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# Contents

1	Introduction			
<b>2</b>	Us	er guide		
	2.1	Initialize the package		
		2.1.1 Extra variables		
	2.2	Processing of a 'raw'1D spectrum		
		2.2.1 The class pSpectrum_1D		
	2.3	Processing of a 'raw'2D spectrum		
	2.0	2.3.1 Computing projections		
	2.4	Simulating data		
		2.4.1 Simulate 1D data		
		2.4.2 Simulate 2D data		
	2.5	The Pseudo_2D class		
	_	_		
3	Lis	t of modules and functions		
	3.1	MISC package 13		
		3.1.1 misc.SNR		
		3.1.2 misc.SNR_2D		
		3.1.3 misc.avg_antidiag		
		3.1.4 misc.binomial_triangle		
		3.1.5 misc.calcres		
		3.1.6 misc.cmap2list		
		3.1.7 misc.edit_checkboxes		
		3.1.8 misc.find_nearest		
		3.1.9 misc.freq2ppm		
		3.1.10 misc.get_trace		
		3.1.11 misc.get_ylim		
		3.1.12 misc.hankel		
		3.1.13 misc.hz2pt		
		3.1.14 misc.in2px		
		3.1.15 misc.makeacqus_1D		
		3.1.16 misc.makeacqus_1D_oxford		
		3.1.17 misc.makeacqus_2D		
		3.1.18 misc.mathformat		
		3.1.19 misc.molfrac		
		3.1.20 misc.noise_std		
		3.1.21 misc.nuc_format		
		3.1.22 misc.polyn		
		3.1.23 misc.ppm2freq		
		3.1.24 misc.ppmfind		
		3.1.25 misc.pretty_scale		

		misc.print_dict
		misc.print_list
		misc.procpar
		misc.px2in
		misc.readlistfile
	3.1.31	misc.select_for_integration
	3.1.32	misc.select_traces
	3.1.33	misc.set_fontsizes
	3.1.34	misc.set_ylim
	3.1.35	misc.show_cmap
	3.1.36	misc.split_acqus_2D
	3.1.37	misc.split_procs_2D
	3.1.38	misc.trim_data
		misc.trim_data_2D
		misc.unhankel
	3.1.41	misc.write_acqus_1D
		misc.write_acqus_2D
		misc.write help
	3.1.44	misc.write_ser56
0.0	DRC	
3.2		CESSING package 57
	3.2.1	processing. Cadzow
	3.2.2	processing.Cadzow_2D
	3.2.3	processing.EAE
	3.2.4	processing.LRD
	3.2.5	processing.MCR
	3.2.6	processing.MCR_ALS
	3.2.7	processing.MCR_unpack
	3.2.8	processing.SIMPLISMA
	3.2.9	processing.acme
		processing.align
		processing.baseline_correction
		processing.calc_nc
		processing.calibration
		processing.convdta
		processing.em
		processing.fp
		processing.ft
		processing.gm
		processing.gmb
	3.2.20	processing.ift
	3.2.21	processing.integral
	3.2.22	processing.integral_2D
	3.2.23	processing.interactive_basl_windows
	3.2.24	processing.interactive_echo_param
	3.2.25	processing.interactive_fp
	3.2.26	processing.interactive_phase_1D
	3.2.27	processing.interactive_phase_2D
	3.2.28	processing.interactive_qfil
	3.2.29	processing.interactive_xfb
	3.2.30	processing.inv_fp
	3.2.31	processing.inv_xfb

		processing.iterCadzow
	3.2.33	processing.load_baseline
		processing.make polynomion baseline
		processing.make_scale
		processing.new MCR 92
		processing.new_MCR_ALS
		processing.pknl
		processing.ps
	3.2.40	processing.qfil
		processing.qpol
		processing.qsin
		processing.quad
		processing.repack_2D
		processing.rev
		processing.rpbc
		processing.sin
		processing.split_echo_train
		processing.spate_ceno_train
		processing.sum_echo_train
		processing.tabula rasa
		processing.td eff
		processing.tp hyper
		processing.uppack 2D
		processing.whittaker_smoother
		processing.write_basl_info
		processing.zf
3.3	FIG	URES package
	3.3.1	figures.ax $1\bar{D}$
	3.3.2	$figures.ax2D \dots \dots$
	3.3.3	figures.ax_heatmap
	3.3.4	figures.dotmd
	3.3.5	figures.dotmd_2D
	3.3.6	figures.figure1D
	3.3.7	figures.figure1D multi
	3.3.8	figures.figure2D
	3.3.9	figures.figure2D multi
	3.3.10	figures.fitfigure
	3.3.11	figures.heatmap
		figures.plot fid
	3.3.13	figures.plot_fid_re
		figures.redraw contours
		figures.sns heatmap
		figures.stacked plot
0.4		l package
3.4		
	3.4.1	sim.calc_splitting
	3.4.2	sim.f_gaussian
	3.4.3	sim.f_lorentzian
	3.4.4	sim.f_pvoigt
	3.4.5	sim.gaussian_filter

	3.4.6	sim.load_sim_1D
	3.4.7	sim.load_sim_2D
	3.4.8	sim.mult_noise
	3.4.9	sim.multiplet
	3.4.10	sim.noisegen
	3.4.11	sim.sim_1D
	3.4.12	sim.sim_2D
		sim.t_2Dgaussian
		sim.t_2Dlorentzian
		sim.t_2Dpvoigt
		sim.t_2Dvoigt
		sim.t_gaussian
		sim.t_lorentzian
		sim.t_pvoigt
		sim.t_voigt
		sim.water7
3.5	FII	' package 160
	3.5.1	fit.CostFunc
	3.5.2	fit.LR
	3.5.3	fit.LSP
	3.5.4	fit.SINC_ObjFunc
	3.5.5	fit.SINC phase
	3.5.6	fit.Voigt_Fit
	3.5.7	fit.ax histogram
	3.5.8	fit.bin data
	3.5.9	fit.build_2D_sgn
	3.5.10	fit.build_baseline
	3.5.11	fit.calc_fit_lines
	3.5.12	fit.dic2mat
	3.5.13	fit.fit_int
	3.5.14	fit.gaussian_fit
	3.5.15	fit.gen_iguess
	3.5.16	fit.gen_iguess_2D
	3.5.17	fit.get_region
	3.5.18	fit.histogram
	3.5.19	fit.integrate
	3.5.20	fit.integrate_2D
		fit.interactive_smoothing
	3.5.22	fit.join_par
	3.5.23	fit.make_iguess
	3.5.24	fit.make_signal
		fit.peak_pick
	3.5.26	fit.polyn_basl
	3.5.27	fit.print_par
	3.5.28	fit.read_par
		fit.smooth_spl
	3.5.30	fit.test_residuals
	3.5.31	fit.voigt_fit
	3.5.32	fit.voigt_fit_2D
	3.5.33	fit.write_log
	3.5.34	fit.write_par

1		
	771	

3.6	SPE	ECTRA package 201
		Spectra.Pseudo 2D
	3.6.2	Spectra.Spectrum_1D
	3.6.3	Spectra.Spectrum_2D
	3.6.4	Spectra.pSpectrum_1D
	3.6.5	Spectra.pSpectrum 2D

## 1. Introduction

KLASSEZ is a python package written to handle 1D and 2D NMR data. The aim of the project is to provide a toolkit, consisting of 'black-box' functions organized in modules, that could be used to read, process and analyze such data in a flexible manner, so to adapt to the needs of the individual users. However, the open-source nature of the package grants the user the chance to open the lid of these black-boxes and understand the gears that stand behind the function call.

The development of the toolkit started with python 3.8 and therefore it is compatible with that version. Nevertheless, the use of python 3.10 is advised.

The key objects provided by KLASSEZ are the classes Spectrum\_1D and Spectrum\_2D, that are able to fulfil the aims of the package with a few lines of code. The classes are able to read both simulated (i.e. generated with a custom-made input file) and experimental datasets. The latter feature was tested with Bruker data after the removal of the digital filter (run command convdta in TopSpin), but should be compatible with other kind of spectrometers, thanks to the remarkable work made by J. J. Helmus and coworkers with their nmrglue package<sup>1</sup>. Either the FID or the spectrum processed with external solver can be read from KLASSEZ by using the classes Spectrum\_nD or pSpectrum\_nD, respectively.

The processing module, besides the classical functions used for the processing of NMR data (window functions, Fourier transform, etc.), includes denoising algorithms based on Multivariate Curve Resolution<sup>2</sup> and on Cadzow method<sup>3</sup>. Details are illustrated in the description of the functions.

Functions to show and analyze data in real time are provided, with dedicated GUIs. However, it is better to rely on the standalone functions, enclosed in the single modules, to save the figures. In fact, the figures module offers a wide plethora of functions (all based on matplotlib) to plot the data with a high degree of customization for the appearance.

The fitting functions use lmfit to build the initial guess and to minimize the difference between the experimental data and the model, generated with a Voigt profile in the time domain and then Fourier-transformed, in the least-square sense (employing the Levenberg-Marquardt algorithm implemented in scipy). For this purpose, the class Voigt\_fit of the fit module includes attribute functions to construct an initial guess interactively, fit the data, and save the parameters in dedicated files.

Regarding the development of the package, I would like to acknowledge Letizia Fiorucci for her contribution in the design and the implementation of several functions, and for the alpha-testing.

<sup>1</sup>https://www.nmrglue.com/

 $<sup>^2</sup>$ Multivariate Curve Resolution: 50 years addressing the mixture analysis problem - A review

<sup>&</sup>lt;sup>3</sup>Denoising NMR time-domain signal by singular-value decomposition accelerated by graphics processing units

# 2. User guide

# 2.1 Initialize the package

Initialize the package by writing, at the top of your file:

```
from klassez import *
```

This line also implies:

```
import os
import sys
import numpy as np
from scipy import linalg, stats
import matplotlib
import matplotlib.pyplot as plt
import matplotlib.cm as cm
from pprint import pprint
```

This means these can be not imported in your code, as KLASSEZ already does it for you.

#### 2.1.1 Extra variables

Initializing KLASSEZ also grants access to CM and COLORS.

CM is a dictionary of colormaps taken from seaborn and saved in a dictionary whose keys are their names, so that also matplotlib can use them. You can inspect the keys through:

```
print(CM.keys())
```

```
COLORS is:
```

```
colors = [ 'tab:blue', 'tab:red', 'tab:green', 'tab:orange', 'tab:cyan', 'tab:purple',
    'tab:pink', 'tab:gray', 'tab:brown', 'tab:olive', 'salmon', 'indigo', 'm', 'c', 'g',
    'r', 'b', 'k', ]
```

repeated cyclically ten times and stored as tuple.

Other two 'quality of life' variables are figures.figsize\_small and figures.figsize\_large, which correspond to figure panel sizes of  $3.59 \times 2.56$  inches and  $15 \times 8$  inches, respectively. The former suits well for saving figures of spectra with font sizes of about 10 pt, whereas the latter are best for GUIs and withstand font sizes of about 14 pt.

For NMR: the variable sim.gamma is a dictionary containing the gyromagnetic ratio, in MHz/T, of all the magnetically-active nuclei. For instance:

```
print(sim.gamma['13C'])
>>> 10.70611
```

A decorator function called **cron** is defined in the top-level script **config**, and imported by **\_\_init\_\_**, so that you can use it after writing:

```
from klassez import cron
```

This decorator allows to measure the runtime of a function, and print it on standard output once it ended.

# 2.2 Processing of a 'raw' 1D spectrum

Let us say that your spectrum is saved in the folder /home/myself/spectra/mydataset/1/. Initialize the spectrum object through:

```
Path = "/home/myself/spectra/mydataset/1/"
s = Spectrum_1D(Path)
```

This command will do three main tasks:

- read the binary FID of your spectrum and store it in a complex array s.fid;
- load the acquisition parameters, read the interesting keys and store them in a dictionary s.acqus;
- initialize a dictionary s.procs which contains the processing parameters.

If there is the group delay at the beginning of the FID, you are advised to write

```
s.convdta()
```

to remove it. However, this is tested for Bruker data and does not always work, because it depends on the version of TopSpin that the spectrometer ran and several other construction parameters. Therefore, if you see a residual of the digital filter in you spectrum, the easiest solution is to run CONVDTA inside TopSpin.

KLASSEZ is able to read also Varian and Spinsolve (Magritek) data, by specifying the option 'spect'.

A detailed description of acqus and procs is shown in table 2.1 and table 2.2.

Please note that reading the spectrum causes the program to save a file called 'name.procs', where 'name' is the path name.

To make the Fourier transform of the FID to obtain the spectrum, you must invoke the **process** method, which reads the **procs** dictionary to get the instructions on the processing you want to make on your spectrum. For instance, if you want to obtain a final spectrum of 8k points with an exponential broadening of  $25 \,\mathrm{Hz}$ :

```
s.procs["wf"]["mode"] = "em"
s.procs["wf"]["lb"] = 25
s.procs["zf"] = 8192
s.process()
```

Calling the process method generates new attributes of the class:

- freq: the frequency scale, in Hz;
- ppm: the ppm scale;
- r: the real part of the spectrum;

Table 2.1: Description of the acqus dictionary of a Spectrum\_1D object.

Key	Explanation	
В0	Magnetic field strength /T	
BYTORDA	A Endianness of binary data: 0 little endian, 1 big endian	
DTYPA	Binary data type: 0 int32, 2 float64	
GRPDLY	Number of points of the digital filter	
nuc	Observed nucleus	
o1p	Carrier frequency i.e. center of the spectrum, in ppm	
01	Same as o1p, but in Hz	
SWp	Sweep width, in ppm	
SW	Sweep width, in Hz	
SF01	Larmor frequency of the observed nucleus at field B0	
TD	Number of sampled complex points	
dw	dw Dwell time, i.e. the sampling interval, in seconds	
AQ	Time duration of the FID	
t1	Acquisition timescale	

- i: the imaginary part of the spectrum;
- S: the complex spectrum (S = r + ii).

After the Fourier transform, the process method applies the phase correction and the calibration using the phase angles and the calibration value saved in the *procs* dictionary automatically. This allows the user to not phase their spectra every time, as well as keeping a record of the processing.

If the spectrum requires phase correction, you can perform it interactively:

#### s.adjph()

or by passing the phase angles, in degrees, to adjph. Example, if you know you need to phase your spectrum with 30 degrees of  $\phi^{(0)}$  and -55 degrees of  $\phi^{(1)}$  with the pivot set at 7.32 ppm:

```
s.adjph(p0=30, p1=-55, pv=7.32)
```

In both cases, the phase angles are updated in the procs dictionary.

The spectrum can be calibrated using a dedicated GUI:

```
s.cal()
```

or specifying the shift value in ppm or in Hz (in this case, be sure to set the isHz keyword to True).

```
s.cal(-3) # Shift of -3 ppm
s.cal(1000, isHz=True) # Shift of +1 kHz
```

Both ppm and freq are updated according to the given values.

A tool for baseline correction and lineshape fitting are also provided. Regarding the baseline correction, there are now two methods available: the 'old' one, named baseline\_correction, allows to build and read a manually-designed baseline; the rpbc method automatically tries to fit the baseline with a polynomion and correct the phase together.

```
s.basl() # subtract s.baseline from s.S, then unpack s.r and s.i
```

For the fitting:

Table 2.2: Description of the procs dictionary of a Spectrum\_1D object.

#### Key Explanation Window function. This is a dictionary itself: wf • 'mode': choose function between - 'em': exponential - 'sin': sine - 'qsin': squared sine - 'gm': mixed lorentzian-gaussian - 'gmb': mixed lorentzian-gaussian, Bruker style • 'lb': Exponential line-broadening. Read by em, gm and gmb • 'gb': Gaussian line-broadening. Read by gm and gmb • 'gc': Center of the gaussian $\in [0,1]$ . Read by gm • 'ssb': Shift of the sine bell. Read by sin and qsin 'sw': Sweep width. Automatically set according to acqus['SW'] Zero-filling. Set the *final* number of points! zf Number of points to be used for processing tdeff Scaling factor for the first point of the FID before Fourier transform fcor р0 Frequency-independent phase correction /degrees First order phase correction /degrees р1 Pivot point for the first order phase correction /ppm pν Set of coefficients of a polynomion to be used as baseline, starting from basl\_c the 0-order coefficient Offset, in ppm, to be added to the frequency and ppm scales for calical bration # Make an initial guess interactively and save all the parameters in a file called "myquess.inp" s.F.iguess(input\_file="myguess.inp") # Save a figure of the initial guess in "myguess.png" s.F.plot(what='iguess', name="myguess") # Do the fit s.F.dofit( log\_file="myfit.log", # Write the log of the fit in a file called "myfit.log" output\_file="myfit.out", # Write the output of my fit in a file called "myfit.out" # Tolerance on the chemical shift of +/-0.5 ppm utol=0.5, # Allow/Not allow to fit dephased peaks vary\_phi=False, # Allow/Not allow to change fraction of gaussianity vary\_xg=False, res\_hist\_name="myres", # Make a figure of the residuals called "myres.png" # Perform tests on the goodness of the fit test\_res=True # Plot the output and show it

s.F.plot(what='fit')

### 2.2.1 The class pSpectrum\_1D

The class Spectrum\_1D does not work if you want to read the processed data directly from TopSpin (or whatever software you used to acquire and process them). Instead, you should use the class pSpectrum\_1D, which is designed to perform exactly this task. It inherits most of the attributes and methods of the Spectrum\_1D class, therefore its usage closely resembles the example reported in the previous section.

