# **NMRPy Documentation**

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## Contents

| 1  | Introduction   | 3                                     |  |  |  |  |  |
|----|--|---------------------------------------|--|--|--|--|--|
| 2  | Installation2.1Virtual environments2.2Pip install2.3Testing the installation2.4Working with NMRPy2.5Documentation  | <b>5</b><br>5<br>6<br>6<br>7          |  |  |  |  |  |
| 3  | Quickstart Tutorial3.1Importing3.2Apodisation and Fourier-transformation3.3Phase-correction3.4Peak-picking3.5Deconvolution3.6Saving / Loading3.7Full tutorial script | 9<br>10<br>11<br>14<br>16<br>19<br>19 |  |  |  |  |  |
| 4  | Basic Data Objects   | 23                                    |  |  |  |  |  |
| 5  | Plotting Objects   | 31                                    |  |  |  |  |  |
| 6  | 5 Indices and tables   |                                       |  |  |  |  |  |
| Ру | Python Module Index  |                                       |  |  |  |  |  |
| In | Index  |                                       |  |  |  |  |  |

Contents:

Introduction

NMRPy is a Python3 module for the processing and analysis of NMR spectra. The functionality of NMRPy is structured to make the analysis of arrayed NMR spectra more intuitive.

## Installation

The following are some general guidelines for installing NMRPy, and are by no means the only way to install a Python package. First be sure to have Python 3 and pip installed. Pip is a useful Python package management system.

Note: NMRPy will not work using Python 2.

On Debian and Ubuntu-like systems these can be installed with the following terminal commands:

```
$ sudo apt install python3
$ sudo apt install python-pip
```

On Windows, the CPython download from https://www.python.org/ comes pre-installed with pip.

The Anaconda Distribution, which is available for Windows, macOS and Linux, comes pre-installed with pip as well as most of the other dependencies required for NMRPy.

## 2.1 Virtual environments

Virtual environments are a great way to keep package dependencies separate from your system files. There are several options for setting up your working environment. We will use virtualenvwrapper, which works out of the box on Linux and MacOS. On Windows, virtualenvwrapper can be used under an MSYS environment in a native Windows Python installation. Alternatively, you can use virtualenvwrapper-win. This will take care of managing your virtual environments by maintaining a separate Python *site-directory* for you.

Install virtualenvwrapper using pip. On Linux and MacOS:

```
$ sudo pip install virtualenv
$ sudo pip install virtualenvwrapper
```

On Windows in a Python command prompt:

```
pip install virtualenv
pip install virtualenvwrapper-win
```

Make a new virtual environment for working with NMRPy (e.g. nmr), and specify that it use Python 3 (we used Python 3.7):

\$ mkvirtualenv -p python3.7 nmr

The new virtual environment will be activated automatically, and this will be indicated in the shell prompt. E.g.:

(nmr) \$

If you are not yet familiar with virtual environments we recommend you survey the basic commands (https://virtualenvwrapper.readthedocs.io/en/latest/) before continuing.

### 2.2 Pip install

The NMRPy code and its dependencies can be installed directly from PyPI into a virtual environment (if you are currently using one) using pip.

\$ pip install nmrpy

## 2.3 Testing the installation

Various tests are provided to test aspects of the NMRPy functionality within the unittest framework. The tests should be run from a terminal and can be invoked with nmrpy.test() after importing the *nmrpy* module.

Only a specific subset of tests can be run by providing an additional argument:

When testing the plotting utilities, a number of matplotlib plots will appear. This tests that the peak and range selection widgets are working properly; the plot windows can be safely closed.

## 2.4 Working with NMRPy

Though the majority of NMRPy functionality can be used purely in a scripting context and executed by the Python interpreter, it will often need to be used interactively. We suggest two ways to do this:

### 2.4.1 IPython

IPython is an interactive Python shell with some useful functionalities like tab-completion. This has been installed by default with NMRPy and can be launched from the command line with:

\$ ipython

### 2.4.2 Jupyter Notebook

For those who prefer a "notebook"-like experience, the Jupyter Notebook may be more appropriate. It has also been installed by default with NMRPy and can be launched with:

```
$ jupyter-notebook
```

If you will be using the peak-picking and range-selection widgets from within the Jupyter Notebook, this requires the jupyter-matplotlib extension (*ipympl*). This is installed automatically but the extension needs to be activated at the beginning of every notebook thus:

In [1]: %matplotlib ipympl

## 2.5 Documentation

Online documentation is available at https://nmrpy.readthedocs.io. The documentation is also distributed in PDF format in the docs subfolder of the nmrpy folder in site-packages where the package is installed.

The docs folder also contains an example Jupyter notebook (quickstart\_tutorial.ipynb) that mirrors the *Quickstart Tutorial*.

