This indicates that for the transition with the smallest energy difference between valence and the conduction band, both, energy and momentum have to change.

In contrast to the 3 dimensional extended MoSe₂ crystal, a MoSe₂ monolayer (see Fig. 5.3) has a direct band gap but is still a semiconductor. Furthermore, the MoSe₂ monolayer has a defect atom in every 9th unit cell and the DFT computation is therefore done in a 3×3 supercell to restore periodicity. This is the reason for the much greater number of states

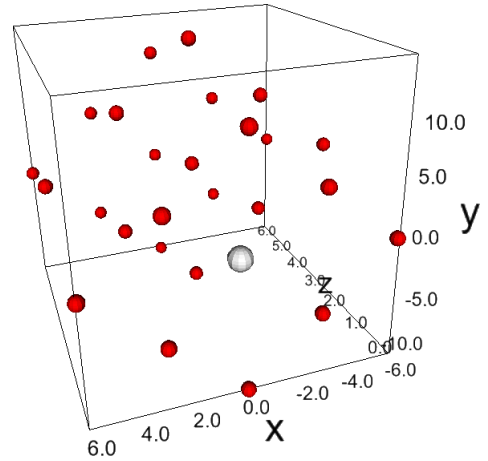


Figure 5.1: Atom Plot of a MoSe₂ monolayer with the defect atom selected in the web frontend

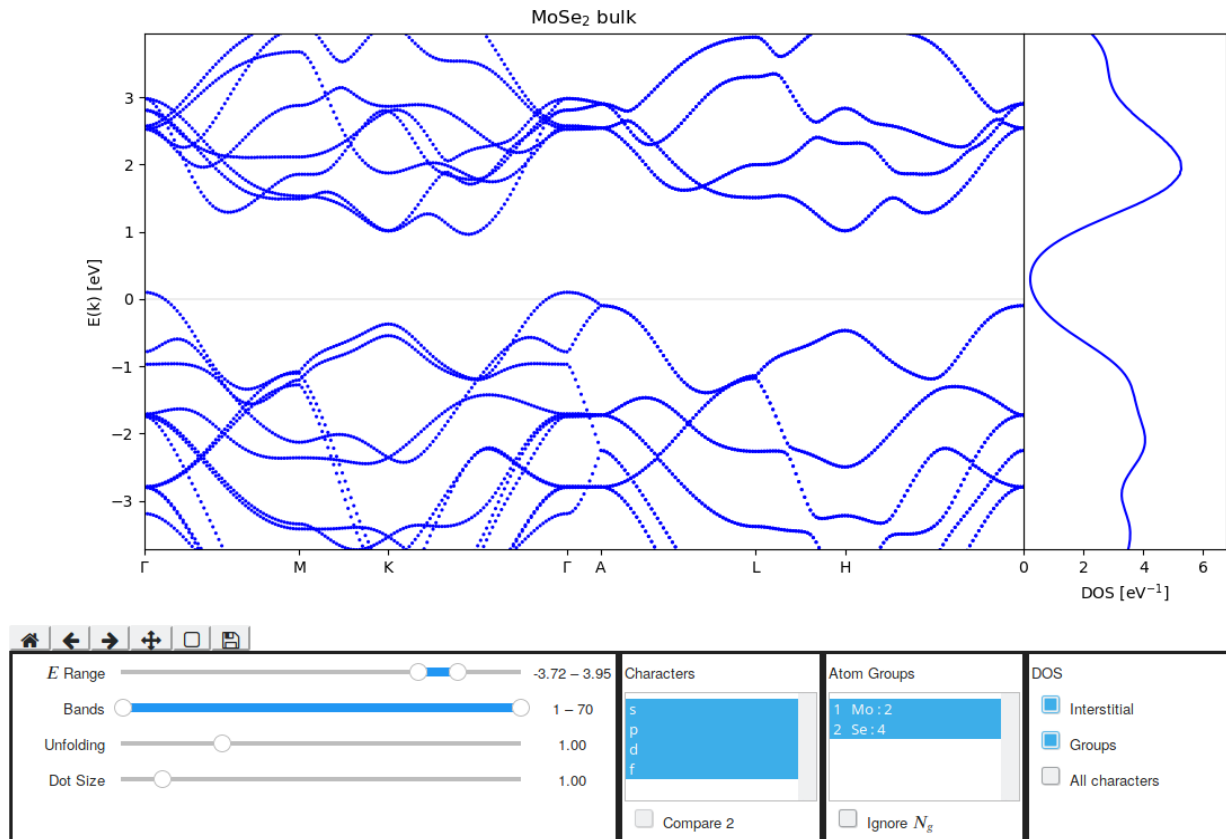


Figure 5.2: Bandstructure of a MoSe_2 crystal using the default settings of the web frontend

in the band structure plot.