## 2.3 Processing of a 'raw' 2D spectrum

Let us say that your spectrum is saved in the folder /home/myself/spectra/mydataset/21/. Initialize the spectrum object through:

```
Path = "/home/myself/spectra/mydataset/21/"
s = Spectrum_2D(Path)
```

The generated acqus and procs dictionaries include informations on both dimensions.

Table 2.3: Description of the acqus dictionary of a Spectrum\_2D object.

Key	Explanation		
ВО	Magnetic field strength /T		
BYTORDA	Endianness of binary data: 0 little endian, 1 big endian		
DTYPA	Binary data type: 0 int32, 2 float64		
GRPDLY	Number of points of the digital filter		
nuc1	Observed nucleus in the indirect dimension		
nuc2	Observed nucleus in the direct dimension		
o1p	Carrier frequency i.e. center of the indirect dimension, in ppm		
o2p	Carrier frequency i.e. center of the direct dimension, in ppm		
o1	Same as olp, but in Hz		
o2	Same as o2p, but in Hz		
SW1p	Sweep width of the indirect dimension, in ppm		
SW2p	Sweep width of the direct dimension, in ppm		
SW1	Sweep width of the indirect dimension, in Hz		
SW2	Sweep width of the indirect dimension, in Hz		
SF01	Larmor frequency of the observed nucleus in F1 at field B0		
SF02	Larmor frequency of the observed nucleus in F2 at field B0		
TD1	Number of $t_1$ -increments		
TD2	Number of sampled complex points		
dw1	$t_1$ increments, in seconds		
dw2	Dwell time, i.e. the sampling interval, in seconds		
AQ1	Sampled timescale of the indirect dimension		
AQ2	Time duration of the FID		
t1	Evolution timescale		
t2	Acquisition timescale		

Then, the sequence of commands resembles the ones of the 1D spectra.

s.convdta() # If there is the digital filter

s.process()

<sup>#</sup> Also in this case, phase correction and calibration are performed automatically with the values in procs

Table 2.4: Description of the procs dictionary of a Spectrum\_2D object. Each of these dictionary entry is a list of two elements: the first one (index 0) is the processing to apply on the indirect dimension, the second (index 1) on the direct dimension. For instance, procs[tdeff] = [64, 1024] means to truncate the indirect evolutions to 64 points and the FIDs to 1024 points.

Key	Explanation		
wf	Window function. This is a dictionary itself:		
	• 'mode': choose function between		
	- 'em': exponential		
	- 'sin': sine		
	- 'qsin': squared sine		
	- 'gm': mixed lorentzian-gaussian		
	- 'gmb': mixed lorentzian-gaussian, Bruker style		
	• 'lb': Exponential line-broadening. Read by em, gm and gmb		
	• 'gb': Gaussian line-broadening. Read by gm and gmb		
	• 'gc': Center of the gaussian $\in [0,1]$ . Read by gm		
	• 'ssb': Shift of the sine bell. Read by sin and qsin		
	• 'sw': Sweep width. Automatically set according acqus['SW']		
zf	Zero-filling. Set the <i>final</i> number of points!		
tdeff			
fcor			
p0			
p1	First order phase correction /degrees		
pv	Pivot point for the first order phase correction /ppm		
cal_1	Calibration offset for F1 /ppm		
cal_2	_2 Calibration offset for F2 /ppm		

s.adjph()
s.plot()

The keys for adjph are of the kind: pXY, where X is the order of the phase correction (0 or 1) and Y is the dimension on which to apply it (1 or 2). Explicative table below:

	<b>F</b> 1	<b>F2</b>
$\phi^{(0)}$	p01	p02
$\phi^{(1)}$	p11	p12
pivot	pv1	pv2

For further information, rely on the help python builtin function. To read the processed data, use the pSpectrum\_2D class instead.

#### 2.3.1 Computing projections

While the 2D spectra give an overall look on the whole experiment, the user might want to extract projection of the direct or the indirect dimension, to focus onto particular features in the spectrum. In order to do so, klassez offers two commands: projf1 and projf2, which compute the sum projections on the indirect or on the direct dimension, respectively, and store the result in dictionaries called trf1 and trf2, whose keys are the ppm values correspondant to the projections. Actually, the capitalized versions of the two dictionaries (with the same keys), i.e. Trf1 and Trf2, can be more useful, as they are instances of the pSpectrum\_1D class and therefore are initialized with ppm scales and other parameters.

Example:

```
# Supposed to have a 1H-15N HSQC spectrum

# Extract the direct dimension trace at 115 ppm, 15N scale
s.projf2(115)

# Access to it through
Proj_115 = s.Trf2['115']

# Extract the indirect dimension trace from 6 to 8 ppm, 1H scale
s.projf1(6, 8)
Proj_indim = s.Trf1['6:8']

# You can plot them:
Proj_115.plot()
Proj_indim.plot()
```

## 2.4 Simulating data

The classes Spectrum\_1D and Spectrum\_2D are also able to generate simulated data by reading a custom-written input file. The functions they use are sim.sim\_1D and sim.sim\_2D.

#### 2.4.1 Simulate 1D data

The input file you have to write *must* have the following keys:

- B0: Magnetic field strength /T;
- nuc: Observed nucleus (e.g. 13C);
- o1p: Carrier frequency i.e. centre of the spectrum /ppm;
- SWp: Sweep width /ppm. The spectrum will cover the range [o1p SWp/2, o1p + SWp/2];
- TD: Number of sampled (complex) points;
- shifts: sequence of peak positions /ppm;
- fwhm: Full-width at half-maximum of the peaks /Hz;
- amplitudes: Intensity of the peaks in the FID;
- x\_g: Fraction of gaussianity.  $x_g = 0 \implies$  pure Lorentzian peak,  $x_g = 1 \implies$  pure Gaussian peak;

and can have the following keys:

- phases: phases of the peaks /degrees. Default: all zeros;
- mult: fine structures of the peaks (e.g. doublets of triplets: dt). Default: all singlets;
- Jconst: coupling constants of the fine structures /Hz. If more of one coupling is expected, provide them as a sequence. Default: not used as the peaks are all singlets.

Key and value must be separated by a tab character. You are allowed to leave empty rows to improve the readibility and to insert comments using the # character.

Example:

```
B0 16.4 # 700 MHz 1H
nuc 1H
o1p 4.7
SWp 40
TD 8192

shifts 1, 3, 5, 7
fwhm [10 for k in range(4)]
amplitudes 10, 20, 15, 10
x_g 0, 0.4, 0.6, 1
phases 5, 0, 10, 0

mult s, t, dt, ddd
Jconst 0, 15, [12, 9.5], [25, 15, 10]
```

This input file generates the spectrum in Figure 2.1. Code:

```
#! /usr/bin/env python3
from klassez import *
s = Spectrum_1D('sim_in_1D', isexp=False)
s.process()
figures.figure1D(s.ppm, s.r, name='test_1D', X_label='$\delta\, ^1$H /ppm',
    Y_label='Intensity /a.u.')
```

#### 2.4.2 Simulate 2D data

The same procedure can be followed to simulate 2D spectra. The input file to write is very similar to the one for 1D data, except for the quantities that clearly span over two dimensions. As in NMR textbook, the direct and indirect dimensions will be named F2 and F2 respectively, and dimension-specific quantities will feature the 1 or 2 labels accordingly.

- B0: Magnetic field strength /T;
- nuc1: Observed nucleus in F1(e.g. 13C);
- nuc2: Observed nucleus in F2(e.g. 1H);
- olp: Carrier frequency i.e. centre of F1 /ppm;

- o2p: Carrier frequency i.e. centre of F2 /ppm;
- SW1p: Sweep width /ppm. The indirect dimension will cover the range [o1p SW1p/2, o1p + SW1p/2];
- SW2p: Sweep width /ppm. The direct dimension will cover the range [o2p SW2p/2, o2p + SW2p/2];
- TD1: Number of sampled (complex) points in F1;
- TD2: Number of sampled (complex) points in F2;
- shifts\_f1: sequence of peak positions in F1 /ppm;
- shifts\_f2: sequence of peak positions in F2 /ppm;
- fwhm\_f1: Full-width at half-maximum of the peaks in F1 /Hz;
- fwhm\_f2: Full-width at half-maximum of the peaks in F2 /Hz;
- amplitudes: Intensity of the peaks in the FID;
- x\_g: Fraction of gaussianity.  $x_g = 0 \implies$  pure Lorentzian peak,  $x_g = 1 \implies$  pure Gaussian peak;

Phase distortions and fine structures are not allowed for multidimensional spectra. The indirect dimension will be generated employing the *States-TPPI* sampling scheme.

Example:

```
B0 28.2
nuc1
       15N
nuc2
       1H
o1p 115
o2p 5
SW1p
       40
SW2p
       20
TD1 512
TD2 8192
shifts_f1 130.0, 105.0, 120.0, 1.25e2, 130.0, 105.0
shifts_f2 0.0, 0.0, 4.0, 7.0, 1.1e1, 10.5
fwhm_f1 100, 100, 100, 100, 100, 100
fwhm_f2 50, 50, 50, 50, 50, 50
amplitudes 10, 20, 10, 20, 10, 10
x_g 0.0, 0.2, 0.4, 0.6, 0.8, 1.0
```

This input file generates the spectrum in Figure 2.2. Code:

```
#! /usr/bin/env python3
```

```
from klassez import *

s = Spectrum_2D('sim_in_2D', isexp=False)
s.process()

figures.figure2D(s.ppm_f2, s.ppm_f1, s.rr, lvl=0.005, name='test_2D', X_label='$\delta\, ^1$H /ppm', Y_label='$\delta\, ^{15}$N /ppm')
```



Figure 2.1: Example of a simulated 1D spectrum.



Figure 2.2: Example of a simulated 2D spectrum.

### 2.5 The Pseudo\_2D class

Sometimes, the spectroscopist might find interesting to acquire a series of 1D experiments in which one (or more) parameters are changed according to a certain schedule. This kind of experiments are 2D in principle, but their processing and analysis resemble the one of 1D spectra. Therefore, they lie somewhere in between 1D spectra and 2D spectra, hence they are often referred to as pseudo 2D.

Also in this case, klassez offers a specific class to deal with this kind of data: Pseudo\_2D. Pseudo\_2D is a subclass of Spectrum\_2D; however, many functions have been adapted to resemble the 1D version.

Pseudo\_2D does not encode for a routine to automatically simulate data. If you want to, you should give a 1D-like input file (just like the one in section 2.4.1), and replace the attribute fid with your FID, generated as you wish. With a real dataset this is not required, as it is able to read everything automatically.

```
path_to_pseudo = "/home/myself/spectra/mydataset/899/"
s = Pseudo_2D(path_to_pseudo)
s.convdta() # If there is the digital filter
```

The process() function applies apodization, zero-filling and Fourier transform only on the direct dimension, reading the parameters from a procs dictionary like the one of  $pectrum_1D$ . The attributes  $processive freq_f1$  and  $ppm_f1$  are initialized with p.arange(N), where N is the number of experiments that your FID comprises of.

The phase adjustment is performed on a reference spectrum, then applied on the whole 2D matrix. By default, the chosen spectrum is the first one, but you can choose the one that fits the most your needs.

```
s.process()
s.adjph(expno = 10) # Calls interactive_phase_1D on the 10th experiment
```

The method plot shows the 2D contour map of the spectrum, just like the one of Spectrum\_2D. However, this is not always the most intelligent way to plot the data in order to gather information. This is the reason why this class features two unique additional methods that plot data: plot\_md and plot\_stacked. Both rely on the parameter which, that is a string of code (i.e. it should be interpreted by eval) that identifies which experiment to show by pointing at their index. which = 'all' results in pointing at all spectra.

```
s.plot() # 2D contour map
s.plot_md(which="3, 5, 11") # Plot the 3rd, the 5th and the 11th spectrum, superimposed
s.plot_stacked(which="np.arange(0,100,5)") # Makes a stacked plot with a spectrum every 5
```

The method integrate differs a little bit from the one coded in Spectrum\_1D.

```
s.integrate(which=2) # Interactive panel on the 3rd spectrum
```

Even if you select the integration limits on a single spectrum, the method integrate will compute the integrals throughout the whole range of experiment. This means that each entry of integrals will be an array as long as the number of experiment.

# 3. List of modules and functions

# 3.1 MISC package

This package contains miscellaneous functions for the calculation of several properties, and generally for the handling of NMR spectra.

## 3.1.1 misc.SNR(data, signal=None, n reg=None)

Computes the signal to noise ratio of a 1D spectrum, as:

$$SNR = \frac{S}{2\sigma_n}$$

#### **Parameters**

- data: 1darray

  The spectrum of which you want to compute the SNR
- signal : float, optional
  If provided, uses this value as maximum signal. Otherwise, it is selected as the maximum value in data
- n\_reg: list or tuple, optional If provided, contains the points that delimit the noise region. Otherwise, the whole spectrum is used.

#### Returns

• snr : float
The SNR of the spectrum

## 3.1.2 misc.SNR 2D(data, n reg=None)

Computes the signal to noise ratio of a 2D spectrum.

#### **Parameters**

• data: 1darray

The spectrum of which you want to compute the SNR

• n\_reg: list or tuple
If provided, the points of F1 scale and F2 scale, respectively, of which to extract the projections.
Otherwise, opens the tool for interactive selection.

#### Returns

• snr\_f1 : float
The SNR of the indirect dimension

• snr\_f2 : float The SNR of the direct dimension

# $3.1.3 \quad misc.avg\_antidiag(X)$

Given a matrix X without any specific structure, finds the closest Hankel matrix in the Frobenius norm sense by averaging the antidiagonals.

#### Parameters:

• X: 2darray Input matrix

#### Returns:

• Xp: 2darray
Hankel matrix obtained from X

# 3.1.4 misc.binomial\_triangle(n)

Calculates the n-th row of the binomial triangle. The first row is n=1, not 0. Example:

binomial\_triangle(4)
>>> 1 3 3 1

#### Parameters:

• n: *int* Row index

#### Returns:

 $\bullet$  row: 1darray The n-th row of binomial triangle.

# 3.1.5 misc.calcres(fqscale)

Calculates the frequency resolution of an axis scale, i.e. how many Hz is a 'tick'.

#### **Parameters**

• fqscale : *1darray* Scale to be processed

#### Returns

ullet res : float The resolution of the scale

## 3.1.6 misc.cmap2list(cmap, N=10, start=0, end=1)

Extract the colors from a colormap and return it as a list.

#### Parameters:

- cmap: matplotlib.Colormap Object

  The colormap from which you want to extract the list of colors
- N: *int* Number of samples to extract
- start: float
  Start point of the sampling. 0 = beginning of the cmap; 1 = end of the cmap.
- end: float
  End point of the sampling. 0 = beginning of the cmap; 1 = end of the cmap.

#### Returns:

• colors: *list*List of the extracted colors.

# $3.1.7 \quad misc.edit\_checkboxes(checkbox, xadj=0, yadj=0, length=None, height=None, length=None, height=None, height=None,$

Edit the size of the box to be checked, and adjust the lines accordingly.

#### Parameters:

- checkbox: matplotlib.widgets.CheckBox Object
  The checkbox to edit
- xadj: *float* modifier value for bottom left corner x-coordinate of the rectangle, in checkbox.ax coordinates
- yadj: float modifier value for bottom left corner y-coordinate of the rectangle, in checkbox.ax coordinates
- length: *float* length of the rectangle, in checkbox.ax coordinates
- height: *float* height of the rectangle, in checkbox.ax coordinates
- color: str or list or None

  If it is not None, change color to the lines

# 3.1.8 misc.find\_nearest(array, value)

Finds the value in array which is the nearest to value.