## **Quickstart Tutorial**

This is a "quickstart" tutorial for NMRPy in which an Agilent (Varian) NMR dataset will be processed. The following topics are explored:

- Importing
- Apodisation and Fourier-transformation
- Phase-correction
- Peak-picking
- Deconvolution
- Saving / Loading
- Full tutorial script

This tutorial will use the test data in the nmrpy install directory:

```
nmrpy/tests/test_data/test1.fid
```

The dataset consists of a time array of spectra of the phosphoglucose isomerase reaction:

*fructose-6-phosphate -> glucose-6-phosphate* 

An example Jupyter notebook is provided in the docs subdirectory of the nmrpy install directory, which mirrors this Quickstart Tutorial.

```
nmrpy/docs/quickstart_tutorial.ipynb
```

## 3.1 Importing

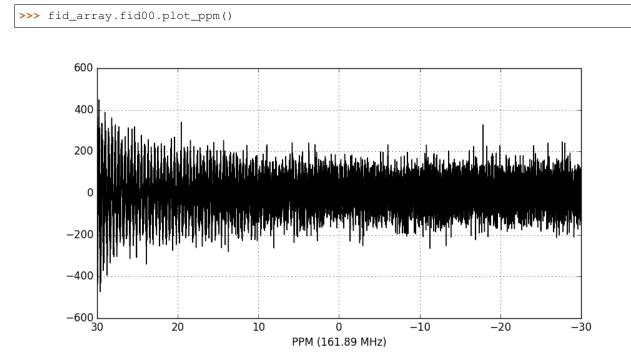
The basic NMR project object used in NMRPy is the *FidArray*, which consists of a set of *Fid* objects, each representing a single spectrum in an array of spectra.

The simplest way to instantiate an *FidArray* is by using the *from\_path()* method, and specifying the path of the *.fid* directory:

You will notice that the fid\_array object is instantiated and now owns several attributes, most of which are of the form fidXX where XX is a number starting at 00. These are the individual arrayed *Fid* objects.

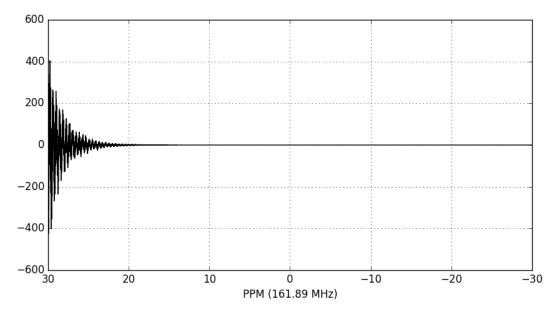
## 3.2 Apodisation and Fourier-transformation

To quickly visualise the imported data, we can use the plotting functions owned by each *Fid* instance. This will not display the imaginary portion of the data:



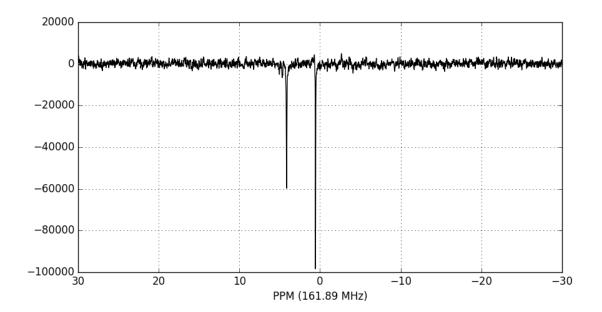
We now perform apodisation of the FIDs using the default value of 5 Hz, and visualise the result:

```
>>> fid_array.emhz_fids()
>>> fid_array.fid00.plot_ppm()
```



Finally, we Fourier-transform the data into the frequency domain:

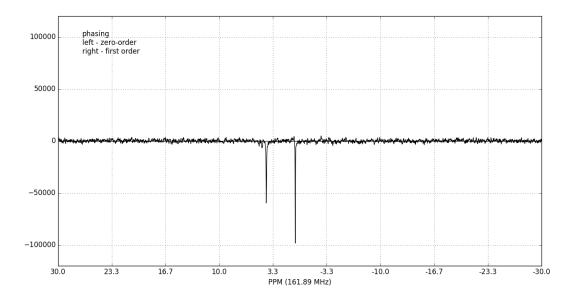
```
>>> fid_array.ft_fids()
>>> fid_array.fid00.plot_ppm()
```



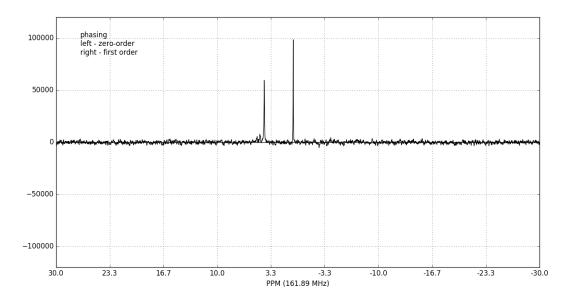
## 3.3 Phase-correction

It is clear from the data visualisation that at this stage the spectra require phase-correction. NMRPy provides a number of GUI widgets for manual processing of data. In this case we will use the *phaser()* method on fid00:

>>> fid\_array.fid00.phaser()



Dragging with the left mouse button and right mouse button will apply zero- and first-order phase-correction, respectively.