Since the Brillouin zone of the supercell is smaller than the Brillouin zone of the same crystal without the defect, the supercell Brillouin zone is unfolded to the same size as the Brillouin zone of the unperturbed lattice. To account for the fact that the defect is only present in every 9th cell and its relative importance for the spectrum is therefore degraded, an unfolding weight is introduced to visualize the relative importance of bands in the unfolded Brillouin zone. By default, the unfolding weight is used by our visualization tool, but it can gradually be turned off in order to highlight the impact of defect states. This is shown in figure 5.4. To even better visualize the defect state, it would also be possible to select the atom group belonging the defect atoms only. In this example, the analysis with reduced band unfolding shows, that there are many more direct band gaps origination from the defect state.

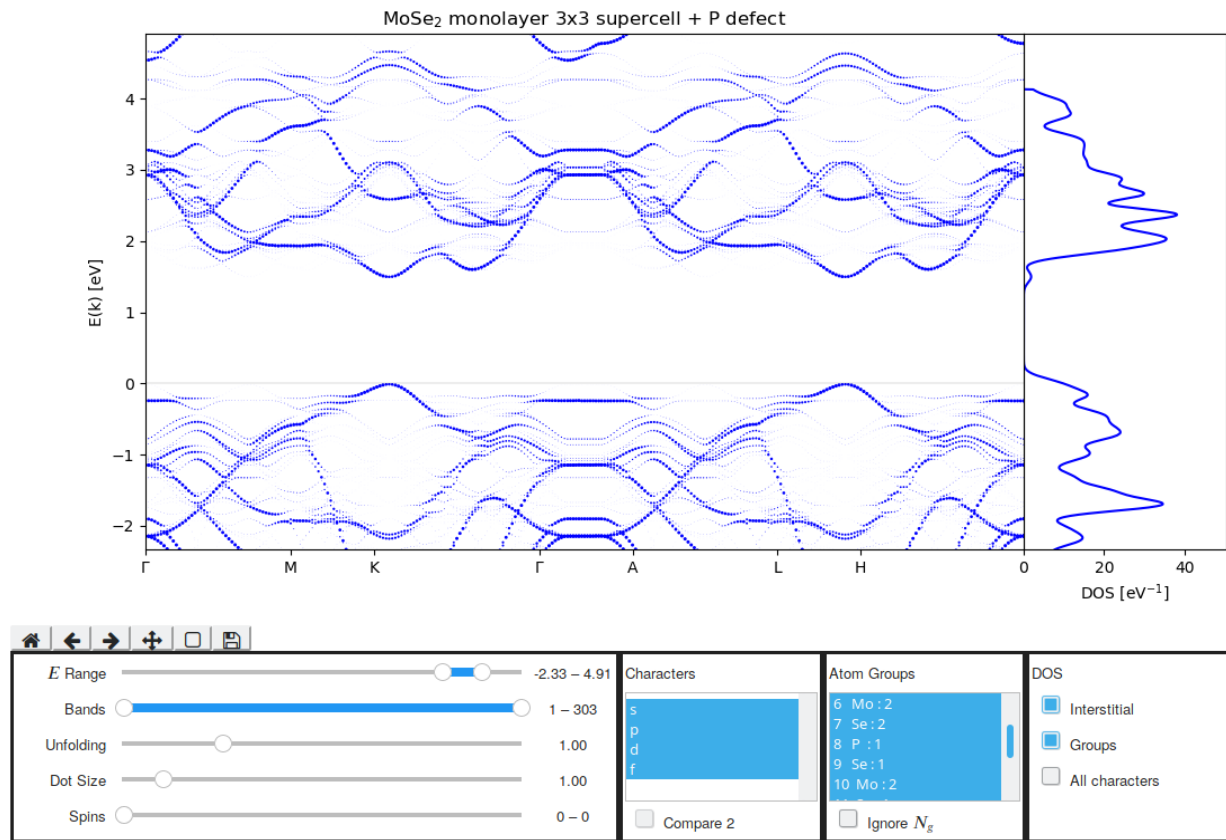


Figure 5.3: Bandstructure of a MoSe_2 monolayer using the default settings of the web frontend