#### **Parameters**

• array : *1darray* Self-explanatory

• value : *float* Value to be found

#### Returns

• val: *float*The closest value in array to value

# $3.1.9 \quad misc.freq2ppm(x, B0=701.125, o1p=0)$

Converts x from Hz to ppm.

#### **Parameters**

• x : float Value to be converted

• B0 : float Field frequency, in MHz. Default: 700 MHz

• olp: float Carrier frequency, in ppm. Default: 0.

#### Returns

• y : *float*The converted value

## 3.1.10 misc.get trace(data, ppm f2, ppm f1, a, b=None, column=True)

Takes as input a 2D dataset and the ppm scales of direct and indirect dimensions respectively. Calculates the projection on the given axis summing from a (ppm) to b (ppm). Default: indirect dimension projection (i.e. column=True), change it to 'False' for the direct dimension projection. Returns the calculated 1D projection.

#### **Parameters**

- data : 2darray

  Spectrum of which to extract the projections
- ppm\_f2 : *1darray* ppm scale of the direct dimension
- ppm\_f1 : *1darray* ppm scale of the indirect dimension
- ullet a : float The ppm value from which to start extracting the projection.
- b: float, optional
  If provided, the ppm value at which to stop extracting the projection. Otherwise, returns only the a trace.
- column: bool

  If True, extracts the F1 projection. If False, extracts the F2 projection.

#### Returns

• y : *1darray* Computed projection

# 3.1.11 misc.get\_ylim(data\_inp)

Calculates the y-limits of a plot as follows:

- Bottom: min(data) 5% max(data)
- Top: max(data) + 5% max(data)

#### Parameters:

• data\_inp: ndarray or list Input data. If it is a list, data\_inp is converted to array.

## 3.1.12 misc.hankel(data, n=None)

Computes a Hankel matrix from data. If data is a 1darray of length N, computes the correspondant Hankel matrix of dimensions (N - n + 1, n). If data is a 2darray, computes the closest Hankel matrix in the Frobenius norm sense by averaging the values on the antidiagonals.

#### Parameters:

- data: 1darrayVector to be Hankel-ized, of length N
- $\bullet$  n: int Number of columns that the Hankel matrix will have

#### Returns:

• H: 2darrayHankel matrix of dimensions (N - n + 1, n)

# 3.1.13 misc.hz2pt(fqscale, hz)

Converts hz from frequency units to points, on the basis of its scale.

#### **Parameters**

• fqscale : *1darray* Scale to be processed

• hz : *float* Value to be converted

#### Returns

• pt : *float*The frequency value converted in points

# $3.1.14 \quad misc.in2px(*in\_args)$

Converts a sequence of numbers from inches to pixels by multiplying  $\times 96$ .

#### Parameters:

• in\_args: sequence of floats Values in inches to convert

#### Returns:

• px\_args: tuple of ints Values in pixels

# 3.1.15 misc.makeacqus\_1D(dic)

Given a NMRGLUE dictionary from a 1D spectrum (generated by ng.bruker.read), this function builds the acqus file with only the 'important' parameters.

#### **Parameters**

• dic: dict

NMRglue dictionary returned by ng.bruker.read

#### Returns

• acqus : *dict*Dictionary with only few parameters

# $3.1.16 \quad misc.makeacqus\_1D\_oxford(dic)$

Given a NMRGLUE dictionary from a 1D spectrum (generated by ng.jcampdx.read), this function builds the acqus file with only the 'important' parameters.

#### Parameters:

#### Returns:

• acqus : *dict*Dictionary with only few parameters

# $3.1.17 \quad misc.makeacqus\_2D(dic)$

Given a NMRGLUE dictionary from a 2D spectrum (generated by ng.bruker.read), this function builds the acqus file with only the 'important' parameters.

#### **Parameters**

• dic: dict

NMRglue dictionary returned by ng.bruker.read

#### Returns

• acqus : *dict*Dictionary with only few parameters

# 3.1.18 misc.mathformat(ax, axis='y', limits=(-2,2))

Apply exponential formatting to the given axis of the given figure panel. The offset text size is uniformed to the tick labels' size.

- ax: matplotlib.Subplot Object Panel of the figure to edit
- axis: *str*'x', 'y' or 'both'.
- limits: tuple tuple of ints that indicate the order of magnitude range outside which the exponential format is applied.

# 3.1.19 misc.molfrac(n)

Computes the 'molar fraction'  ${\tt x}$  of the array  ${\tt n}.$  Returns also the total amount.

$$x_i = \frac{n_i}{N}, \qquad N = \sum_i n_i$$

#### **Parameters**

• n: list or 1darray list of values

- x: *list or 1darray* molar fraction array
- N: float sum of all the elements in n

# 3.1.20 misc.noise std(y)

Calculates the standard deviation of the noise using the Bruker formula. Taken y as an array of N points, and  $y_i$  its i-th entry:

$$\sigma_n = \frac{1}{\sqrt{N-1}} \sqrt{\sum_{i=1}^{N} y_i^2 - \frac{1}{N} \left[ \left( \sum_{i=1}^{N} y_i \right)^2 + \frac{3 \left( \sum_{i=1}^{N/2} i (y_{N-i+1} - y_i)^2 \right)}{N^2 - 1} \right]}$$

#### **Parameters**

 $\bullet$  y : 1darray The spectral region you would like to use to calculate the standard deviation of the noise.

#### Returns

• noisestd : *float*The standard deviation of the noise.

# $3.1.21 \quad misc.nuc\_format(nuc)$

Converts the nuc key you may find in acqus in the formatted label, e.g. '13C'  $\rightarrow$  '\${13}\$C'

#### Parameters:

• nuc: strUnformatted string

#### Returns:

• fnuc: str Formatted string.

# 3.1.22 misc.polyn(x, c)

Computes p(x), polynomion of degree n-1, where n is the number of provided coefficients.

#### **Parameters**

- x : 1darray
  Scale upon which to build the polynomion
- c : list or 1darray Sequence of the polynomion coefficient, starting from the 0-th order coefficient

#### Returns

• px : 1darrayPolynomion of degree n-1.

# $3.1.23 \quad misc.ppm2freq(x, B0=701.125, o1p=0)$

Converts x from ppm to Hz.

#### **Parameters**

• x : float Value to be converted

• B0 : float Field frequency, in MHz. Default: 700 MHz

• o1p : float Carrier frequency, in ppm. Default: 0.

#### Returns

• y : *float*The converted value

# $3.1.24 \quad misc.ppmfind(ppm\_scale, \, value)$

Finds the exact value in ppm scale.

#### **Parameters**

• ppm\_scale : 1darray Self-explanatory

ullet value : float

The value to be found

#### Returns

• V: float
The closest value to value in ppm\_scale

## 3.1.25 misc.pretty scale(ax, limits, axis='x', n major ticks=10)

This function computes a pretty scale for your plot. Calculates and sets a scale made of n\_major\_ticks numbered ticks, spaced by 5 \* n\_major\_ticks unnumbered ticks. After that, the plot borders are trimmed according to the given limits.

- ax: matplotlib.AxesSubplot object
  Panel of the figure of which to calculate the scale
- limits: *tuple* limits to apply of the given axis. (left, right)
- axis: str'x' for x-axis, 'y' for y-axis
- $\bullet$  n\_major\_ticks: int Number of numbered ticks in the final scale. An oculated choice gives very pleasant results.

# 3.1.26 misc.print\_dict(mydict)

Prints a dictionary one entry per row, in the format key: value. Nested dictionaries are printed with an indentation

#### Parameters:

• mydict: dict The dictionary you want to print

#### Returns:

 $\bullet$  outstring: strThe printed text formatted as single string

# $3.1.27 \quad misc.print\_list(mylist)$

Prints a list, one entry per row.

#### Parameters:

 $\bullet$  mylist: listThe list you want to print

#### Returns:

ullet outstring: strThe printed text formatted as single string

## 3.1.28 misc.procpar(txt)

Takes as input the path of a file containing a 'key' in the first column and a 'value' in the second column. Returns a dictionary of shape key: value.

#### **Parameters**

 $\bullet$  txt : strPath to a file that contains 'key' in first column and 'value' in the second

#### Returns

• procpars : *dict*Dictionary of shape key : value.

# $3.1.29 \quad misc.px2in(*px\_args)$

Converts a sequence of numbers from inches to pixels by multiplying  $\times 96$ .

#### Parameters:

• px\_args: sequence of ints Values in pixels to convert

#### Returns:

• in\_args: tuple of floats Values in inches

## 3.1.30 misc.readlistfile(datafile)

Takes as input the path of a file containing one entry for each row. Returns a list of the aforementioned entries.

#### Parameters

ullet datafile: str Path to a file that contains one entry for each row

#### Returns

• files : *list*List of the entries contained in the file

# 3.1.31 misc.select for integration(ppm f1, ppm f2, data, Neg=True)

Select the peaks of a 2D spectrum to integrate. First, select the area where your peak is located by dragging the red square. Then, select the center of the peak by right\_clicking. Finally, click 'ADD' to store the peak. Repeat the procedure for as many peaks as you want.

#### Parameters:

- ppm\_f1: *1darray* ppm scale of the indirect dimension
- ppm\_f2: 1darray ppm scale of the direct dimension
- data : 2darray Spectrum
- Neg: bool
  Choose if to show the negative contours (True) or not (False)

#### Returns:

• peaks: list of dict
For each peak there are two keys, 'f1' and 'f2', whose meaning is obvious. For each of these keys, you have 'u': center of the peak /ppm, and 'lim': the limits of the square you drew before.

## 3.1.32 misc.select traces(ppm f1, ppm f2, data, Neg=True, grid=False)

Select traces from a 2D spectrum, save the coordinates in a list. Left click to select a point, right click to remove it.

#### **Parameters**

- ppm\_f1 : *1darray* ppm scale of the indirect dimension
- ppm\_f2 : *1darray* ppm scale of the direct dimension
- data : 2darray Spectrum
- Neg: bool
  Choose if to show the negative contours (True) or not (False)
- grid : *bool*Choose if to display the grid (True) or not (False)

#### Returns

• coord: *list*List containing the [x,y] coordinates of the selected points.

## 3.1.33 misc.set fontsizes(ax, fontsize=10)

Automatically adjusts the fontsizes of all the figure elements. In particular:

- title = fontsize
- axis labels = fontsize -2
- ticks labels = fontsize -3
- legend entries = fontsize -4

- ax: matplotlib.Subplot Object Subplot of interest
- fontsize: *float* Starting fontsize

# 3.1.34 misc.set\_ylim(ax, data\_inp)

Sets the y-limits of ax as follows:

- Bottom: min(data) 5% max(data)
- Top: max(data) + 5% max(data)

- ax: matplotlib.Subplot Object
  Panel of the figure where to apply this scale
- data\_inp: ndarray or list
  Input data. If it is a list, data\_inp is converted to array.

## 3.1.35 misc.show cmap(cmap, N=10, start=0, end=1)

Plot the colors extracted from a colormap.

- cmap: matplotlib.Colormap Object

  The colormap from which you want to extract the list of colors
- N: *int* Number of samples to extract
- start: float
  Start point of the sampling. 0 = beginning of the cmap; 1 = end of the cmap.
- end: float
  End point of the sampling. 0 = beginning of the cmap; 1 = end of the cmap.

# $3.1.36 \quad misc.split\_acqus\_2D(acqus)$

Split the acqus dictionary of a 2D spectrum into two separate 1D-like acqus dictionaries.

#### Parameters:

• acqus: *dict* acqus dictionary of a 2D spectrum

- acqu1s: *dict* acqus dictionary of the indirect dimension
- acqu2s: *dict* acqus dictionary of the direct dimension

# $3.1.37 \quad misc.split\_procs\_2D(procs)$

Split the procs dictionary of a 2D spectrum into two separate 1D-like procs dictionaries.

#### **Parameters**

• procs: *dict* procs dictionary of a 2D spectrum

- proc1s: *dict* procs dictionary of the indirect dimension
- proc2s: dict procs dictionary of the direct dimension

# 3.1.38 misc.trim\_data(ppm\_scale, y, sx, dx)

Trims the frequency scale and correspondant 1D dataset y from sx (ppm) to dx (ppm).

#### **Parameters**

- ppm\_scale : 1darray ppm scale of the spectrum
- y : 1darray spectrum
- sx : *float* ppm value where to start trimming
- dx : *float* ppm value where to finish trimming

#### Returns

• xtrim : 1darray
Trimmed ppm scale

• ytrim : *1darray* Trimmed spectrum

# $3.1.39 \quad misc.trim\_data\_2D(x\_scale, y\_scale, data, xlim=None, ylim=None)$

Trims data and scales according to xlim and ylim. Returns the trimmed data and the correspondant trimmed scales.

#### Parameters:

- x\_scale: *1darray*Scale for the rows of data
- y\_scale: 1darray
  Scale for the columns of data
- data: *2darray*Data to be trimmed
- xlim: tuple
  Limits for x\_scale (L, R)
- ylim: tupleLimits for y\_scale (L, R)

- trimmed\_x: 1darray
  Trimmed x-scale
- trimmed\_y: 1darray Trimmed y-scale
- trimmed\_data: 2darray Trimmed data

## 3.1.40 misc.unhankel(H)

Concatenates the first row and the last column of the matrix H, which should have Hankel-like structure, so to build the array of independent parameters.

#### Parameters:

• H: 2darray Hankel-like matrix

#### Returns:

• h: *1darray*First row and last column of H, concatenated

## 3.1.41 misc.write acqus 1D(acqus, path='./', filename=None)

Writes the input file for a simulated spectrum, basing on a dictionary of parameters.

#### **Parameters**

ullet acqus: dict The dictionary containing the parameters for the simulation

• path: str, optional
Directory where the file will be saved.

• filename: str, optionalName of the file to be saved in path. The default name is  $sim_in_1D$ 

# 3.1.42 misc.write\_acqus\_2Dacqus, path='sim\_in\_2D')

Writes the input file for a simulated spectrum, basing on a dictionary of parameters.

#### Parameters

ullet acqus: dict The dictionary containing the parameters for the simulation

• path: str, optional
Directory where the file will be saved. The default name is sim\_in\_1D.

# 3.1.43 misc.write\_help(request, file=None)

Gets the documentation of request, and tries to save it in a text file.

- request: function or class or package Whatever you need documentation of
- file: str or None or False

  Name of the output documentation file. If it is None, a default name is given. If it is False, the output is printed on screen.

# $3.1.44 \quad misc.write\_ser(fid, path='./', BYTORDA=0, DTYPA=0, overwrite=True)$

Writes the FID file in directory 'path', in a TopSpin-readable way (i.e. little endian, int32). The binary file is named 'fid' if 1D, 'ser' if multiD. The parameters BYTORDA and DTYPA can be found in the acqus file.

- BYTORDA =  $1 \Rightarrow \text{big endian} \Rightarrow '>'$
- BYTORDA =  $0 \Rightarrow$  little endian  $\Rightarrow$  '<'
- DTYPA =  $0 \Rightarrow int32 \Rightarrow 'i4'$
- DTYPA =  $2 \Rightarrow \text{float}64 \Rightarrow \text{'f8'}$

- fid : *ndarray* FID array to be written
- path : strDirectory where to save the file

# 3.2 PROCESSING package

This package contains functions for the processing of NMR spectra, either in time domain or in frequency domain, and the transition between the two domains.

### 3.2.1 processing.Cadzow(data, n, nc, print head=True)

This functions performs Cadzow denoising on  $\mathtt{data}$ , which is a 1D array of N points. The algorithm works as follows:

- 1. Transform data in a Hankel matrix  $\mathbb{H}$  of dimensions (N-n,n)
- 2. Make SVD on  $\mathbb{H} = \mathbb{USV}^{\dagger}$
- 3. Keep only the first nc singular values, and put all the rest to  $0 (S \to S')$
- 4. Rebuild  $\mathbb{H}' = \mathbb{US}'\mathbb{V}^{\dagger}$
- 5. Average the antidiagonals to rebuild the Hankel-type structure, then make 1D array

#### **Parameters**

- data: *1darray* Input data
- n: *int*Number of columns of the Hankel matrix.
- nc: *int*Number of singular values to keep.
- print\_head: bool
  Set it to True to display the fancy heading.