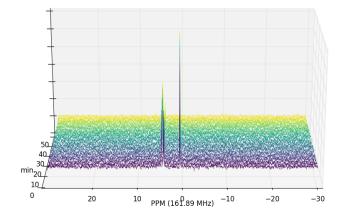


Alternatively, automatic phase-correction can be applied at either the *FidArray* or *Fid* level. We will apply it to the whole array:

```
>>> fid_array.phase_correct_fids()
```

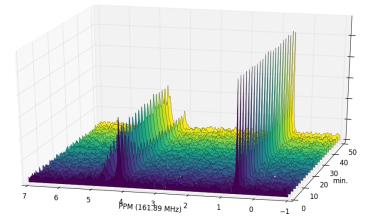
And plot an array of the phase-corrected data:

>>> fid\_array.plot\_array()



Zooming in on the relevant peaks, changing the view perspective, and filling the spectra produces a more interesting plot:

| >>> fid_array.plot_array(upper_ppm=7, lower_ppm=-1, filled= <b>True</b> , | azim=-76, elev=23) |
|---|--------------------|
|---|--------------------|



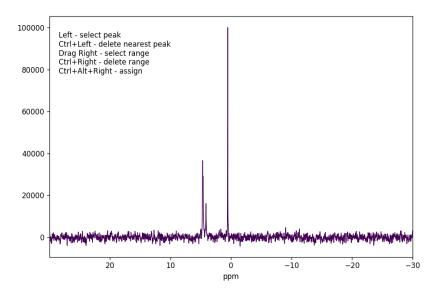
At this stage it is useful to discard the imaginary component of our data, and possibly normalise the data (by the maximum data value amongst the *Fid* objects):

>>> fid\_array.real\_fids()
>>> fid\_array.norm\_fids()

## 3.4 Peak-picking

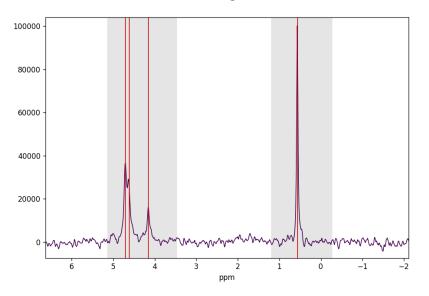
To begin the process of integrating peaks by deconvolution, we will need to pick some peaks. The *peaks* attribute of a *Fid* is an array of peak positions, and *ranges* is an array of range boundaries. These two objects are used in deconvolution to integrate the data by fitting Lorentzian/Gaussian peak shapes to the spectra. *peaks* and *ranges* may be specified programatically, or picked using the interactive GUI widget:

>>> fid\_array.peakpicker(fid\_number=10)



Peak and range selector

Left-clicking specifies a peak selection with a vertical red line. Dragging with a right-click specifies a range to fit independently with a grey rectangle:



Peak and range selector

Inadvertent wrongly selected peaks can be deleted with Ctrl+left-click; wrongly selected ranges can be deleted with Ctrl+right-click. Once you are done selecting peaks and ranges, these need to be assigned to the *FidArray*; this is

achieved with a Ctrl+Alt+right-click.

Ranges divide the data into smaller portions, which significantly speeds up the process of fitting of peakshapes to the data. Range-specification also prevents incorrect peaks from being fitted by the fitting algorithm.

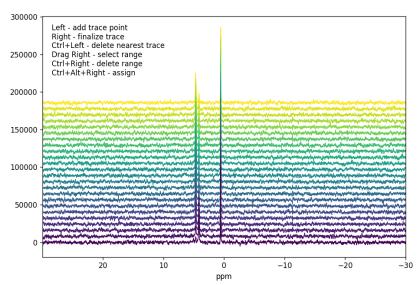
Having used the *peakpicker()* FidArray method (as opposed to the *peakpicker()* on each individual Fid instance), the peak and range selections have now been assigned to each Fid in the array:

```
>>> print(fid_array.fid00.peaks)
[ 4.73 4.63 4.15 0.55]
>>> print(fid_array.fid00.ranges)
[[ 5.92 3.24]
[ 1.19 -0.01]]
```

### 3.4.1 Peak-picking trace selector

Sometimes peaks are subject to drift so that the chemical shift changes over time; this can happen, e.g., when the pH of the reaction mixture changes as the reaction proceeds. NMRPy offers a convenient trace selector, with which the drift of the peaks can be traced over time and the chemical shift selected accordingly as appropriate for the particular Fid.

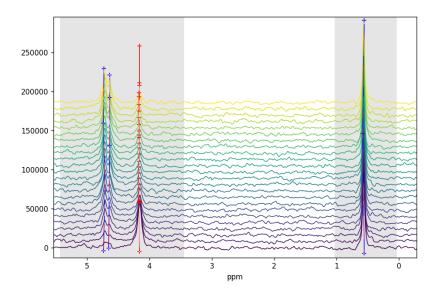
>>> fid\_array.peakpicker\_traces(voff=0.08)



### Peak and range trace selector

As for the *peakpicker()*, ranges are selected by dragging the right mouse button and can be deleted with Ctrl+right-click. A peak trace is initiated by left-clicking below the peak underneath the first Fid in the series. This selects a point and anchors the trace line, which is displayed in red as the mouse is moved. The trace will attempt to follow the highest peak. Further trace points can be added by repeated left-clicking, thus tracing the peak through the individual Fids in the series. It is not necessary to add an anchor point for every Fid, only when the trace needs to change direction. Once the trace has traversed all the Fids, select a final trace point (left-click) and then finalize the trace with a right-click. The trace will change colour from red to blue to indicate that it has been finalized.

Additional peaks can then be selected by initiating a new trace. Wrongly selected traces can be deleted by Ctrl+leftclick at the bottom of the trace that should be removed. Note that the interactive buttons on the matplotlib toolbar for the figure can be used to zoom and pan into a region of interest of the spectra.