Desktop Frontend: Co Crystal

Figure 5.5 shows the visualization of a band structure calculation of a 3 dimensional Co crystal with settings of dual spin, one atom group(since it is the only crystal in the cell structure) with all characters and whole band selected with exponent being at 0.0. We can see in the figure that band plot of any spin intersecting with the Fermi energy level. This means that many electrons excite to valance band which is a characteristic of a conductor. Hence the Co Crystal can be considered and concluded as conductor material.

One more physical property of Co crystal can be determined by considering and reading the density of state(DOS) files and plotting the density of state along with the band plots. In the figure 5.5, on the right one can see the Density of states of both the spins. If the area covered under the density is considered, it is clear that under the Fermi energy, electrons of one spin covers more area than other spin in the density of states plot. This means there are more electrons with positive spin under the Fermi energy. This is the character exhibited by the Ferro magnetic substance, so it be considered and concluded that this Co crystal is a Ferro magnetic.

There are two properties which are concluded by observing the band-dos plot but depending on

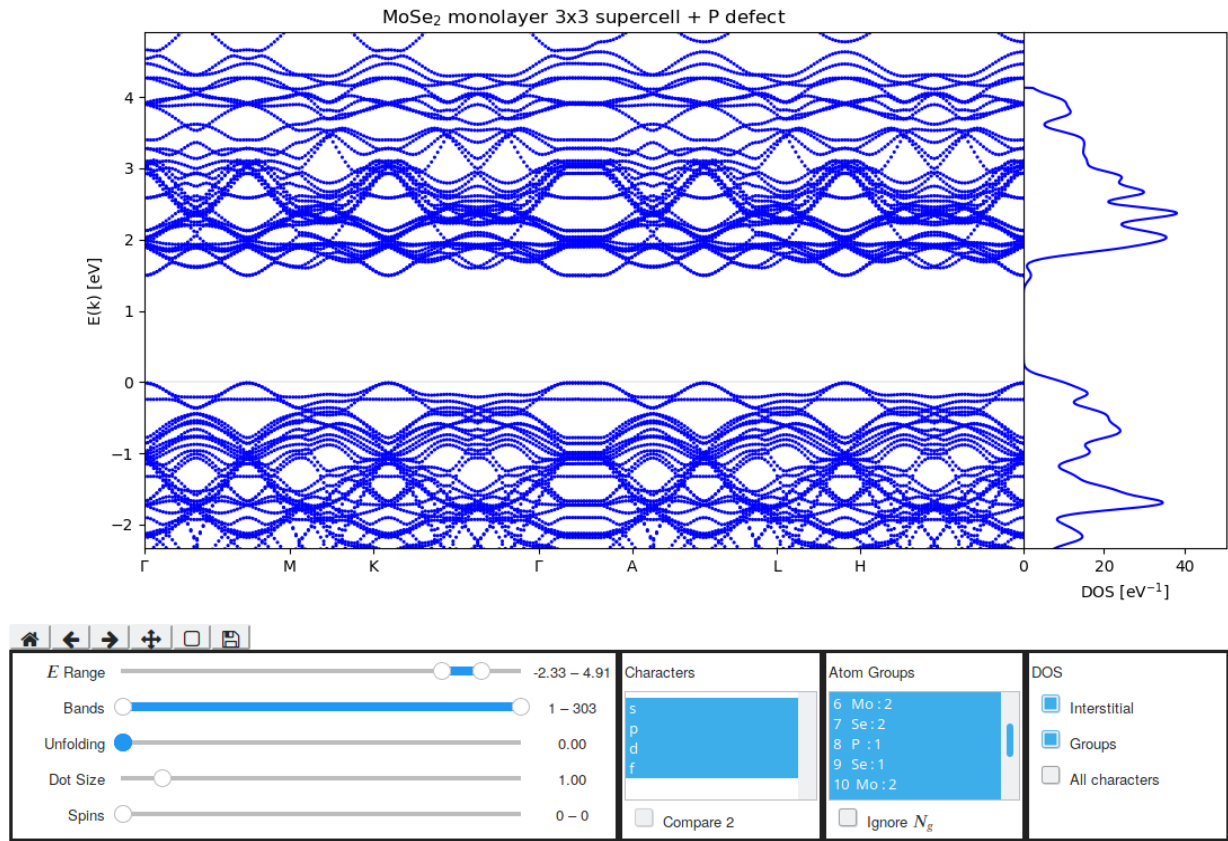


Figure 5.4: Bandstructure of a MoSe₂ monolayer without unfolding weights

the material and band-dos plot many properties can be extracted and concluded.

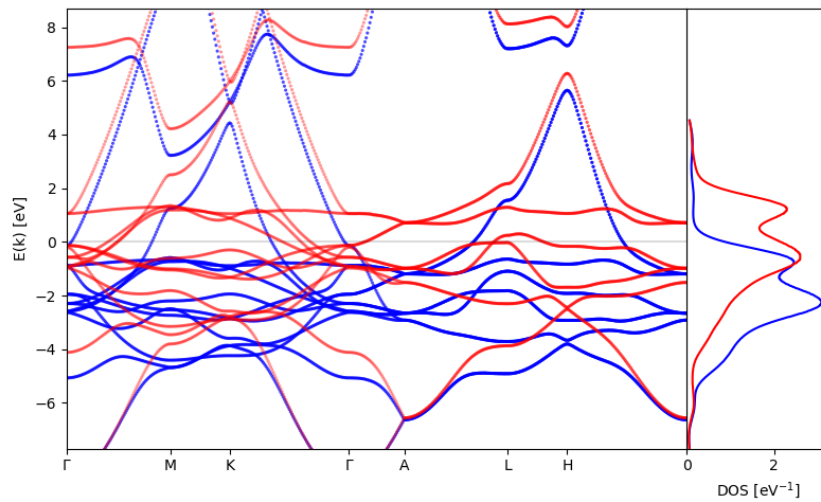


Figure 5.5: Bandstructure of a Co using the above settings of the Desktop frontend

Derived physical Quantities: Differentiation

The kind of datasets handled in the scope of the project did not lend themselves readily to applications of automatic differentiation techniques. Nevertheless, it is possible to derive meaningful physical quantities from the band structure using numerical differentiation techniques.