#### Returns

• datap: *1darray*Denoised data

# $3.2.2 \quad processing. Cadzow\_2D(data, n, nc, i=True, itermax=100, f=0.005, \\ print \quad time=True)$

Performs the Cadzow denoising method on a 2D spectrum, one transient at the time. This function calls Cadzow if i=False, or iterCadzow if i=True.

#### **Parameters**

- data: 2darray Input data
- n: *int*Number of columns of the Hankel matrix.
- nc: *int*Number of singular values to keep.
- i: bool

  Calls processing.Cadzow if i=False, or processing.iterCadzow if i=True.
- itermax: *int*Maximum number of iterations allowed.
- f: float
  Factor for the arrest criterion.
- print\_time: *bool*Set it to True to display the time spent.

#### Returns

• datap: 2darray
Denoised data

## 3.2.3 processing.EAE(data)

Shuffles data if the spectrum is acquired with 'FnMODE': 'Echo-Antiecho'. **NOTE:** introduces -90° phase shift in F1, to be corrected after the processing.

```
pdata = np.zeros_like(data)

pdata[::2] = (data[::2].real - data[1::2].real) + 1j*(data[::2].imag - data[1::2].imag)

pdata[1::2] = -(data[::2].imag + data[1::2].imag) + 1j*(data[::2].real + data[1::2].real)
```

#### **Parameters**

• data : ndarray
Data to be shuffled.

#### Returns

• pdata : ndarray Shuffled data.

## 3.2.4 processing.LRD(data, nc)

Denoising method based on Low-Rank Decomposition. The algorithm performs a singular value decomposition on data, then keeps only the first nc singular values while setting all the others to 0. Finally, rebuilds the data matrix using the modified singular values.

#### Parameters:

• data: *2darray*Data to be denoised

• nc: *int*Number of components, i.e. number of singular values to keep

#### Returns:

• data\_out: 2darray
Denoised data

# 3.2.5 processing.MCR(input\_data, nc, f=10, tol=1e-5, itermax=1e4, H=True, oncols=True)

This is an implementation of Multivariate Curve Resolution for the denoising of 2D NMR data.

Let us consider a matrix  $\mathbb{D}$ , of dimensions  $m \times n$ , where the starting data are stored. The final purpose of MCR is to decompose the  $\mathbb{D}$  matrix as follows:

$$\mathbb{D} = \mathbb{CS} + \mathbb{E}$$

where  $\mathbb{C}$  and  $\mathbb{S}$  are matrices of dimension  $m \times nc$  and  $nc \times n$ , respectively, and  $\mathbb{E}$  contains the part of the data that are not reproduced by the factorization.

Being  $\mathbb{D}$  the FID of a NMR spectrum,  $\mathbb{C}$  will contain time evolutions of the indirect dimension, and  $\mathbb{S}$  will contain transients in the direct dimension.

The total MCR workflow can be separated in two parts: a first algorithm that produces an initial guess for the three matrices  $\mathbb{C}$ ,  $\mathbb{S}$  and  $\mathbb{E}$  (SIMPLISMA), and an optimization step that aims at the removal of the unwanted features of the data by iteratively filling the  $\mathbb{E}$  matrix (MCR ALS).

This function returns the denoised datasets,  $\mathbb{CS}$ , and the single  $\mathbb{C}$  and  $\mathbb{S}$  matrices.

#### **Parameters**

- input\_data: 2darray or 3darray a 3D array containing the set of 2D NMR datasets to be coprocessed stacked along the first dimension. A single 2D array can be passed, if the denoising of a single dataset is desired.
- nc: *int* number of purest components to be looked for;
- f: float percentage of allowed noise;
- tol: *float* tolerance for the arrest criterion;
- itermax: *int* maximum number of allowed iterations
- H: bool
  True for horizontal stacking of data (default), False for vertical;
- oncols: boolTrue to estimate  $\mathbb S$  with processing.SIMPLISMA, False to estimate  $\mathbb C$ .

- CS\_f: 2darray or 3darray Final denoised data matrix
- C\_f: 2darray or 3darray Final C matrix
- S\_f: 2darray or 3darray Final S matrix

## 3.2.6 processing.MCR ALS(D, C, S, itermax=10000, tol=1e-5)

Performs alternating least squares to get the final  $\mathbb C$  and  $\mathbb S$  matrices. Being the fundamental MCR equation:

$$\mathbb{D} = \mathbb{CS} + \mathbb{E}$$

At the k-th step of the iterative cycle:

1. 
$$\mathbb{C}^{(k)} = \mathbb{DS}^{+(k-1)}$$

$$2. \ \mathbb{S}^{(k)} = \mathbb{C}^{+(k)} \mathbb{D}$$

3. 
$$\mathbb{E}^{(k)} = \mathbb{D} - \mathbb{C}^{(k)}\mathbb{S}^{(k)}$$

Defined  $r_C$  and  $r_S$  as the Frobenius norm of the difference of  $\mathbb C$  and  $\mathbb S$  matrices between two subsequent steps:

$$r_C = \|\mathbb{C}^{(k)} - \mathbb{C}^{(k-1)}\|_F$$
  $r_S = \|\mathbb{S}^{(k)} - \mathbb{S}^{(k-1)}\|_F$ 

The convergence is reached when:

$$r_C \leq au$$
 ol  $r_S \leq au$  ol

#### **Parameters**

- D: 2darrayInput data, of dimensions  $m \times n$
- C: 2darray Estimation of the  $\mathbb C$  matrix, of dimensions  $m \times \mathtt{nc}$ .
- S: 2darrayEstimation of the S matrix, of dimensions  $nc \times n$ .
- itermax: *int*Maximum number of iterations
- tol: *float*Threshold for the arrest criterion.

- C: 2darray Optimized  $\mathbb{C}$  matrix, of dimensions  $m \times nc$ .
- S: 2darray Optimized  $\mathbb{S}$  matrix, of dimensions  $nc \times n$ .

## 3.2.7 processing.MCR unpack(C, S, nds, H=True)

Reverts matrix augmentation of  $stack\_MCR$ . If H=True, converts C from dimensions (Y,nds) to (X,Y,nds) and S from dimensions (nds,X\*Z) to (X,nds,Z); if H=False converts C from dimensions (Y,nds) to (X,Y,nds) and S from dimensions (nds,X\*Z) to (X,nds,Z).

#### **Parameters**

- C: 2darrayMCR  $\mathbb{C}$  matrix
- S: 2darray MCR S matrix
- nds: *int*Number of datasets to be unpacked.
- H: bool
  True for horizontal stacking, False for vertical stacking.

- C\_f: 3darray Not-augmented C matrix.
- S\_f: 3darray Not-augmented S matrix.

### 3.2.8 processing.SIMPLISMA(D, nc, f=10, oncols=True)

Finds the first nc purest components of matrix D using the SIMPLISMA algorithm, proposed by Windig and Guilment (DOI: 10.1021/ac00014a016). If oncols=True, this function estimates  $\mathbb{S}$  with SIMPLISMA, then calculates  $\mathbb{C} = \mathbb{DS}^+$ . If oncols=False, this function estimates  $\mathbb{C}$  with SIMPLISMA, then calculates  $\mathbb{S} = \mathbb{C}^+\mathbb{D}$ . f defines the percentage of allowed noise.

#### **Parameters**

- D: 2darrayInput data, of dimensions  $m \times n$
- nc: intNumber of components to be found. This determines the final size of the  $\mathbb C$  and  $\mathbb S$  matrices.
- f: float
  Percentage of allowed noise.
- ullet oncols: bool If True, SIMPLISMA estimates the  $\mathbb S$  matrix, otherwise estimates  $\mathbb C$ .

- C: 2darray Estimation of the  $\mathbb{C}$  matrix, of dimensions  $m \times nc$ .
- S: 2darrayEstimation of the  $\mathbb{S}$  matrix, of dimensions  $nc \times n$ .

# 3.2.9 processing.acme(data, m=1, a=5e-05)

Automated phase Correction based on Minimization of Entropy. This algorithm allows for automatic phase correction by minimizing the entropy of the m-th derivative of the spectrum, as explained in detail by L. Chen et.al. in *Journal of Magnetic Resonance 158 (2002) 164-168*.

Defined the entropy of h as:

$$S = -\sum_{j} h_j \ln h_j$$

and

$$h = \frac{\left| R_j^{(m)} \right|}{\sum_j \left| R_j^{(m)} \right|}$$

where

$$R = \text{Re}\{\text{spectrum}\,e^{-\mathrm{i}\phi}\}$$

and  $R^{(m)}$  is the m-th derivative of R, the objective function to minimize is:

$$Q = S + P(R)$$

where P(R) is a penalty function for negative values of the spectrum.

The phase correction is applied using processing.ps. The values p0 and p1 are fitted using Nelder-Mead algorithm.

## Parameters:

- data: *1darray*Spectrum to be phased, complex
- $\bullet$  m: int Order of the derivative to be computed
- a: *float*Weighting factor for the penalty function

## Returns:

- p0f: float
  Fitted zero-order phase correction, in degrees
- p1f: float Fitted first-order phase correction, in degrees

# 3.2.10 processing.align(ppm scale, data, lims, u off=0.5, ref idx=0)

Performs the calibration of a pseudo-2D experiment by circular-shifting the spectra of an appropriate amount. The target function aims to minimize the superimposition between a reference spectrum and the others using a brute-force method.

#### **Parameters:**

- ppm\_scale: 1darray ppm scale of the spectrum to calibrate
- data: *2darray*Complex-valued spectrum
- lims: *tuple* (ppm sx, ppm dx) of the calibration region
- u\_off: float
  Maximum offset for the circular shift, in ppm
- ref\_idx: *int*Index of the spectrum to be used as reference

#### Returns:

- data\_roll: 2darray Calibrated data
- u\_cal: *list*Number of point of which the spectra have been circular-shifted
- u\_cal\_ppm: *list* Correction for the ppm scale of each experiment

# 3.2.11 processing.baseline\_correction(ppm, data, basl\_file='spectrum.basl', winlim=None)

Interactively corrects the baseline of a given spectrum and saves the parameters in a file. The program starts with an interface to partition the spectrum in windows to correct separately. Then, for each window, an interactive panel opens to allow the user to compute the baseline.

#### Parameters:

• ppm: *1darray* PPM scale of the spectrum

• data: *1darray*The spectrum of which to adjust the baseline

• basl\_file: *str*Name for the baseline parameters file

• winlim: list or str or None
List of the breakpoints for the window. If it is str, it points to a file to be read with np.loadtxt.

If it is None, the partitioning is done interactively.

# 3.2.12 processing.calc\_nc(data, s\_n)

Calculates the optimal number of components, given the standard deviation of the noise. The threshold value is calculated as stated in Theorem 1 of reference: https://arxiv.org/abs/1710.09787v2

# Parameters:

• data: *2darray* Input data

• s\_n: *float*Noise standard deviation

# Returns:

• n\_c: *int* Number of components

# 3.2.13 processing.calibration(ppmscale, S)

Scroll the ppm scale of spectrum to make calibration. The interface offers two guidelines: the red one, labelled 'reference signal' remains fixed, whereas the green one ('calibration value') moves with the ppm scale.

The ideal calibration procedure consists in placing the red line on the signal you want to use as reference, and the green line on the ppm value that the reference signal must assume in the calibrated spectrum. Then, scroll with the mouse until the two lines are superimposed.

#### **Parameters**

• ppmscale: *1darray*The ppm scale to be calibrated

• S: *1darray*The spectrum to calibrate

#### Returns

• offset: *float*Difference between original scale and new scale. This must be summed up to the original ppm scale to calibrate the spectrum.

# 3.2.14 processing.convdta(data, grpdly=0, scaling=1)

Removes the digital filtering to obtain a spectrum similar to the command CONVDTA performed by TopSpin. However, they will differ a little bit because of the digitization. These differences are not invisible to human's eye.

## Parameters:

• data: *ndarray* FID with digital filter

• grpdly: int

Number of points that the digital filter consists of. Key \$GRPDLY in acqus file

• scaling: float
Scaling factor of the resulting FID. Needed to match TopSpin's intensities.

## Returns:

• data\_in: ndarray
FID without the digital filter. It will have grpdly points less than data.

# 3.2.15 processing.em(data, lb, sw)

Exponential apodization.

Being the FID an array of N points in its last dimension, and taken x = np.arange(N)/N:

$$\mathtt{apod} = \exp \biggl[ - \pi \frac{\mathtt{lb}}{2\mathtt{sw}} x \biggr]$$

# **Parameters**

- data : *ndarray* FID to be apodized
- lb : floatLorentzian broadening, in Hz. 1b > 0
- sw : *float* Spectral width in Hz, used for normalization.

## Returns

• apod \* data : ndarray Apodized FID.

# 3.2.16 processing.fp(data, wf=None, zf=None, fcor=0.5, tdeff=0)

Performs the full processing of a 1D NMR FID.

# **Parameters**

- data: *1darray*Data to be processed.
- wf: dict {'mode': function to be used, 'parameters': see window functions documentation}
- zf: *int* final size of spectrum
- fcor: *float*Weighting factor for the first point of the FID.
- tdeff: int

  Number of FID points to be employed for the processing. tdeff = 0 means whole FID.

## Returns

• data: *ndarray*Processed spectrum.

# 3.2.17 processing.ft(data, alt=False, fcor=0.5, Numpy=True)

Fourier transform in NMR sense. This means it returns the reversed spectrum.

#### **Parameters**

- data : *ndarray* FID to be Fourier-transformed.
- alt: bool
  Negates the sign of the odd points, then takes the complex conjugate. Required for States-TPPI processing.
- fcor: float
  Weighting factor for FID 1-st point. Default value (0.5) prevents baseline offset
- Numpy: bool

  If True (\*\*\*STRONGLY ADVISED\*\*\*) performs the FT using the FFT algorithm encoded in Numpy. Otherwise, performs it manually using the definition of discrete FT.

#### Returns

• dataft : ndarray Transformed FID.

# 3.2.18 processing.gm(data, lb, gb, sw, gc=0)

Gaussian apodization.

Being the FID an array of N points in its last dimension, and taken x = np.arange(N):

$$\mathrm{apod} = \exp\big[a - b^2\big], \qquad a = \pi \frac{\mathrm{lb}}{\mathrm{sw}} x, \qquad b = 0.6\pi \frac{\mathrm{gb}}{\mathrm{sw}} \big(\mathrm{gc}(N-1) - x\big)$$

## **Parameters**

- data : *ndarray* FID to be apodized
- lb : float Lorentzian broadening, in Hz. 1b < 0
- gb : float Gaussian broadening, in Hz. gb > 0
- ullet sw : float Spectral width in Hz, used for normalization.
- $\bullet$ gc: float Gaussian center, relatively to the FID length:  $0 \leq \texttt{gc} \leq 1$

### Returns

• apod \* data : ndarray Apodized FID.

# 3.2.19 processing.gmb(data, lb, gb, sw)

Gaussian apodization, Bruker-like. It does not work very well.

Being the FID an array of N points in its last dimension, and taken t = np.arange(N)/sw:

$$\mathrm{apod} = \exp\bigl[\alpha t - (\beta t)^2\bigr], \qquad \alpha = \pi \mathrm{lb}, \qquad \beta = \frac{-\alpha N}{2\,\mathrm{gb}\,\mathrm{sw}}\bigr)$$

# Parameters

- data : *ndarray* FID to be apodized
- lb: float
  Lorentzian broadening, in Hz. 1b < 0
- gb : floatGaussian broadening, in Hz. gb > 0
- sw: float Spectral width in Hz, used for normalization.
- gc: float Gaussian center, relatively to the FID length:  $0 \le gc \le 1$

#### Returns

• apod \* data : ndarray Apodized FID.

# 3.2.20 processing.ift(data, alt=False, fcor=0.5, Numpy=True)

Inverse Fourier transform in NMR sense. This means that the spectrum is reversed before to be inverse-Fourier transformed.

## **Parameters**

- data : *ndarray* FID to be inverse Fourier-transformed.
- alt: bool Negates the sign of the odd points, then takes the complex conjugate. Required for States-TPPI processing.
- fcor: float
  Weighting factor for FID 1-st point. Default value (0.5) prevents baseline offset
- Numpy: bool
  If True (\*\*\*STRONGLY ADVISED\*\*\*) performs the FT using the FFT algorithm encoded in Numpy. Otherwise, performs it manually using the definition of discrete FT.

#### Returns

• dataft : ndarray
Inverse transformed FID.

# 3.2.21 processing.integral(fx, x=None, lims=None)

Calculates the primitive of fx. If fx is a multidimensional array, the integrals are computed along the last dimension.