#### Peak and range trace selector

As previously, peaks and ranges need to be assigned to the *FidArray* with Ctrl+Alt+right-click. As can be seen below, the individual peaks have different chemical shifts for the different Fids, although the drift in these spectra is not significant so that *peakpicker\_traces()* need not have been used and *peakpicker()* would have been sufficient. This is merely for illustrative purposes.

```
>>> print(p.fid00.peaks)
[4.73311164 4.65010807 0.55783899 4.15787759]
>>> print(p.fid10.peaks)
[4.71187817 4.6404565 0.5713512 4.16366854]
>>> print(p.fid20.peaks)
[4.73311164 4.63466555 0.57907246 4.16366854]
```

## 3.5 Deconvolution

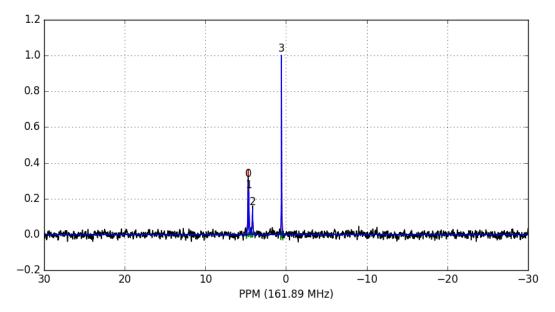
Individual Fid objects can be deconvoluted with deconv(). FidArray objects can be deconvoluted with deconv\_fids(). By default this is a multiprocessed method (*mp=True*), which will fit pure Lorentzian lineshapes (*frac\_gauss=0.0*) to the *peaks* and *ranges* specified in each *Fid*.

We shall fit the whole array at once:

>>> fid\_array.deconv\_fids()

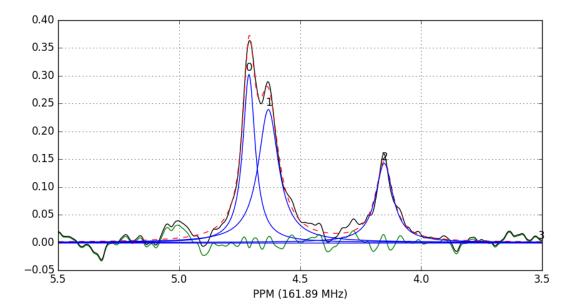
And visualise the deconvoluted spectra:

```
>>> fid_array.fid10.plot_deconv()
```



Zooming-in to a set of peaks makes clear the fitting result:

>>> fid\_array.fid10.plot\_deconv(upper\_ppm=5.5, lower\_ppm=3.5)
>>> fid\_array.fid10.plot\_deconv(upper\_ppm=0.9, lower\_ppm=0.2)



In this case, peaks 0 and 1 belong to glucose-6-phosphate, peak 2 belongs to fructose-6-phosphate, and peak 3 belongs to triethyl-phosphate.

We can view the deconvolution result for the whole array using *plot\_deconv\_array()*. Fitted peaks appear in red:

```
>>> fid_array.plot_deconv_array(upper_ppm=6, lower_ppm=3)
```

Peak integrals of the entire FidArray are stored in deconvoluted\_integrals, or in each individual Fid as

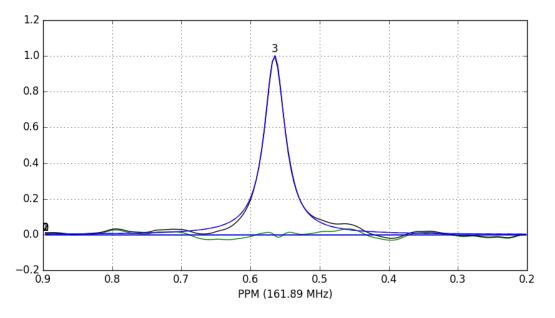
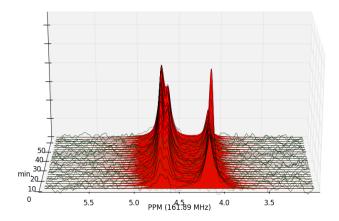


Fig. 1: The lines are colour-coded according to:

- *Blue*: individual peak shapes (and peak numbers above);
- Black: original data;
- *Red*: summed peak shapes;
- Green: residual (original data summed peakshapes).



#### deconvoluted\_integrals.

We could easily plot the species integrals using the following code:

```
from matplotlib import pyplot as plt
integrals = fid_array.deconvoluted_integrals.transpose()
g6p = integrals[0] + integrals[1]
f6p = integrals[2]
tep = integrals[3]
#scale species by internal standard tep (5 mM)
q6p = 5.0 \times q6p/tep.mean()
f6p = 5.0 \star f6p/tep.mean()
tep = 5.0 \star tep/tep.mean()
species = {'g6p': g6p,
           'f6p': f6p,
           'tep': tep}
fig = plt.figure()
ax = fig.add_subplot(111)
for k, v in species.items():
    ax.plot(fid_array.t, v, label=k)
ax.set_xlabel('min')
ax.set_ylabel('mM')
ax.legend(loc=0, frameon=False)
plt.show()
```

## 3.6 Saving / Loading

The current state of any FidArray object can be saved to file using the save\_to\_file() method:

```
>>> fid_array.save_to_file(filename='fidarray.nmrpy')
```