The effective mass m^* represents the mass, that an electron appears to have due to the inter-atomic forces in the crystal. At every \mathbf{k}_0 , where $E(\mathbf{k})$ has a local extremum, $E(\mathbf{k})$ can be expanded in a Taylor series with a vanishing first order term $E(\mathbf{k}) = E_0 + \frac{\partial E(\mathbf{k})}{\partial \mathbf{k}} \cdot (\mathbf{k} - \mathbf{k}_0)^2$. Comparing this to the dispersion relation of a free electron $E(\mathbf{k})_{free} = \frac{\hbar^2 \mathbf{k}^2}{2m_e}$ motivates the general definition

$$m_{k_i, k_j}^* = \hbar^2 \left(\frac{\partial^2 E(\mathbf{k})}{\partial k_i \partial k_j} \right)^{-1} \quad (5.1)$$

Since $\frac{\partial^2 E(\mathbf{k})}{\partial k_i \partial k_j}$ depends on the direction of the partial derivatives, m_{k_i, k_j}^* is a tensor. Because the band structure files only contain a discrete sampled path in the Brillouin zone, only the derivatives that correspond to the direction from one high-symmetry point to the next can be computed. We are only interested in the diagonal terms of $m^*|_{i,j}$.

A second potentially interesting quantity is the group velocity $v_G(\mathbf{k})$ associated to each band. The group velocity $v_G(\mathbf{k})$ at the Fermi energy $E = E_F$ is called Fermi velocity.

$$v_G(\mathbf{k})_{k_i} = \frac{1}{\hbar} \frac{\partial E(\mathbf{k})}{\partial k_i} \quad (5.2)$$

Differentiation

Since the k -mesh in DFT calculations is potentially very sparse, low order finite difference schemes are not expected to work well. Alternatively, one way to exploit all data points efficiently is to use fast Fourier transform methods, which are equivalent to the derivation of a truncated Fourier series. This method is expected to be well-suited for the problem since the graph of the band structure $(k, E(|k|))$ with $k \in \{-\Gamma, \dots, H, \dots, \Gamma\}$ is periodic, where Γ and H are arbitrary representatives of the high symmetry points. This periodicity in the reciprocal space is a direct consequence of the periodicity of the crystal.