## Parameters:

- fx: *ndarray*Function (array) to integrate
- x: 1darray or None Independent variable. Determines the integration step. If None, it is set as the point scale
- lims: *tuple or None*Integration range. If None, the whole function is integrated

#### Returns:

• Fx: *ndarray* Integrated function.

Calculate the integral of a 2D peak. The idea is to extract the traces correspondent to the peak center and fit them with a gaussian function in each dimension. Then, once got the intensity of each of the two gaussians, multiply them together in order to obtain the 2D integral. This procedure should be equivalent to what CARA does.

#### Parameters:

- ppm\_f1: *1darray* PPM scale of the indirect dimension
- t\_f1: 1darray
  Trace of the indirect dimension, real part
- SFO1: float
  Larmor frequency of the nucleus in the indirect dimension
- ppm\_f2: *1darray* PPM scale of the direct dimension
- t\_f2: 1darray
  Trace of the direct dimension, real part
- SFO2: float
  Larmor frequency of the nucleus in the direct dimension
- u\_1: float Chemical shift in F1 /ppm. Defaults to the center of the scale
- fwhm\_1: float Starting FWHM /Hz in the indirect dimension
- utol\_1: float
  Allowed tolerance for u\_1 during the fit. (u\_1-utol\_1, u\_1+utol\_1)
- u\_2: float Chemical shift in F2 /ppm. Defaults to the center of the scale
- fwhm\_2: float Starting FWHM /Hz in the direct dimension
- utol\_2: float
  Allowed tolerance for u\_2 during the fit. (u\_2-utol\_2, u\_2+utol\_2)
- plot\_result: bool

  True to show how the program fitted the traces.

## Returns:

• I\_tot: *float* Computed integral.

# 3.2.23 processing.interactive basl windows(ppm, data)

Allows for interactive partitioning of a spectrum in windows. Double left click to add a bar, double right click to remove it. Returns the location of the red bars as a list.

# Parameters:

• ppm: *1darray* PPM scale of the spectrum

• data: *1darray* Spectrum to be partitioned

## Returns:

• coord: *list*List containing the coordinates of the windows, plus ppm[0] and ppm[-1]

# 3.2.24 processing.interactive echo param(data0)

Interactive plot that allows to select the parameters needed to process a CPMG-like FID. Use the TextBox or the arrow keys to adjust the values. You can call processing.sum\_echo\_train or processing.split\_echo\_train by starring the return statement of this function, i.e.:

processing.sum\_echo\_train(data0, \*interactive\_echo\_train(data0))

as they are in the correct order to be used in this way.

#### Parameters:

• data0: ndarray CPMG FID

## Returns:

- n: *int*Distance between one echo and the next one
- n\_echoes: *int* Number of echoes to sum/split
- i\_p: *int* Offset points from the start of the FID

# 3.2.25 processing interactive fp(fid0, acqus, procs)

Perform the processing of a 1D NMR spectrum interactively. The GUI offers the opportunity to test different window functions, as well as different tdeff values and final sizes. The active parameters appear as blue text.

## Parameters:

• fid0: *1darray* FID to process

• acqus: *dict*Dictionary of acquisition parameters

• procs: *dict*Dictionary of processing parameters

## Returns:

• pdata: *1darray* Processed spectrum

• procs: *dict*Updated dictionary of processing parameters

# 3.2.26 processing.interactive\_phase\_1D(ppmscale, S)

Allows for interactive phase adjustment of 1D NMR spectra. Employs processing.ps for the actual phase correction. Press the **z** key on the keyboard to toggle the automatic adjustment of vertical scale on or off.

## **Parameters**

• ppmscale: *1darray* PPM scale of the spectrum

• S: *1darray*The spectrum to be phased

## Returns

• phased\_data: *1darray* Phased spectrum.

# 3.2.27 processing.interactive phase 2D(ppm f1, ppm f2, S)

Interactively adjust the phases of a 2D spectrum. S must be hypercomplex, therefore must be passed before to unpack it into the 4 real files. The phase correction is done using processing.ps as follows:

- 1. S = processing.ps(S, p0=p0\_f2, p1=p1\_f2, pivot=pivot\_f2)
- 2. Transpose: normal if FnMODE='QF', hypercomplex otherwise
- 3.  $S = phase(S, p0=p0_f1, p1=p1_f1, pivot=pivot_f1)$
- 4. Transpose back

## **Parameters**

- ppm\_f1: *1darray*Indirect dimension ppm scale
- ppm\_f1: 1darray Direct dimension ppm scale
- S: 2darray Hypercomplex spectrum

#### Returns

• S:2darray Phased spectrum

# 3.2.28 processing.interactive qfil(ppm, data in)

Interactive function to design a gaussian filter with the aim of suppressing signals in the spectrum. You can adjust position and width of the filter scrolling with the mouse.

## Parameters:

• ppm: *1darray* Scale on which the filter will be built

• data\_in: *1darray*Spectrum on which to apply the filter.

## Returns:

• u: *float*Position of the gaussian filter

• s: float
Width of the gaussian filter (Standard deviation)

# $3.2.29 \quad \text{processing.interactive\_xfb} \\ \text{(fid0, acqus, procs, lvl0=0.1, show\_cnt=True)} \\$

Perform the processing of a 2D NMR spectrum interactively. The GUI offers the opportunity to test different window functions, as well as different tdeff values and final sizes. The active parameters appear as blue text. When changing the parameters, give it some time to compute. The figure panel is quite heavy.

## Parameters:

- fid0: 2darray
  FID to process
- acqus: *dict*Dictionary of acquisition parameters
- procs: *dict*Dictionary of processing parameters
- lvl0: float
  Starting level of the contours
- show\_cnt: bool
  Choose if to display data using contours (True) or heatmap (False)

#### Returns:

- pdata: *2darray*Processed spectrum
- procs: *dict*Updated dictionary of processing parameters

# 3.2.30 processing.inv fp(data, wf=None, size=None, fcor=0.5)

Performs the full inverse processing of a 1D NMR spectrum (data). Required parameters are:

## **Parameters**

- data: *1darray* Input data
- wf: dict {'mode': function to be used, 'parameters': different from each function}
- size: *int* initial size of the FID
- fcor: *float* weighting factor for the FID first point

#### Returns

• data: *1darray* Processed data

# 3.2.31 processing.inv\_xfb(data, wf=[None, None], size=[None, None], fcor=[0.5, 0.5], FnMODE='States-TPPI')

Performs the full processing of a 2D NMR FID (data). Required parameters are:

## **Parameters**

- data: *2darray*Input data
- wf: list of dict list of two entries [F1, F2]. Each entry is a dictionary of window functions
- zf: *list*list of two entries [zf F1, zf F2]
- fcor: *list* first fid point weighting factor [F1, F2]
- u: bool

  If True, unpacks the hypercomplex spectrum and returns the 4 real files, using processing.unpack\_2D
- tdeff: list of int number of points of the FID to be used for the processing, [F1, F2]

#### Returns

• data: *2darray* Processed data

# 3.2.32 processing.iterCadzow(data, n, nc, itermax=100, f=0.005, print\_head=T print\_time=True)

This functions performs Cadzow denoising on  $\mathtt{data}$ , which is a 1D array of N points, in an iterative manner. The algorithm works as follows:

- 1. Transform data in a Hankel matrix  $\mathbb{H}$  of dimensions (N-n,n)
- 2. Make SVD on  $\mathbb{H} = \mathbb{USV}^{\dagger}$
- 3. Keep only the first nc singular values, and put all the rest to  $0 (S \to S')$
- 4. Rebuild  $\mathbb{H}' = \mathbb{US}'\mathbb{V}^{\dagger}$
- 5. Average the antidiagonals to rebuild the Hankel-type structure, then make 1D array
- 6. Check arrest criterion: if it is not reached, go to step 1, otherwise exit from the cycle and return the processed data.

The arrest criterion is on the array of singular values S, which is the main diagonal of the matrix  $\mathbb{S}$ . At step k and Python indexing system:

$$\left|\frac{S^{(k-1)}[\text{nc}-1]}{S^{(k-1)}[\textbf{0}]} - \frac{S^{(k)}[\text{nc}-1]}{S^{(k)}[\textbf{0}]}\right| < f\frac{S^{(0)}[\text{nc}-1]}{S^{(0)}[\textbf{0}]}$$

## **Parameters**

- data: *1darray* Input data
- n: *int*Number of columns of the Hankel matrix.
- nc: *int*Number of singular values to keep.
- itermax: *int*Maximum number of iterations allowed.
- f: float
  Factor for the arrest criterion.
- print\_head: *bool*Set it to True to display the fancy heading.
- print\_time: bool
  Set it to True to display the time spent.

#### Returns

• datap: *1darray* Denoised data

# 3.2.33 processing.load baseline(filename, ppm, data)

Read the baseline parameters from a file and builds the baseline itself.

#### Parameters:

 $\bullet$  filename: str Location of the baseline file

• ppm: *1darray* PPM scale of the spectrum

• data: *1darray*Spectrum of which to correct the baseline

## Returns:

• baseline: *1darray* Computed baseline

# 3.2.34 processing.make polynomion baseline(ppm, data, limits)

Interactive baseline correction with 4th degree polynomion.

#### Parameters:

- ppm: *1darray* PPM scale of the spectrum
- data: *1darray* spectrum
- limits: *tuple*Window limits (left, right).

## Returns:

- mode: str Baseline correction mode: 'polynomion' as default, 'spline' if you press the button
- $\bullet$  C\_f: 1darray or str Baseline polynomion coefficients, or 'callintsmooth' if you press the spline button

# 3.2.35 processing.make scale(size, dw, rev=True)

Computes the frequency scale of the NMR spectrum, given the number of points and the employed dwell time (the REAL one, not the TopSpin one!). rev=True is required for the correct frequency arrangement in the NMR sense.

## **Parameters**

ullet size: int Number of points of the frequency scale

• dw: *float*Time spacing in the time dimension

• rev: bool Reverses the scale

#### Returns

• fqscale: *1darray*The computed frequency scale.

3.2.36 processing.new\_MCR(input\_data, nc, f=10, tol=1e-05, itermax=10000. H=True, oncols=True, our\_function=None, fargs=[], our\_function2=None f2args=[])

# This is an implementation of Multivariate Curve Resolution # for the denoising of 2D NMR data. It requires: # - input\_data: a tensor containing the set of 2D NMR datasets to be coprocessed # stacked along the first dimension; # - nc: number of purest components; # - f: percentage of allowed noise; # - tol: tolerance for the arrest criterion; # - itermax: maximum number of allowed iterations, default  $10000 \ \#$  - H: True for horizontal stacking of data (default), False for vertical; # - oncols: True to estimate S with purest components, False to estimate C # This function returns the denoised datasets, 'CS', and the 'C' and 'S' matrices.

 $3.2.37 \quad processing.new\_MCR\_ALS(D,~C,~S,~itermax=10000,~tol=1e-05,\\ reg\_f=None,~reg\_fargs=[])$ 

Modified function to do ALS

# 3.2.38 processing.pknl(data, grpdly=0, onfid=False)

Compensate for the Bruker group delay at the beginning of FID through a first-order phase correction of

$$\phi^{(1)} = 360*\mathrm{GRPDLY}$$

This should be applied after apodization and zero-filling.

# Parameters:

- data: *ndarray*Input data. Be sure it is complex!
- grpdly: *int*Number of points that make the group delay.
- onfid: bool

  If it is True, performs FT before to apply the phase correction, and IFT after.

## Returns:

• datap: ndarray Corrected data

# 3.2.39 processing.ps(data, ppmscale=None, p0=None, p1=None, pivot=None, interactive=False)

Applies phase correction on the last dimension of data. The pivot is set at the center of the spectrum by default. Missing parameters will be inserted interactively.

Being data a 1D array of N points, as well as ppmscale, the following parameters are defined:

$$p_V = (\text{index of pivot on ppmscale})/N$$
  
 $x = \text{np.arange}(N)/N - p_V$ 

Basically,  $p_V$  and x are pivot and ppmscale, normalized in order to fit a scale that ranges from 0 to 1. This is done with the aim to 'standardize' the behaviour of the function, making it nucleus-independent.

The following phase-correction function is applied:

$$\Phi = \exp \left[ \mathrm{i} (\mathtt{p0} + \mathtt{p1} x) \right]$$

If data is a 2D array, the function  $\Phi$  is applied row-wise.

## Parameters:

- data: *ndarray* Input data
- ppmscale: 1darray or None PPM scale of the spectrum. Required for pivot and interactive phase correction
- p0: float
  Zero-order phase correction angle /degrees
- p1: float
  First-order phase correction angle /degrees
- pivot: *float or None*.

  First-order phase correction pivot /ppm. If None, it is set at the center of the spectrum.
- interactive: bool

  If True, all the parameters will be ignored and the interactive phase correction panel will be opened.

#### Returns:

- datap: ndarray
  Phased data
- final\_values: tuple
  Employed values of the phase correction. (p0, p1, pivot)

# 3.2.40 processing.qfil(ppm, data, u, s)

Suppress signals in the spectrum using a gaussian filter.

#### Parameters:

• ppm: 1darray
Scale on which to build the filter

ullet data: ndarray Data to be processed. The filter is applied on the last dimension

• u: *float*Position of the filter

• s: *float*Width of the filter (standard deviation)

## Returns:

• pdata: ndarray Filtered data

# 3.2.41 processing.qpol(fid)

Fits the FID with a 4-th degree polynomion, then subtracts it from the original FID. The real and imaginary channels are treated separately.

# Parameters

• fid : ndarray Self-explanatory.

# Returns

• fid : ndarray
Processed FID.

# 3.2.42 processing.qsin(data, ssb)

Sine-squared apodization.

Being data an array of N points in its last dimension, and taken x = np.arange(N)/N, if ssb = 0 or ssb = 1:

$$\mathtt{apod} = \sin^2\!\left(\pi\,x\right)$$

else if  $ssb \geq 2$ :

$$\mathtt{apod} = \sin^2 \biggl( \pi \frac{1}{\mathtt{ssb}} + \pi \bigl( 1 - \frac{1}{\mathtt{ssb}} \bigr) x \biggr)$$

#### **Parameters**

- data : *ndarray* FID to be apodized.
- ssb : *int* Shifting parameter for the sine bell.

- ssb = 0: from 0 to  $\pi$ 

- ssb = 1 : same as ssb = 0

- ssb = 2: from  $\pi/2$  to  $\pi$ 

- ssb = 3: from  $\pi/3$  to  $\pi$ 

# Returns

• apod \* data : ndarray Apodized FID.

# 3.2.43 processing.quad(fid)

Subtracts from the FID the arithmetic mean of its last quarter. The real and imaginary channels are treated separately.

# **Parameters**

• fid : ndarray Self-explanatory.

# Returns

• fid : ndarray
Processed FID.

# 3.2.44 processing.repack\_2D(rr, ir, ri, ii)

Renconstruct hypercomplex 2D NMR data given the 4 real arrays. See  $processing.unpack_2D$  for details.

# Returns

• data

## 3.2.45 processing.rev(data)

Reverses the last dimension of data.

### **Parameters**

• data : *ndarray* matrix to be reverted

### Returns

 $\bullet \ \, \mathrm{data[...,::-1]}: \ \, ndarray \\ \mathrm{Self-explanatory}$ 

# 3.2.46 processing.rpbc(data, split\_imag=False, n=5, basl\_method='tq', basl\_thresh=0.1, basl\_itermax=500, \*\*phase\_kws)

Reversed Phase and Baseline Correction. Allows for the automatic phase correction and baseline subtraction of NMR spectra. It is called 'reversed' because the baseline is actually computed and subtracted before to perform the phase correction.

The baseline is computed using a low-order polynomion, built on a scale that goes from -1 to 1, whose coefficients are obtained minimizing a non-quadratic cost function. It is recommended to use either 'tq' (truncated quadratic, much faster) or 'huber' (Huber function, slower but sometimes more accurate). The user is requested to choose between separating the real and imaginary channel in this step. The order of the polynomion and the threshold value are the key parameters for obtaining a good baseline. The used function is processing polyn basl

The phase correction is computed on the baseline-subtracted complex data as described in the SINC algorithm (M. Sawall et al., *Journal of Magnetic Resonance* 289 (2018), 132-141). The default parameters are generally fine, but in case of data with poor SNR (approximately SNR < 10) better results can be obtained by increasing the value of the e1 parameter. The employed function is processing.SINC phase

### **Parameters:**

- data: *1darray*Data to be processed, complex-valued
- split\_imag: bool

  If True, computes the baseline on the real and imaginary part separately; else, the set of polynomion coefficients are forced to be the same for both
- $\bullet$  n: intNumber of coefficients of the polynomion, i.e. it will be of degree n-1
- basl\_method: str Cost function to be minimized for the baseline computation. Look for fit.CostFunc, 'method' attribute
- basl\_thresh: float Relative threshold value for the non-quadratic behaviour of the cost function. Look for fit.CostFunc, 's' attribute
- basl\_itermax: int Maximun number of iterations allowed during the baseline fitting procedure
- phase\_kws: keyworded arguments
  Optional arguments for the phase correction. Look for fit.SINC\_phase keyworded arguments
  for details.