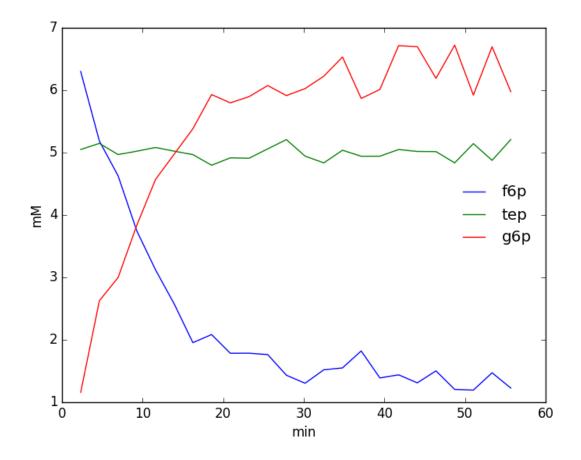
And reloaded using from\_path():

>>> fid\_array = nmrpy.data\_objects.FidArray.from\_path(fid\_path='fidarray.nmrpy')

## 3.7 Full tutorial script

The full script for the quickstart tutorial:

(continues on next page)



```
(continued from previous page)
```

```
fid_array.ft_fids()
#fid_array.fid00.plot_ppm()
#fid_array.fid00.phaser()
fid_array.phase_correct_fids()
#fid_array.fid00.plot_ppm()
fid_array.real_fids()
fid_array.norm_fids()
#fid_array.plot_array()
#fid_array.plot_array(upper_ppm=7, lower_ppm=-1, filled=True, azim=-76, elev=23)
peaks = [ 4.73, 4.63, 4.15, 0.55]
ranges = [[5.92, 3.24], [1.19, -0.01]]
for fid in fid_array.get_fids():
    fid.peaks = peaks
    fid.ranges = ranges
fid_array.deconv_fids()
#fid_array.fid10.plot_deconv(upper_ppm=5.5, lower_ppm=3.5)
#fid_array.fid10.plot_deconv(upper_ppm=0.9, lower_ppm=0.2)
#fid_array.plot_deconv_array(upper_ppm=6, lower_ppm=3)
integrals = fid_array.deconvoluted_integrals.transpose()
g6p = integrals[0] + integrals[1]
f6p = integrals[2]
tep = integrals[3]
#scale species by internal standard tep at 5 mM
g6p = 5.0 \times g6p/tep.mean()
f6p = 5.0 \star f6p/tep.mean()
tep = 5.0 * tep / tep . mean()
species = {'g6p': g6p,
           'f6p': f6p,
           'tep': tep}
fig = plt.figure()
ax = fig.add_subplot(111)
for k, v in species.items():
   ax.plot(fid_array.t, v, label=k)
ax.set_xlabel('min')
ax.set_ylabel('mM')
ax.legend(loc=0, frameon=False)
plt.show()
#fid_array.save_to_file(filename='fidarray.nmrpy')
#fid array = nmrpy.data objects.FidArray.from path(fid path='fidarray.nmrpy')
```

## **Basic Data Objects**

#### class nmrpy.data\_objects.Fid(\*args, \*\*kwargs)

The basic FID (Free Induction Decay) class contains all the data for a single spectrum (data), and the necessary methods to process these data.

#### baseline\_correct(deg=2)

Perform baseline correction by fitting specified baseline points (stored in \_bl\_ppm) with polynomial of specified degree (stored in \_bl\_ppm) and subtract this polynomial from *data*.

Parameters deg - degree of fitted polynomial

#### baseliner()

Instantiate a baseline-correction GUI widget. Right-click-dragging defines a range. Ctrl-Right click deletes previously selected range. Indices selected are stored in \_bl\_ppm, which is used for baseline-correction ().

#### data

The spectral data. This is the primary object upon which the processing and analysis functions work.

#### **deconv** (*method='leastsq'*, *frac\_gauss=0.0*)

Deconvolute data object by fitting a series of peaks to the spectrum. These peaks are generated using the parameters in *peaks*. *ranges* splits *data* up into smaller portions. This significantly speeds up deconvolution time.

#### Parameters

- **frac\_gauss** (0-1) determines the Gaussian fraction of the peaks. Setting this argument to None will fit this parameter as well.
- **method** The fitting method to use. Default is 'leastsq', the Levenberg-Marquardt algorithm, which is usually sufficient. Additional options include:

Nelder-Mead (nelder)

L-BFGS-B (l-bfgs-b)

Conjugate Gradient (cg)

Powell (powell)

Newton-CG (newton)

#### deconvoluted\_integrals

An array of integrals for each deconvoluted peak.

#### **emhz** (lb=5.0)

Apply exponential line-broadening to data array data.

Parameters 1b - degree of line-broadening in Hz.

#### classmethod from\_data(data)

Instantiate a new Fid object by providing a spectral data object as argument. Eg.

fid = Fid.from\_data(data)

#### ft()

Fourier Transform the data array data.

Calculates the Discrete Fourier Transform using the Fast Fourier Transform algorithm as implemented in NumPy (*Cooley, James W., and John W. Tukey, 1965, 'An algorithm for the machine calculation of complex Fourier series,' Math. Comput. 19: 297-301.*)

#### peakpick (thresh=0.1)

Attempt to automatically identify peaks. Picked peaks are assigned to peaks.

Parameters thresh – fractional threshold for peak-picking

#### peakpicker()

Instantiate a peak-picking GUI widget. Left-clicking selects a peak. Right-click-dragging defines a range. Ctrl-left click deletes nearest peak; ctrl-right click deletes range. Peaks are stored in *peaks*; ranges are stored in *ranges*: both are used for deconvolution (see *deconv()*).

#### peaks

Picked peaks for deconvolution of data.

#### phase\_correct (method='leastsq')

Automatically phase-correct *data* by minimising total absolute area.