In Fourier space, spatial derivatives transform into multiplications, which can easily be shown by partial integration.

$$f^{(n)}(x) = \mathcal{F}^{-1}((ik)^n \mathcal{F}(f(x))) \quad (5.3)$$

To test the FFT differentiation method, a band was selected that did not have intersections

with other bands within the interval between two high symmetry points and a stationary point at the high symmetry points. Then a resolution study was done to investigate the impact of the number of points within the interval.

The comparison between the FFT and a first-order central difference approximation of the second derivative (FD) is shown in Fig. 5.6, where different k-mesh resolutions are compared to a derivative that is almost fully converged. The comparison indicates that for small N, the error of the FFT method is significantly smaller than the error of the FD derivative. This is especially striking in the vicinity of the high symmetry points, where the error of the differentiation is required to be small in order to get good approximations for m^* , which is only meaningful close to the high symmetry points.

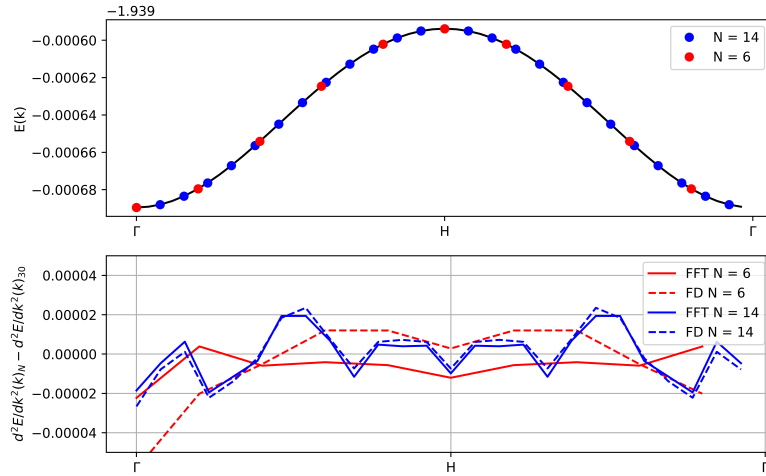


Figure 5.6: Comparison between FFT differentiation method and central finite difference method

When using more points, the difference between both approximation schemes is not significant. It is interesting to note, that the FFT method does not work anymore in the limit of extremely many points. In this case the derivative is dominated by Gibbs oscillations. These are most likely caused by small discontinuities in $E(k)$, due to basis changes inside the DFT computation.

At the current state the question remains to be answered whether the computation of m^* and v_g is useful. The number of bands, which the computation can be applied to is extremely limited. Since the bands in the data files are not labeled according to their corresponding eigenfunction but sorted by value, there are discontinuities at each point, where two bands intersect. This problem might be solved in the future.

Chapter 6

Conclusion

Outlook

- A package to developed to visualize DFT data in an appropriate way. The data is read, extracted and visualized using the package.
- An API and GUI are developed which are easy to use, easy to get the visualization output and visually appealing.
- The API is used and studies data to external physical feature such as effective mass (m), Fermi Velocity(v_{fermi}) using numerical differentiation techniques which is helpful in analyzing materials characteristics.
- Fluer output data are now easily accesible to non-experts since processing and visualization done by this API.
- Integrating of API and GUI into AiiDA workflow makes it easier for anyone to access this and understand the simulation much faster.
- Features extracted through the simulation data can also be saved for further research and analysis.
- A directly executable GUI which makes it easier to use in any computer just by copying the GUI.

TODO Module design:

- Preprocessor module: dataset dependencies for new recipes have to be explicitly stated. This could also be automatically resolved by type inspection.
- Frontends:
 - PyViz Param [\[Ste+19\]](#) makes it possible to decouple the formal description of a particular GUI from the GUI library used. This would serve to separate interface and

implementation like it is done in the preprocessor and visualization submodules.

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