### Returns:

- y: *1darray*Processed data
- p0: float
  Zero-order phase correction angle, in degrees
- p1: float
  First-order phase correction angle, in degrees

• c: 1darray

Set of coefficients to be used for the baseline computation, starting from the 0-order coefficient

## 3.2.47 processing.sin(data, ssb)

Sine apodization.

Being data an array of N points in its last dimension, and taken x = np.arange(N)/N, if ssb = 0 or ssb = 1:

$$\mathtt{apod} = \sin \biggl( \pi \, x \biggr)$$

else if  $ssb \ge 2$ :

$$\mathtt{apod} = \sin \left( \pi \frac{1}{\mathtt{ssb}} + \pi \left( 1 - \frac{1}{\mathtt{ssb}} \right) x \right)$$

### **Parameters**

- data : *ndarray* FID to be apodized.
- ssb: *int* Shifting parameter for the sine bell.

- ssb = 0: from 0 to  $\pi$ 

- ssb = 1 : same as ssb = 0

- ssb = 2: from  $\pi/2$  to  $\pi$ 

- ssb = 3: from  $\pi/3$  to  $\pi$ 

### Returns

• apod \* data : ndarray Apodized FID.

### 3.2.48 processing.split echo train(datao, n, n echoes, i p=0)

Separate a CPMG echo-train FID into echoes so to be processed separately. The first decay, i.e. the native FID, is extracted, and corresponds to echo number 0. Then, for each echo, the left side (reversed) is summed up to its right part. You can use processing.interactive\_echo\_param to calculate the parameters interactively.

### Parameters:

- datao: *ndarray* FID with an echo train on its last dimension
- $\bullet$  n: int number of points that separate one echo from the next
- n\_echoes: *int* number of echoes to extract. If it is 0, extracts only the first decay
- i\_p: int Number of offset points

### **Returns:**

• data\_p: (n+1)darray Separated echoes

## 3.2.49 processing.stack MCR(input data, H=True)

Performs matrix augmentation converting input\_data from dimensions (X,Y,Z) to (Y,X\*Z) if H=True, or (X\*Y,Z) if H=False.

### **Parameters**

- input\_data: 3darray Contains the spectra to be stacked together. The index that runs on the datasets must be the first one.
- H: bool

  True for horizontal stacking, False for vertical stacking.

### Returns

• data: *2darray*Augmented data matrix.

## 3.2.50 processing.sum\_echo\_train(datao, n, n\_echoes, i\_p=0)

Sum up a CPMG echo-train FID into echoes so to enchance the SNR. This function calls processing.split\_e with the same parameters. You can use processing.interactive\_echo\_param to calculate the parameters interactively.

### Parameters:

- datao: *ndarray* FID with an echo train on its last dimension
- $\bullet$  n: int number of points that separate one echo from the next
- n\_echoes: *int* number of echoes to sum
- i\_p: int Number of offset points

### Returns:

• data\_p: ndarray Summed echoes

## 3.2.51 processing.tabula rasa(data, lvl=0.05, cmap=cm.Blues r)

This function is to be used in SIFT algorithm. Allows interactive selection using a Lasso widget of the region of the spectrum which contain signal. Returns a masking matrix, of the same shape as data, whose entries are 1 inside the selection and 0 outside.

### **Parameters**

• data : *2darray*The data to be plotted

• lvl : *float*Level of the contours, expressed as fraction of the maximum intensity

• cmap : matplotlib.cm Object Color of the contours

### Returns

• mask: 2darray
Matrix that contains 1 inside the selection and 0 outside.

## 3.2.52 processing.td eff(data, tdeff)

Uses only the first tdeff points of data. It is applied before any other processing. The length of the list tdeff must match the dimension of data, if data is multidimensional.

### **Parameters**

• data: *ndarray*The FID to be trimmed.

• tdeff: int or list

The number of points to be used. If data is 1D, tdeff can be passed as integer, otherwise it has to be: [F1, F2, ..., Fn]. A 0 entry in the list means to not trim the corresponding dimension.

### Returns

• data: *ndarray*Trimmed data according to tdeff.

### 3.2.53 processing.tp hyper(data)

Computes the hypercomplex transpose of data. Needed for the processing of data acquired in a phase-sensitive manner in the indirect dimension.

To explain how a hypercomplex transposition works, we will focus on the simpler example of a  $4 \times 6$  matrix,  $\mathbb{A}$ , with complex coefficients  $a_{i,j} = x_{i,j} + iy_{i,j}$ .

$$\mathbb{A} = \begin{pmatrix} x_{1,1} + \mathrm{i}y_{1,1} & x_{1,2} + \mathrm{i}y_{1,2} & x_{1,3} + \mathrm{i}y_{1,3} & x_{1,4} + \mathrm{i}y_{1,4} \\ x_{2,1} + \mathrm{i}y_{2,1} & x_{2,2} + \mathrm{i}y_{2,2} & x_{2,3} + \mathrm{i}y_{2,3} & x_{2,4} + \mathrm{i}y_{2,4} \\ x_{3,1} + \mathrm{i}y_{3,1} & x_{3,2} + \mathrm{i}y_{3,2} & x_{3,3} + \mathrm{i}y_{3,3} & x_{3,4} + \mathrm{i}y_{3,4} \\ x_{4,1} + \mathrm{i}y_{4,1} & x_{4,2} + \mathrm{i}y_{4,2} & x_{4,3} + \mathrm{i}y_{4,3} & x_{4,4} + \mathrm{i}y_{4,4} \\ x_{5,1} + \mathrm{i}y_{5,1} & x_{5,2} + \mathrm{i}y_{5,2} & x_{5,3} + \mathrm{i}y_{5,3} & x_{5,4} + \mathrm{i}y_{5,4} \\ x_{6,1} + \mathrm{i}y_{6,1} & x_{6,2} + \mathrm{i}y_{6,2} & x_{6,3} + \mathrm{i}y_{6,3} & x_{6,4} + \mathrm{i}y_{6,4} \end{pmatrix}$$

The hypercomplex transpose of the matrix  $\mathbb{A}$ , here referred as  $\mathbb{B}$ , is:

$$\mathbb{B} = \begin{pmatrix} x_{1,1} + \mathrm{i} x_{2,1} & x_{3,1} + \mathrm{i} x_{4,1} & x_{5,1} + \mathrm{i} x_{6,1} \\ y_{1,1} + \mathrm{i} y_{2,1} & y_{3,1} + \mathrm{i} y_{4,1} & y_{5,1} + \mathrm{i} y_{6,1} \\ x_{1,2} + \mathrm{i} x_{2,2} & x_{3,2} + \mathrm{i} x_{4,2} & x_{5,2} + \mathrm{i} x_{6,2} \\ y_{1,2} + \mathrm{i} y_{2,2} & y_{3,2} + \mathrm{i} y_{4,2} & y_{5,2} + \mathrm{i} y_{6,2} \\ x_{1,3} + \mathrm{i} x_{2,3} & x_{3,3} + \mathrm{i} x_{4,3} & x_{5,3} + \mathrm{i} x_{6,3} \\ y_{1,3} + \mathrm{i} y_{2,3} & y_{3,3} + \mathrm{i} y_{4,3} & y_{5,3} + \mathrm{i} y_{6,3} \\ x_{1,4} + \mathrm{i} x_{2,4} & x_{3,4} + \mathrm{i} x_{4,4} & x_{5,4} + \mathrm{i} x_{6,4} \\ y_{1,4} + \mathrm{i} y_{2,4} & y_{3,4} + \mathrm{i} y_{4,4} & y_{5,4} + \mathrm{i} y_{6,4} \end{pmatrix}$$

#### **Parameters**

• data : *2darray*Complex data matrix.

### Returns

• datatp: 2darray
Hypercomplex transpose of data.

## 3.2.54 processing.unpack 2D(data)

Separates fully processed 2D NMR data into 4 distinct ser files:

```
\begin{split} &\text{rr} = \operatorname{Re}\{\operatorname{\mathtt{data}}\}[::2] \\ &\text{ir} = \operatorname{Im}\{\operatorname{\mathtt{data}}\}[::2] \\ &\text{ri} = \operatorname{Re}\{\operatorname{\mathtt{data}}\}[1::2] \\ &\text{ii} = \operatorname{Im}\{\operatorname{\mathtt{data}}\}[1::2] \end{split}
```

### **Parameters**

• data : *2darray* Complex data matrix.

### Returns

• rr, ir, ri, ii: 2darray, 2darray, 2darray, 2darray See above.

## 3.2.55 processing.whittaker smoother(data, n=2, s f=1, w=None)

Adapted from P.H.C. Eilers, *Anal. Chem* 2003, 75, 3631-3636. Implementation of the smoothing algorithm proposed by Whittaker in 1923.

### Parameters:

- data: *1darray*Data to be smoothed
- n: *int* Order of the difference to be computed
- s\_f: *float* Smoothing factor
- w: 1darray or None
  Array of weights. If None, no weighting is applied.

### Returns:

• z: 1darray Smoothed data

## 3.2.56 processing.write basl info(f, limits, mode, data)

Writes the baseline parameters of a certain window in a file.

- f: TextIO object
  File where to write the parameters
- limits: *tuple*Limits of the spectral window. (left, right)
- ullet mode: str Baseline correction mode: 'polynomion' or 'spline'
- data: float or 1darray
  It can be either the spline smoothing factor or the polynomion coefficients

# 3.2.57 processing.xfb(data, wf=[None, None], zf=[None, None], fcor=[0.5,0.5], tdeff=[0,0], u=True, FnMODE='States-TPPI')

Performs the full processing of a 2D NMR FID.

### **Parameters**

- data: *2darray* FID data matrix
- wf: list
  List of two entries [F1, F2]. Each entry is a dictionary: {'mode': function to be used, 'parameters': different from each function}
- zf: *list*List of two entries [zf F1, zf F2]
- fcor: *list*Weighting factor for the first point of the FID. [F1, F2]
- tdeff: *list*Number of FID points to be employed for the processing. tdeff = [0,0] means whole FID.
  [F1, F2]
- u: bool if True, unpacks the hypercomplex spectrum and returns the 4 files.

### Returns

• data : 2darray Processed data.

## 3.2.58 processing.zf(data, size)

Zero-filling of data up to size.

### **Parameters**

• data : *ndarray* FID / Spectrum to be zero-filled

### Returns

## 3.3 FIGURES package

This package contains a series of functions to make plots of various nature.

3.3.1 figures.ax1D(ax, ppm, datax, norm=False, xlims=None, ylims=None, c='tab:blue', lw=0.5, X\_label='\$ delta \$ F1 /ppm', Y\_label='Intensity /a.u.', n\_xticks=10, n\_yticks=10, label=None, fontsize=10)

Makes the figure of a 1D NMR spectrum, placing it in a given figure panel. This allows the making of modular figures.

The plot can be customized in a very flexible manner by setting the function keywords properly.

- ax: matplotlib.subplot Object panel where to put the figure
- ppm: *1darray* ppm scale of the spectrum
- data: *1darray* spectrum to be plotted
- norm: bool if True, normalizes the intensity to 1.
- xlims: *list or tuple*Limits for the x-axis. If None, the whole scale is used.
- ylims: *list or tuple*Limits for the y-axis. If None, the whole scale is used.
- c: str Colour of the line.
- lw: float linewidth
- X\_label: str text of the x-axis label;
- Y\_label: str text of the y-axis label;
- n\_xticks: *int* Number of numbered ticks on the x-axis of the figure
- n\_yticks: *int* Number of numbered ticks on the x-axis of the figure
- label: str label to be put in the legend.

• fontsize: *float*Biggest font size in the figure.

## Returns:

• line: Line2D Object Line object returned by plt.plot.

```
3.3.2 figures.ax2D(ax, ppm_f2, ppm_f1, datax, xlims=None, ylims=None, cmap='Greys_r', c_fac=1.4, lvl=0.1, lw=0.5, X_label='$ delta ,$F2 /ppm', Y_label='$ delta ,$F1 /ppm', title=None, n xticks=10, n yticks=10, fontsize=10)
```

Makes a 2D contour plot like the one in figures.figure2D, but in a specified panel. Allows for the buildup of modular figures. The contours are drawn according to the formula:

```
cl = contour_start * contour_factor ** np.arange(contour_num)
where
contour_start = np.max(data) * lvl
contour_num = 16
contour_factor = c_fac.
```

Increasing the value of  $c_fac$  will decrease the number of contour lines, whereas decreasing the value of  $c_fac$  will increase the number of contour lines.

- ax: matplotlib.subplot Object panel where to put the figure
- ppm\_f2: 1darray ppm scale of the direct dimension
- ppm\_f1: *1darray* ppm scale of the indirect dimension
- datax: 2darray the 2D NMR spectrum to be plotted
- xlims: *tuple* limits for the x-axis (left, right). If None, the whole scale is used.
- ylims: *tuple* limits for the y-axis (left, right). If None, the whole scale is used.
- cmap: strColormap identifier for the contour
- c\_fac: *float*Contour factor parameter
- lvl: float
  height with respect to maximum at which the contour are computed
- X\_label: str text of the x-axis label;
- Y\_label: str text of the y-axis label;

- lw: *float* linewidth of the contours
- title: str Figure title.
- n\_xticks: *int* Number of numbered ticks on the x-axis of the figure
- $\bullet\,$  n\_yticks: int Number of numbered ticks on the x-axis of the figure
- fontsize: *float*Biggest font size in the figure.

### Returns:

ullet cnt:  $matplotlib.QuadContour\ object$  Drawn contour lines

3.3.3 figures.ax\_heatmap(ax, data, zlim='auto', z\_sym=True, cmap=None, xscale=None, yscale=None, rev=(False, False), n\_xticks=10, n\_yticks=1 n\_zticks=10, fontsize=10)

Computes a heatmap of data on the given 'ax'

### Parameters:

- ax: matplotlib.Subplot object
  Panel where to draw the heatmap
- data: 2darray Input data
- zlim: tuple or 'auto' or 'abs'

  Vertical limits of the heatmap, that determines the extent of the colorbar. 'auto' means (min(data), max(data)), 'abs' means (min(|data|), max(|data|)).
- z\_sym: bool
  True to symmetrize the vertical scale around 0.
- cmap: matplotlib.cm object Colormap of the heatmap.
- xscale: *1darray or None* x-scale. None means np.arange(data.shape[1])
- yscale: *1darray or None* y-scale. None means np.arange(data.shape[0])
- rev: tuple of bool Reverse scale (x, y).
- n\_xticks: *int*Number of ticks of the x axis
- n\_yticks: *int*Number of ticks of the y axis
- n\_zticks: *int* Number of ticks of the color bar
- fontsize: *float*Biggest font size to apply to the figure.

### Returns:

- im: matplotlib.AxesImageThe heatmap
- cax: figure panel where the colorbar is drawn

## 3.3.4 figures.dotmd(ppmscale, S, labels=None, n xticks=10)

Interactive display of multiple 1D spectra.

- ppmscale: *1darray or list* ppm scale of the spectra. If only one scale is supplied, all the spectra are plotted using the same scale. Otherwise, each spectrum is plotted using its scale. There is a 1:1 correspondence between ppmscale and S.
- S: *list* spectra to be plotted
- labels: *list* labels to be put in the legend.
- n\_xticks: *int* Number of numbered ticks on the x-axis of the figure

 $\begin{array}{lll} 3.3.5 & figures.dotmd\_2D(ppm\_f1,ppm\_f2,S,labels=None,name='dotmd\_2D\\ & X\_label='\$\backslash delta\backslash,\ \$\ F2\ /ppm',\ Y\_label='\$\backslash delta\backslash,\ \$\ F1\ /ppm',\\ & n\_xticks=10,\ n\_yticks=10,\ Neg=True) \end{array}$ 

Interactive display of multiple 2D spectra. They have to share the same scales.