**Parameters method** – The fitting method to use. Default is 'leastsq', the Levenberg-Marquardt algorithm, which is usually sufficient. Additional options include:

Nelder-Mead (nelder)

L-BFGS-B (l-bfgs-b)

Conjugate Gradient (cg)

Powell (powell)

Newton-CG (newton)

#### phaser()

Instantiate a phase-correction GUI widget which applies to data.

#### plot\_deconv(\*\*kwargs)

Plot data with deconvoluted peaks overlaid.

#### Parameters

- upper\_ppm upper spectral bound in ppm
- lower\_ppm lower spectral bound in ppm
- **lw** linewidth of plot
- colour colour of the plot

- peak\_colour colour of the deconvoluted peaks
- residual\_colour colour of the residual signal after subtracting deconvoluted peaks

plot\_ppm(\*\*kwargs)

Plot data.

#### Parameters

- upper\_ppm upper spectral bound in ppm
- lower\_ppm lower spectral bound in ppm
- **lw** linewidth of plot
- colour colour of the plot

#### **ps** (*p*0=0.0, *p*1=0.0)

Linear phase correction of data

#### Parameters

- **p0** Zero order phase in degrees
- **p1** First order phase in degrees

#### ranges

Picked ranges for deconvolution of *data*.

#### real()

Discard imaginary component of *data*.

#### zf()

Apply a single degree of zero-filling to data array *data*.

Note: extends data to double length by appending zeroes. This results in an artificially increased resolution once Fourier-transformed.

#### class nmrpy.data\_objects.FidArray(\*args, \*\*kwargs)

This object collects several *Fid* objects into an array, and it contains all the processing methods necessary for bulk processing of these FIDs. It should be considered the parent object for any project. The class methods *from\_path()* and *from\_data()* will instantiate a new *FidArray* object from a Varian/Bruker .fid path or an iterable of data respectively. Each *Fid* object in the array will appear as an attribute of *FidArray* with a unique ID of the form 'fidXX', where 'XX' is an increasing integer .

#### add\_fid(fid)

Add an Fid object to this FidArray, using a unique id.

#### **Parameters fid** – an *Fid* instance

```
add_fids (fids)
```

Add a list of *Fid* objects to this *FidArray*.

**Parameters fids** – a list of *Fid* instances

#### baseline\_correct\_fids(deg=2)

Apply baseline-correction to all Fid objects owned by this FidArray

**Parameters deg** – degree of the baseline polynomial (see baseline\_correct())

#### baseliner\_fids()

Instantiate a baseline-correction GUI widget. Right-click-dragging defines a range. Ctrl-Right click deletes previously selected range. Indices selected are stored in \_bl\_ppm, which is used for baseline-correction ().

#### data

An array of all data objects belonging to the Fid objects owned by this FidArray.

#### **deconv\_fids** (*mp=True*, *cpus=None*, *method='leastsq'*, *frac\_gauss=0.0*)

Apply deconvolution to all *Fid* objects owned by this *FidArray*, using the *peaks* and *ranges* attribute of each respective *Fid*.

#### Parameters

- method see phase\_correct ()
- **mp** parallelise the phasing process over multiple processors, significantly reduces computation time
- cpus defines number of CPUs to utilise if 'mp' is set to True, default is n-1 cores

#### deconvoluted\_integrals

Collected deconvoluted\_integrals

#### $\texttt{del\_fid}(\mathit{fid\_id})$

Delete an *Fid* object belonging to this *FidArray*, using a unique id.

Parameters fid\_id - a string id for an Fid

#### $emhz_fids$ (lb=5.0)

Apply line-broadening (apodisation) to all nmrpy.~data\_objects.Fid objects owned by this FidArray

Parameters 1b – degree of line-broadening in Hz.

#### classmethod from\_data(data)

Instantiate a new *FidArray* object from a 2D data set of spectral arrays.

Parameters data - a 2D data array

#### **classmethod from\_path** (*fid\_path='.', file\_format=None*) Instantiate a new *FidArray* object from a .fid directory.

#### **Parameters**

- fid\_path filepath to .fid directory
- file\_format 'varian' or 'bruker', usually unnecessary

#### ft\_fids (mp=True, cpus=None)

Fourier-transform all FIDs.

#### Parameters

- mp parallelise over multiple processors, significantly reducing computation time
- cpus defines number of CPUs to utilise if 'mp' is set to True

#### get\_fid(id)

Return an Fid object owned by this object, identified by unique ID. Eg.:

fid12 = fid\_array.get\_fid('fid12')

#### **Parameters** id – a string id for an *Fid*

#### get\_fids()

Return a list of all *Fid* objects owned by this *FidArray*.

#### get\_integrals\_from\_traces()

Returns a dictionary of integral values for all *Fid* objects calculated from trace dictionary *integral\_traces*.

#### get\_masked\_integrals()

After peakpicker\_traces() and deconv\_fids() this function returns a masked integral array.

#### integral\_traces

Returns the dictionary of integral traces generated by select\_integral\_traces().

#### norm\_fids()

Normalise FIDs by maximum data value in *data*.

#### **peakpicker** (*fid\_number=None*, *assign\_only\_to\_index=True*, *voff=0.02*)

Instantiate peak-picker widget for *data*, and apply selected *peaks* and *ranges* to all *Fid* objects owned by this *FidArray*. See *peakpicker()*.

#### **Parameters**

- **fid\_number** list or number, index of *Fid* to use for peak-picking. If None, data array is plotted.
- **assign\_only\_to\_index** if True, assigns selections only to *Fid* objects indexed by fid\_number, if False, assigns to all
- **voff** vertical offset for spectra

#### peakpicker\_traces(voff=0.02, lw=1)

Instantiates a widget to pick peaks and ranges employing a polygon shape (or 'trace'). This is useful for picking peaks that are subject to drift and peaks that appear (or disappear) during the course of an experiment.

#### **Parameters**

- **voff** vertical offset fraction (0.01)
- **lw** linewidth of plot (1)

#### phase\_correct\_fids (method='leastsq', mp=True, cpus=None)

Apply automatic phase-correction to all Fid objects owned by this FidArray

#### **Parameters**

- method see phase\_correct ()
- **mp** parallelise the phasing process over multiple processors, significantly reducing computation time
- cpus defines number of CPUs to utilise if 'mp' is set to True

#### plot\_array(\*\*kwargs)

Plot data.

#### **Parameters**

- upper\_index upper index of array (None)
- lower\_index lower index of array (None)
- upper\_ppm upper spectral bound in ppm (None)
- **lower\_ppm** lower spectral bound in ppm (None)
- lw linewidth of plot (0.5)
- azim starting azimuth of plot (-90)

- **elev** starting elevation of plot (40)
- **filled** True=filled vertices, False=lines (False)
- show\_zticks show labels on z axis (False)
- **labels** under development (None)
- **colour** plot spectra with colour spectrum, False=black (True)
- filename save plot to .pdf file (None)

#### plot\_deconv\_array(\*\*kwargs)

Plot all data with deconvoluted peaks overlaid.

#### Parameters

- upper\_index upper index of Fids to plot
- lower\_index lower index of Fids to plot
- upper\_ppm upper spectral bound in ppm
- **lower\_ppm** lower spectral bound in ppm
- data\_colour colour of the plotted data ('k')
- summed\_peak\_colour colour of the plotted summed peaks ('r')
- **residual\_colour** colour of the residual signal after subtracting deconvoluted peaks ('g')
- data\_filled fill state of the plotted data (False)
- **summed\_peak\_filled** fill state of the plotted summed peaks (True)
- **residual\_filled** fill state of the plotted residuals (False)
- **figsize** [x, y] size of plot ([15, 7.5])
- **lw** linewidth of plot (0.3)
- azim azimuth of 3D axes (-90)
- **elev** elevation of 3D axes (20)

#### **ps\_fids** (*p0*=0.0, *p1*=0.0)

Apply manual phase-correction to all Fid objects owned by this FidArray

#### Parameters

- **p0** Zero order phase in degrees
- **p1** First order phase in degrees

#### real\_fids()

Discard imaginary component of FID data sets.

#### save\_to\_file (filename=None)

Save FidArray object to file, including all objects owned.

Parameters filename – filename to save FidArray to

#### select\_integral\_traces(voff=0.02, lw=1)

Instantiate a trace-selection widget to identify deconvoluted peaks. This can be useful when data are subject to drift. Selected traces on the data array are translated into a set of nearest deconvoluted peaks, and saved in a dictionary: *integral\_traces*.

#### Parameters

- **voff** vertical offset fraction (0.01)
- **lw** linewidth of plot (1)
- t

An array of the acquisition time for each FID.

**zf\_fids**()

Zero-fill all Fid objects owned by this FidArray

## **Plotting Objects**

Interactive data-selection widget with lines and ranges. Lines and spans are saved as self.peaks, self.ranges.

class nmrpy.plotting.DataPeakSelector (fid, peaks=None, ranges=None, voff=0.001, lw=1, label=None, title=None)

Interactive data-selection widget with lines and ranges for a single Fid. Lines and spans are saved as self.peaks, self.ranges.

Interactive selector widget. can inherit from various mixins for functionality: Lineselectortion:LineSelectorMixinSpan selection:SpanSelectorMixinPolyselection:PolySelectorMixinPolySelectorMixinSpan selection:SpanSelectorMixinSpanSelectorMixin

This class is not intended to be used without inheriting at least one mixin.

class nmrpy.plotting.DataTraceRangeSelector(fid\_array, peaks=None, ranges=None, voff=0.001, lw=1, label=None)

Interactive data-selection widget with traces and ranges. Traces are saved as self.data\_traces (WRT data) and self.index\_traces (WRT index). Spans are saves as self.spans.

- class nmrpy.plotting.DataTraceSelector(fid\_array, extra\_data=None, extra\_data\_colour='b', voff=0.001, lw=1, label=None) Interactive data-selection widget with traces and ranges. Traces are saved as self.data\_traces (WRT data) and self.index traces (WRT index).
- **class** nmrpy.plotting.**FidArrayRangeSelector** (*fid\_array, ranges=None, y\_indices=None, voff=0.001, lw=1, title=None, label=None*) Interactive data-selection widget with ranges. Spans are saved as self.ranges.
- **class** nmrpy.plotting.**FidRangeSelector** (*fid*, *title=None*, *ranges=None*, *y\_indices=None*, *voff=0.001*, *lw=1*, *label=None*) Interactive data-selection widget with ranges. Spans are saved as self.ranges.