- ppm\_f1: 1darray ppm scale of the indirect dimension. If only one scale is supplied, all the spectra are plotted using the same scale. Otherwise, each spectrum is plotted using its scale. There is a 1:1 correspondance between ppm\_f1 and S.
- ppm\_f2: 1darray ppm scale of the direct dimension. If only one scale is supplied, all the spectra are plotted using the same scale. Otherwise, each spectrum is plotted using its scale. There is a 1:1 correspondance between ppm\_f2 and S.
- S: *list* spectra to be plotted
- labels: *list* labels to be put in the legend.
- name: *str*If you choose to save the figure, this is its filename.
- X\_label: str text of the x-axis label;
- Y\_label: str text of the y-axis label;
- n\_xticks: *int*Number of numbered ticks on the x-axis of the figure
- n\_yticks: *int* Number of numbered ticks on the x-axis of the figure
- Neg: *bool*If True, show the negative contours.

3.3.6 figures.figure1D(ppm, datax, norm=False, xlims=None, ylims=None, c='tab:blue', lw=0.5, X\_label='\$ delta \$ F1 /ppm', Y\_label='Intensity /a.u.', n\_xticks=10, n\_yticks=10, fontsize=10, name=None, ext='png', dpi=600)

Makes the figure of a 1D NMR spectrum.

The plot can be customized in a very flexible manner by setting the function keywords properly.

- ppm: *1darray* ppm scale of the spectrum
- datax: *1darray* spectrum to be plotted
- norm: bool if True, normalizes the intensity to 1.
- xlims: *list or tuple*Limits for the x-axis. If None, the whole scale is used.
- ylims: *list or tuple*Limits for the y-axis. If None, the whole scale is used.
- c: str Colour of the line.
- lw: float linewidth
- X\_label: str text of the x-axis label;
- Y\_label: str text of the y-axis label;
- n\_xticks: *int* Number of numbered ticks on the x-axis of the figure
- n\_yticks: *int* Number of numbered ticks on the x-axis of the figure
- fontsize: *float*Biggest font size in the figure.
- name: str or None
  Filename for the figure to be saved. If None, the figure is shown instead.
- ext: strFormat of the image
- dpi: *int*Resolution of the image in dots per inches

3.3.7 figures.figure1D\_multi(ppm0, data0, xlims=None, ylims=None, norm=Fac=None, X\_label='\$ delta \$ F1 /ppm', Y\_label='Intensity /a.u.', n\_xticks=10, n\_yticks=10, fontsize=10, labels=None, name=None, ext='png', dpi=600)

Creates the superimposed plot of a series of 1D NMR spectra.

- ppm0: sequence of 1darray or 1darray ppm scale of the spectra. If only one scale is supplied, it is assumed to be the same for all the spectra
- data0: sequence of 1darray
  List containing the spectra to be plotted
- xlims: tuple or None Limits for the x-axis. If None, the whole scale is used.
- ylims: tuple or None Limits for the y-axis. If None, they are automatically set.
- norm: False or float or str

  If it is False, it does nothing. If it is float, divides all spectra for that number. If it is str('#'), normalizes all the spectra to the '#' spectrum (python numbering). If it is whatever else string, normalizes all spectra to themselves.
- c: tuple or None
  List of the colors to use for the traces. None uses the default ones.
- X\_label: str text of the x-axis label
- Y\_label: str text of the y-axis label
- n\_xticks: *int* Number of numbered ticks on the x-axis of the figure
- n\_yticks: *int* Number of numbered ticks on the x-axis of the figure
- fontsize: *float*Biggest fontsize in the picture
- labels: list or None or False
  List of the labels to be shown in the legend. If it is None, the default entries are used (i.e., '1, 2, 3,...'). If it is False, the legend is not shown.
- name: str or None
  Filename of the figure, if it has to be saved. If it is None, the figure is shown instead.
- ext: strFormat of the image

 3.3.8 figures.figure2D(ppm\_f2, ppm\_f1, datax, xlims=None, ylims=None, cmap='Greys\_r', c\_fac=1.4, lvl=0.09, X\_label='\$ delta \$ F2 /ppm', Y\_label='\$ delta \$ F1 /ppm', lw=0.5, cmapneg=None, n\_xticks=10, n\_yticks=10, fontsize=10, name=None, ext='png', dpi=600)

Makes a 2D contour plot. Allows for the buildup of modular figures. The contours are drawn according to the formula:

```
cl = contour_start * contour_factor ** np.arange(contour_num)
where
contour_start = np.max(data) * lvl
contour_num = 16
contour_factor = c_fac.
```

Increasing the value of c\_fac will decrease the number of contour lines, whereas decreasing the value of c\_fac will increase the number of contour lines.

- ppm\_f2: 1darray ppm scale of the direct dimension
- ppm\_f1: 1darray ppm scale of the indirect dimension
- datax: 2darray the 2D NMR spectrum to be plotted
- xlims: *tuple* limits for the x-axis (left, right). If None, the whole scale is used.
- ylims: *tuple* limits for the y-axis (left, right). If None, the whole scale is used.
- cmap: str Colormap identifier for the contour
- c\_fac: *float* Contour factor parameter
- lvl: *float* height with respect to maximum at which the contour are computed
- X\_label: str text of the x-axis label;
- Y\_label: str text of the y-axis label;

- lw: *float* linewidth of the contours
- cmapneg: str or None Colormap identifier for the negative contour. If None, they are not computed at all
- n\_xticks: *int* Number of numbered ticks on the x-axis of the figure
- n\_yticks: *int* Number of numbered ticks on the x-axis of the figure
- fontsize: *float*Biggest font size in the figure.
- name: str Filename for the figure
- ext: strFormat of the image
- dpi: *int*Resolution of the image in dots per inches

 $\begin{array}{lll} 3.3.9 & \text{figures.figure2D\_multi(ppm\_f2, ppm\_f1, datax, xlims=None, ylims=None, lvl='default', c\_fac=1.4, Negatives=False, X\_label='\$ & \text{delta} & \text{F2 /ppm', Y\_label='\$} & \text{delta} & \text{F1 /ppm', lw=0.5, n\_xticks=10, n\_yticks=10, labels=None, name=Nonext='png', dpi=600)} \\ \end{array}$ 

Generates the figure of multiple, superimposed spectra, using figures.ax2D.

- ppm\_f2: *1darray* ppm scale of the direct dimension
- ppm\_f1: *1darray* ppm scale of the indirect dimension
- datax: sequence of 2darray the 2D NMR spectra to be plotted
- xlims: *tuple* limits for the x-axis (left, right). If None, the whole scale is used.
- ylims: *tuple* limits for the y-axis (left, right). If None, the whole scale is used.
- lvl: 'default' or list height with respect to maximum at which the contour are computed. If 'default', each spectrum is at 10% of maximum height. Otherwise, each entry of the list corresponds to the contour height of the respective spectrum.
- c\_fac: *float*Contour factor
- Negatives: bool set it to True if you want to see the negative part of the spectrum
- X\_label: str text of the x-axis label;
- Y\_label: *str* text of the y-axis label;
- lw: *float* linewidth of the contours
- n\_xticks: *int* Number of numbered ticks on the x-axis of the figure
- n\_yticks: *int* Number of numbered ticks on the x-axis of the figure
- labels: *list* entries of the legend. If None, the spectra are numbered.

ullet name: str

Filename for the figure. If None, it is shown instead of saved

 $\bullet$  ext: str

Format of the image

 $\bullet$  dpi: int

Resolution of the image in dots per inches

3.3.10 figures.fitfigure(S, ppm\_scale, t\_AQ, V, C=False, SFO1=701.125, o1p=0, limits=None, s\_labels=None, X\_label='\$ delta , \$F1 /ppm', n xticks=10, name=None)

Makes the figure to show the result of a quantitative fit.

- S: 1darray
  Spectrum to be fitted
- ppm\_scale : 1darray Self-explanatory
- V : 2darray matrix (# signals, parameters)
- C: 1darray or False

  Coefficients of the polynomion to be used as baseline correction. If the 'baseline' checkbox in the interactive figure panel is not checked, C\_f is False.
- limits: tuple or None
  Trim limits for the spectrum (left, right). If None, the whole spectrum is used.
- s\_labels: list or None or False Legend entries for the single components. If None, they are computed automatically as 1, 2, 3, etc. If False, they are not shown in the legend.
- X\_label : str label for the x-axis.
- n\_xticks: int number of numbered ticks that will appear in the ppm scale. An oculated choice can be very satisfying.
- name: str or None

  Name with which to save the figure. If None, the picture is shown instead of being saved.

3.3.11 figures.heatmap(data, zlim='auto', z\_sym=True, cmap=None, xs-cale=None, yscale=None, rev=(False, False), n\_xticks=10, n\_yticks=10 n\_zticks=10, fontsize=10, name=None)

Computes a heatmap of data.

- data: 2darray Input data
- zlim: tuple or 'auto' or 'abs'
  Vertical limits of the heatmap, that determines the extent of the colorbar. 'auto' means (min(data), max(data)), 'abs' means(min(|data|), max(|data|)).
- z\_sym: bool

  True to symmetrize the vertical scale around 0.
- cmap: matplotlib.cm object Colormap of the heatmap.
- xscale: *1darray or None* x-scale. None means np.arange(data.shape[1])
- yscale: *1darray or None* y-scale. None means np.arange(data.shape[0])
- rev: tuple of bool Reverse scale (x, y).
- n\_xticks: *int*Number of ticks of the x axis
- n\_yticks: *int*Number of ticks of the y axis
- n\_zticks: *int*Number of ticks of the color bar
- fontsize: *float*Biggest font size to apply to the figure.
- name: str or None Filename for the figure. Set to None to show the figure.

## 3.3.12 figures.plot\_fid(fid, name=None, ext='png', dpi=600)

Makes a two-panel figure that shows on the left the real part of the FID, on the right the imaginary part. The x-scale and y-scale are automatically adjusted.

# 3.3.13 figures.plot\_fid\_re(fid, scale=None, c='b', lims=None, name=None, ext='png', dpi=600)

Makes a single-panel figure that shows either the real or the imaginary part of the FID. The x-scale and y-scale are automatically adjusted.

- fid: *ndarray*FID to be plotted
- scale: *1darray or None* x-scale of the figure
- $extbf{c:} str$  Color
- lims: tuple or None Limits
- name: strName of the figure
- ext: strFormat of the image
- dpi: *int*Resolution of the image in dots per inches

# 3.3.14 figures.redraw\_contours(ax, ppm\_f2, ppm\_f1, S, lvl, cnt, Neg=False, Ncnt=None, lw=0.5, cmap=[None, None])

Redraws the contours in interactive 2D visualizations.

### Parameters:

- ax: matplotlib.Subplot Object
  Panel of the figure where to draw the contours
- ppm\_f2: 1darray ppm scale of the direct dimension
- ppm\_f1: 1darray ppm scale of the indirect dimension
- S: 2darray Spectrum
- lvl: *float*Level at which to draw the contours
- ullet cnt:  $matplotlib.contour.QuadContourSet\ object$  Pre-existing contours
- Neg: bool
  Choose if to draw the negative contours (True) or not (False)
- Ncnt: matplotlib.contour.QuadContourSet object Pre-existing negative contours
- lw: float Linewidth
- cmap: *list*Colour of the contours. Format: [cmap +, cmap -]

### Returns:

- ullet cnt: matplotlib.contour.QuadContourSet object Updated contours
- Ncnt: matplotlib.contour.QuadContourSet object or None Updated negative contours if Neg is True, None otherwise

## 3.3.15 figures.sns heatmap(data, name=None, ext='png', dpi=600)

Computes a heatmap of data, which is a matrix. This function employs the *seaborn* package. Specify 'name' if you want to save the figure.

### Parameters:

• data: 2darray

Data of which to compute the heatmap. Make sure the entries are real numbers.

• name:  $str\ or\ None$ Filename of the figure to be saved. If None, the figure is shown instead.

• ext: strFormat of the image

• dpi: *int*Resolution of the image in dots per inches

 $3.3.16 \quad figures.stacked\_plot(ppmscale, S, xlims=None, lw=0.5, X\_label='\$ \\ \quad delta$ 

\$ F1 /ppm', Y\_label='Normalized intensity /a.u.', n\_xticks=10, labels=None, name=None, ext='png', dpi=600)

Creates a stacked plot of all the spectra contained in the list 'S'. Note that 'S' MUST BE a list. All the spectra must share the same scale.

- ppmscale: *1darray* ppm scale of the spectrum
- S: *list* spectra to be plotted
- xlims: *list or tuple*Limits for the x-axis. If None, the whole scale is used.
- lw: float linewidth
- name: *str* filename of the figure, if it has to be saved;
- X\_label: str text of the x-axis label;
- Y\_label: str text of the y-axis label;
- n\_xticks: *int* Number of numbered ticks on the x-axis of the figure
- labels: *list* labels to be put in the legend.

# 3.4 SIM package

This package contains function for the simulation of various features of NMR spectra, being them monodimensional or bidimensional. Functions for the simulation of whole spectra are also provided.

# 3.4.1 sim.calc splitting(u0, I0, m=1, J=0)

Calculate the frequency and the intensities of a NMR signal splitted by scalar coupling.

# Parameters:

- u0: float
  Frequency of the non-splitted signal (Hz)
- I0: float
  Total intensity of the non-splitted signal.
- $\bullet$  m: int Multiplicity, i.e. number of expected signals after the splitting
- J: float Scalar coupling constant (Hz)

### Returns:

- u\_s: 1darray
  Frequencies of the splitted signal (Hz)
- I\_s: 1darray
  Intensities of the splitted signal

# $3.4.2 \quad sim.f_gaussian(x, u, s, A=1)$

Gaussian function in the frequency domain:

$$G(x) = \frac{A}{\sqrt{2\pi s}} \exp\left[-\frac{1}{2} \left(\frac{x-u}{s}\right)^2\right]$$

# **Parameters**

- x: 1darray
  Independent variable
- u: float Peak position
- s: *float* Standard deviation
- A: float Intensity

# Returns

• f: *1darray*Gaussian function.

# $3.4.3 \quad sim.f\_lorentzian(x, u, fwhm, A=1)$

Lorentzian function in the frequency domain:

$$\mathcal{L}(x) = \frac{A}{\pi} \frac{\gamma}{(x-u)^2 + \gamma^2}$$

# **Parameters**

- x: 1darray Independent variable
- u: float Peak position
- fwhm: float Full-width at half-maximum,  $\Gamma = 2\gamma$
- A: float Intensity

# Returns

• f: *1darray* Lorentzian function.

# $3.4.4 \quad sim.f_pvoigt(x, u, fwhm, A=1, x_g=0)$

Pseudo-Voigt function in the frequency domain:

$$S(x) = x_g \mathcal{G}(x) + (1 - x_g) \mathcal{L}(x)$$

This is practically done by:

```
s = fwhm / 2.355

S = A* (sim.f_gaussian(x, u, s, A=x_g) + sim.f_lorentzian(x, u, fwhm, A=1-x_g))
```

### **Parameters**

- x: 1darray Independent variable
- u: float
  Peak position
- $\bullet$  fwhm: float Full-width at half-maximum
- A: float Intensity
- x\_g: *float* Fraction of gaussianity

# Returns

• S: 1darray Pseudo-Voigt function.

# 3.4.5 sim.gaussian\_filter(ppm, u, s)

Compute a gaussian filter to be used in order to suppress signals in the spectrum.

# Parameters:

- ppm: *1darray* Scale on which to build the filter
- u: *float*Position of the filter
- s: *float*Width of the filter (standard deviation)

# Returns:

• G: *1darray* Computed gaussian filter

# 3.4.6 sim.load sim 1D(File)

Creates a dictionary from the spectral parameters listed in the input file. Template of the input file:

```
BO 16.4
                      # Magnetic field strength /Tesla
                      # Observed nucleus
nuc 1H
o1p 4.7
                      # Pulse carrier frequency /ppm
                      # Spectral width /ppm
SWp 30
TD 4096
                      # Number of sampled complex points
shifts 0, 4.5, 3
                      # Peak shifts /ppm, separated by commas
       100, 100, 100 # Full-width at half maximum of the peaks /Hz, separated by commas
amplitudes 1, 4, 3
                      # Amplitudes of the peaks, separated by commas
                      # Fraction of qaussianity of the peaks (1 = 100% qaussian, 0 = 100%
x_g 0.0, 0.5, 1.0
    lorentzian)
```

Use the tab character to separate the variable from its value. Comments are placed here to explain what the variables stand for, they are not needed in a real input file. However, if you want to put them, use the # character to denote them. Note that the variable names CANNOT be changed for any reason: the penalty is a massive sequence of errors. If only one value is supplied for the fields shifts, fwhm, amplitudes and x\_g, that value must be followed by a comma in order to allow the program to recognize the list as a tuple.