| class | nmrpy.plotting.IntegralDataSelector  |                      |                  |             | ata=None,<br>peaks=No                 |      |
|-------|--|----------------------|------------------|-------------|---------------------------------------|------|
|       |  | ranges=N<br>bel=None |                  | title=None, | voff=0.001,                           | la-  |
| class | nmrpy.plotting.LineSpanDataSelector  | tra_data_            | colour=          | ='k',       | peaks=No                              | one, |
|       |  | ranges=N<br>bel=None |                  | title=None, | voff=0.001,                           | la-  |
| class | nmrpy.plotting. <b>PeakTraceDataSelecto</b>  | tra_data             | _coloui<br>None, | r='k',      | lata=None,<br>peaks=Na<br>voff=0.001, | one, |
|       | nmrpy.plotting. <b>Phaser</b> ( <i>fid</i> )<br>nteractive phase-correction widget       |                      |                  |             |                                       |      |
|       | nmrpy.plotting. <b>Plot</b><br>Basic 'plot' class containing functions for various types | of plots.            |                  |             |                                       |      |
| class | <pre>nmrpy.plotting.SpanDataSelector(data,<br/>tra_d</pre>                               | -                    |                  |             | a=None,<br>e, ranges=No               |      |

title=None, voff=0.001, label=None)

Indices and tables

- genindex
- modindex
- search

Python Module Index

n

nmrpy.data\_objects,23
nmrpy.plotting,31

## Index

## Α

add\_fid() (nmrpy.data\_objects.FidArray method), 25 add fids() (nmrpy.data objects.FidArray method), 25

## В

(nmrpy.data\_objects.Fid baseline\_correct() method), 23 baseline\_correct\_fids() (*nm*rpy.data objects.FidArray method), 25 baseliner() (nmrpy.data\_objects.Fid method), 23 (nmrpy.data\_objects.FidArray baseliner\_fids() method), 25

## D

data (nmrpy.data\_objects.Fid attribute), 23 data (nmrpy.data\_objects.FidArray attribute), 25 DataPeakRangeSelector (class in nmrpy.plotting), 31 DataPeakSelector (class in nmrpy.plotting), 31 DataSelector (class in nmrpy.plotting), 31 DataTraceRangeSelector (class in nmrpy.plotting), 31 DataTraceSelector (class in nmrpy.plotting), 31 deconv() (nmrpy.data\_objects.Fid method), 23 deconv\_fids() (nmrpy.data\_objects.FidArray method), 26 deconvoluted\_integrals (nmrpy.data\_objects.Fid attribute), 24 deconvoluted\_integrals (nmrpy.data\_objects.FidArray attribute), 26 del\_fid() (nmrpy.data\_objects.FidArray method), 26 F

emhz() (nmrpy.data\_objects.Fid method), 24 emhz\_fids() (nmrpy.data\_objects.FidArray method), 26

## F

Fid (class in nmrpy.data\_objects), 23

FidArray (class in nmrpy.data objects), 25 FidArrayRangeSelector (class in nmrpy.plotting), 31 FidRangeSelector (class in nmrpy.plotting), 31 from\_data() (nmrpy.data\_objects.Fid class method), 24 from\_data() (*nmrpy.data\_objects.FidArray* class method), 26 from\_path() (*nmrpy.data\_objects.FidArray* class method), 26 ft() (nmrpy.data\_objects.Fid method), 24 ft\_fids() (nmrpy.data\_objects.FidArray method), 26

## G

get\_fid() (nmrpy.data\_objects.FidArray method), 26 get\_fids() (nmrpy.data\_objects.FidArray method), 26 get\_integrals\_from\_traces() (nmrpy.data\_objects.FidArray method), 26 get\_masked\_integrals() (nmrpy.data\_objects.FidArray method), 27

integral\_traces (nmrpy.data\_objects.FidArray attribute), 27 IntegralDataSelector (class in nmrpy.plotting), 32

### L

LineSpanDataSelector (class in nmrpy.plotting), 32

## Ν

nmrpy.data\_objects (module), 23 nmrpy.plotting (module), 31 norm\_fids() (nmrpy.data\_objects.FidArray method), 27

## Ρ

peakpick() (nmrpy.data\_objects.Fid method), 24

peakpicker() (nmrpy.data\_objects.Fid method), 24 (nmrpy.data\_objects.FidArray peakpicker() method), 27 peakpicker\_traces() (*nm*rpy.data\_objects.FidArray method), 27 peaks (nmrpy.data\_objects.Fid attribute), 24 PeakTraceDataSelector (class in nmrpy.plotting), 32 phase\_correct() (nmrpy.data\_objects.Fid method), 24 phase\_correct\_fids() (nmrpy.data\_objects.FidArray method), 27 Phaser (class in nmrpy.plotting), 32 phaser() (nmrpy.data\_objects.Fid method), 24 Plot (class in nmrpy.plotting), 32 plot\_array() (nmrpy.data\_objects.FidArray method), 27 plot\_deconv() (nmrpy.data\_objects.Fid method), 24 plot\_deconv\_array() (*nm*rpy.data\_objects.FidArray method), 28 plot\_ppm() (nmrpy.data\_objects.Fid method), 25 ps() (nmrpy.data\_objects.Fid method), 25 ps\_fids() (nmrpy.data\_objects.FidArray method), 28

## R

## S

## Т

t (nmrpy.data\_objects.FidArray attribute), 29

## Ζ

zf() (nmrpy.data\_objects.Fid method), 25
zf\_fids() (nmrpy.data\_objects.FidArray method), 29