#### **Parameters**

• File: strPath to the input file location

#### Returns

• dic: dict
Dictionary of the parameters, ready to be read from the simulation functions.

# 3.4.7 sim.load sim 2D(File, states=True)

Creates a dictionary from the spectral parameters listed in the input file. Template of the input file:

```
B0 28.2
                  # Magnetic field strength /Tesla
                  # Observed nucleus in indirect dimension (F1)
nuc1
       15N
nuc2
                  # Observed nucleus in direct dimension (F2)
o1p 115
                  # Pulse carrier frequency /ppm in F1
o2p 5
                  # Pulse carrier frequency /ppm in F2
                  # F1 Spectral width /ppm
SW1p
       40
                  # F2 Spectral width /ppm
SW2p
       20
TD1 64
                  # Number of t1 increment in indirect dimension
                  # Number of sampled complex points for each transient
TD2 256
shifts_f1 130.0, 105.0, 120.0
                                     # Peak F1 shifts /ppm, separated by commas
shifts_f2 0.0, 0.0, 7.0
                                     # Peak F2 shifts /ppm, separated by commas
fwhm_f1 100, 100, 100
                                     # Full-width at half maximum of the peaks in F1
    (/Hz), separated by commas
                                     # Full-width at half maximum of the peaks in F2
fwhm_f2 500, 500, 500
    (/Hz), separated by commas
amplitudes 100.0, 200.0, 100.0
                                     # Intensity of the peaks, separated by commas
                                     # # Fraction of gaussianity of the peaks (1 = 100%
x_g 0.0, 0.5, 1.0
   gaussian, 0 = 100% lorentzian)
```

Use the tab character to separate the variable from its value. Comments are placed here to explain what the variables stand for, they are not needed in a real input file. However, if you want to put them, use the # character to denote them. Note that the variable names **CANNOT** be changed for any reason: the penalty is a massive sequence of errors. If only one value is supplied for the fields shifts, fwhm, amplitudes and x\_g, that value must be followed by a comma in order to allow the program to recognize the list as a tuple.

### **Parameters**

- File: strPath to the input file location
- states: bool

  If FnMODE is States or States-TPPI, set it to True to get the correct timescale.

# Returns

• dic: dict

Dictionary of the parameters, ready to be read from the simulation functions.

3.4.8 sim.mult\_noise(data\_size, mean, s\_n)

 $\underline{\text{Multiplicative noise model.}}$ 

# 3.4.9 sim.multiplet(u, I, m='s', J=[])

Split a given signal according to a scalar coupling pattern.

# Parameters:

- u: float
  Frequency of the non-splitted signal (Hz)
- I: *float*Intensity of the non-splitted signal
- m: str Organic chemistry-like multiplet, i.e. s, d, dqt, etc.
- J: float or list
  Scalar coupling constants. The number of constants should match the number of coupling branches

# Returns:

- u\_in: *list*List of the splitted frequencies (Hz)
- I\_in: *list*Intensities of the splitted signal

# $3.4.10 \quad \text{sim.noisegen(size, o2, t2, s n=1)}$

Simulates additive noise in the time domain, in the form of a matrix of dimensions size.

This model for the noise depicts it as a white noise vector (i.e. random number that fit a gaussian distribution with 0 mean and standard deviation equal to s\_n), modulated for the carrier frequency o2.

We consider the noise as the sum of a correlated contribution, due to the fact that the signal in a real spectrometer travels through the same cables until the ADC, and of a non-correlated contribution, which arises from the separation into real channel and imaginary channel.

This translates in the following code:

```
# correlated part of noise until ADC
white_corr = np.random.normal(0, s_n, size)
# white noise in FID has to be centered on the offset frequency
noise_corr = white_corr * np.exp(1j* 2 * np.pi * o2 * t2)

# uncorrelated part of noise: quadrature detection
white_re = np.random.normal(0, s_n, size)
white_im = np.random.normal(0, s_n, size)
# cosine-modulated in the real channel and sine-modulated in the imaginary channel
noise_re = white_re * np.cos( 2* np.pi * o2 * t2)
noise_im = white_im * np.sin( 2* np.pi * o2 * t2)

# final noise is sum of the two parts
noise = noise_corr + (noise_re + 1j*noise_im)
```

### **Parameters**

- size: int or tuple

  Dimension of the noise matrix
- o2: float Carrier frequency, in Hz.
- t2: *1darray*Time scale of the last temporal dimension.
- s\_n: *float*Standard deviation of the noise.

#### Returns

• noise: 2darray
Noise matrix, of dimensions size.

# 3.4.11 sim.sim 1D(File, pv=False)

Simulates a 1D NMR spectrum from the instructions written in File.

The instructions on how to write the input file are reported in the caption of sim.load\_sim\_1D.

# **Parameters**

• File: strPath to the input file location

• pv: bool

True for pseudo-Voigt model, False for Voigt model.

# Returns

• fid: *1darray* FID of the simulated spectrum.

# 3.4.12 sim.sim 2D(File, states=True, alt=True, pv=False)

Simulates a 2D NMR spectrum from the instructions written in File. The indirect dimension is sampled with FnMODE=States-TPPI as default.

The instructions on how to write the input file are reported in the caption of sim.load\_sim\_2D.

# **Parameters**

• File: strPath to the input file location

• states: bool
Set it to True to allow for correct spectral arrangement in the indirect dimension.

• alt: bool
Set it to True to allow for correct spectral arrangement in the indirect dimension.

• pv: bool

True for pseudo-Voigt model, False for Voigt model.

### Returns

• fid: 2darray
FID of the simulated spectrum.

# 3.4.13 sim.t 2Dgaussian(t1, t2, v1, v2, s1, s2, A=1, states=True, alt=True)

Bidimensional gaussian peak in the time domain. The working code requires states=True and alt=True.

The signal is generated as follows:

```
# States acquires twice the same point of the indirect dimension time domain
t1[1::2] = t1[::2]
# TPPI cycles the receiver phase of 90 degrees at each transient acquisition
freq_1 = np.zeros(len(t1), dtype='complex64')
for k in range(4):
    t1t = t1[k::4]
    freq_1[k::4] = np.cos( (2 * np.pi * v1 * t1t) - (0.5 * np.pi * np.mod(k,4) ))

# NMR signal in the direct dimension
F2 = np.exp(1j*2*np.pi*v2*t2) * np.exp(-(s2**2 * t2**2)/2)
# NMR signal in the indirect dimension
F1 = freq_1 * np.exp(-(s1**2 * t1**2)/2)

# The full FID is reconstructed by doing the external product between the two vectors
S = A * F1.reshape(-1,1) @ F2.reshape(1,-1)
```

### **Parameters**

- t1: *1darray*Indirect evolution timescale
- t2: *1darray*Timescale of the direct dimension
- v1: float
  Peak position in the indirect dimension, in Hz
- v2: *float*Peak position in the direct dimension, in Hz
- s1: *float*Standard deviation in the indirect dimension, in rad/s
- s2: float
  Standard deviation in the direct dimension, in rad/s
- A: float Intensity
- states: bool
  Set to True for FnMODE = States-TPPI
- alt: bool
  Set to True for FnMODE = States-TPPI

#### Returns

• S: 2darray
2D Gaussian function.

# $3.4.14 \quad sim.t\_2Dlorentzian(t1, t2, v1, v2, fwhm1, fwhm2, A=1, states=True, alt=True)$

Bidimensional Lorentzian peak. The working code requires states=True and alt=True. The signal is generated as follows:

```
hwhm1 = fwhm1 / 2
hwhm2 = fwhm2 / 2

# States acquires twice the same point of the indirect dimension time domain
t1[1::2] = t1[::2]

# TPPI cycles the receiver phase of 90 degrees at each transient acquisition
freq_1 = np.zeros(len(t1), dtype='complex64')
for k in range(4):
    t1t = t1[k::4]
    freq_1[k::4] = np.cos( (2 * np.pi * v1 * t1t) - (0.5 * np.pi * np.mod(k,4) ))

# NMR signal in the direct dimension
F2 = np.exp(1j*2*np.pi*v2*t2) * np.exp(-(hwhm2 * t2))
# NMR signal in the indirect dimension
F1 = freq_1 * np.exp(-(hwhm1 * t1))

# The full FID is reconstructed by doing the external product between the two vectors
S = A * F1.reshape(-1,1) @ F2.reshape(1,-1)
```

# **Parameters**

- t1: *1darray*Indirect evolution timescale
- t2: *1darray*Timescale of the direct dimension
- v1: float
  Peak position in the indirect dimension, in Hz
- v2: *float*Peak position in the direct dimension, in Hz
- fwhm1: float
  Full-width at half maximum in the indirect dimension, in rad/s
- fwhm2: float
  Full-width at half maximum in the direct dimension, in rad/s
- A: float Intensity
- states: bool Set to True for 'FnMODE': 'States-TPPI
- alt: bool
  Set to True for 'FnMODE': 'States-TPPI

# Returns

• S: *2darray* Lorentzian function.

# 3.4.15 $sim.t_2Dpvoigt(t1, t2, v1, v2, fwhm1, fwhm2, A=1, x_g=0.5, states=True, alt=True)$

Generates a 2D pseudo-voigt signal in the time domain. x\_g states for the fraction of gaussianity, whereas A defines the overall amplitude of the total peak. Indexes '1' and '2' on the variables stand for 'F1' and 'F2', respectively.

```
# stdev computed for the gaussian part.
s1 = fwhm1 / 2.355
s2 = fwhm2 / 2.355
# Passing 's' to 'gaussian' and 'fwhm' to 'lorentzian' makes the two parts of the pseudo-voigt signal to have the same width and allow proper summation
G = sim.t_2Dgaussian(t1, t2, v1, v2, s1, s2, A=x_g, states=states, alt=alt)
L = sim.t_2Dlorentzian(t1, t2, v1, v2, fwhm1, fwhm2, A=(1-x_g), states=states, alt=alt)
fid = A * (G + L)
```

#### **Parameters**

- t1: *1darray*Indirect evolution timescale
- t2: *1darray*Timescale of the direct dimension
- v1: float
  Peak position in the indirect dimension, in Hz
- v2: float
  Peak position in the direct dimension, in Hz
- fwhm1: float
  Full-width at half maximum in the indirect dimension, in rad/s
- fwhm2: float
  Full-width at half maximum in the direct dimension, in rad/s
- A: float Intensity
- x\_g: *float* Fraction of gaussianity
- states: bool Set to True for FnMODE=States-TPPI
- alt: bool
  Set to True for FnMODE=States-TPPI

### Returns

• fid: 2darray
2D Pseudo-Voigt function.

# $3.4.16 \quad sim.t\_2Dvoigt(t1, t2, v1, v2, fwhm1, fwhm2, A=1, x\_g=0.5, states=True)$

Generates a 2D Voigt signal in the time domain. x\_g states for the fraction of gaussianity, whereas A defines the overall amplitude of the total peak. Indexes '1' and '2' on the variables stand for 'F1' and 'F2', respectively.

```
# stdev computed for the gaussian part.
s1 = fwhm1 / 2.355
s2 = fwhm2 / 2.355
# hwhm computed for the lorentzian part.
hwhm1 = fwhm1 / 2
hwhm2 = fwhm2 / 2
# States acquires twice the same point of the indirect dimension time domain
t1[1::2] = t1[::2]
# direct dimension
   frequency
freq_2 = np.exp(1j * 2 * np.pi * v2 * t2)
   Add line-broadening, fist lorentzian then gaussian, using:
   hwhm' = (1 - x_g) * hwhm
                                for L
   s' = x_q * s
                                for G
F2 = freq_2 * np.exp(-(1-x_g)*hwhm2 * t2) * np.exp(-((x_g*s2)**2 * t2**2)/2)
# indirect dimension
# Redfield cycles the receiver phase of 90 degrees at each transient acquisition
freq_1 = np.zeros(len(t1), dtype='complex64')
for k in range(4):
   t1t = t1[k::4]
   freq_1[k::4] = np.cos((2 * np.pi * v1 * t1t) - (0.5 * np.pi * np.mod(k,4)))
  Add line-broadening, fist lorentzian then gaussian, using:
   hwhm' = (1 - x_g) * hwhm
                                for L
   s' = x_g * s
                                for G
F1 = freq_1 * np.exp(-(1-x_g) * hwhm1 * t1) * np.exp(-((x_g*s1)**2 * t1**2)/2)
# The full FID is reconstructed by doing the external product between the two vectors
S = A * F1.reshape(-1,1) @ F2.reshape(1,-1)
return S
```

# **Parameters**

- t1: *1darray*Indirect evolution timescale
- t2: 1darray
  Timescale of the direct dimension
- v1: float
  Peak position in the indirect dimension, in Hz
- v2: float
  Peak position in the direct dimension, in Hz

- $\bullet$  fwhm 1: float Full-width at half maximum in the indirect dimension, in rad/s
- $\bullet$  fwhm2: float Full-width at half maximum in the direct dimension, in rad/s
- A: float Intensity
- x\_g: *float* Fraction of gaussianity
- states: bool Set to True for 'FnMODE': 'States-TPPI
- alt: bool Set to True for 'FnMODE': 'States-TPPI

# Returns

• S: 2darray Voigt function.

# 3.4.17 sim.t\_gaussian(t, u, s, A=1, phi=0)

Gaussian function in the time domain.

$$g(x) = Ae^{i\phi} e^{i2\pi ut} e^{-s^2t^2/2}$$

# **Parameters**

- t: *1darray* Independent variable
- u: float Peak position, in Hz
- A: float Intensity
- phi: *float* Phase, in radians

# Returns

• S: *1darray* Gaussian function.

# 3.4.18 sim.t\_lorentzian(t, u, fwhm, A=1, phi=0)

Lorentzian function in the time domain.

$$\ell(x) = Ae^{i\phi} e^{i2\pi ut} e^{-\gamma t}$$

# **Parameters**

- t: 1darray Independent variable
- u: float Peak position, in Hz
- fwhm: float Full-width at half-maximum,  $\Gamma=2\gamma$ , in rad/s
- A: float Intensity
- phi: *float* Phase, in radians

# Returns

• S: *1darray* Lorentzian function.

# $3.4.19 \quad sim.t_pvoigt(t, u, fwhm, A=1, x_g=0, phi=0)$

Pseudo-Voigt function in the time domain:

$$s(x) = x_g g(x) + (1 - x_g)\ell(x)$$

- s = fwhm / 2.355
- S = A \* (sim.t\_gaussian(t, u, s, A=x\_g, phi=phi) + sim.t\_lorentzian(t, u, fwhm, A=1-x\_g, phi=phi))

### **Parameters**

- t: *1darray*Independent variable
- u: float
  Peak position, in Hz
- ullet fwhm: float Full-width at half-maximum, in rad/s
- A: float Intensity
- x\_g: *float* Fraction of gaussianity
- phi: float
  Phase, in radians

# Returns

• S: *1darray* Pseudo-Voigt function.

# $3.4.20 \quad sim.t\_voigt(t, u, fwhm, A=1, x\_g=0, phi=0)$

Voigt function in the time domain. The parameter  $x_g$  affects the linewidth of the lorentzian and gaussian contributions.

```
s = fwhm / 2.355
S = A * np.exp(1j*phi) * sim.t_gaussian(t, u/2, s*x_g) * sim.t_lorentzian(t, u/2, fwhm*(1-x_g))
```

# **Parameters**

- t: 1darray
  Independent variable
- u: float Peak position, in Hz
- fwhm: float Full-width at half-maximum, in rad/s
- A: float Intensity
- x\_g: *float* Fraction of gaussianity
- phi: float
  Phase, in radians

### Returns

• S: *1darray* Voigt function.

# 3.4.21 sim.water7(N, t2, vW, fwhm=300, A=1, spread=701.125)

Simulates a feature like the water ridge in HSQC spectra, in the time domain.

This signal is modelled as a gaussian signal which does not encode for any frequency in the indirect dimension, and whose chemical shift moves due to field drifts through the various transients according to a gaussian distribution. This signal is on-phase in the even transients and 90°-dephased in the odd transients.

# **Parameters**

- N: *int*Number of transients
- t2: *1darray*Time scale of the last temporal dimension.
- vW: *float*Nominal peak position, in Hz.
- fwhm: float
  Nominal full-width at half maximum of the peak, in rad/s
- A: float
  Signal intensity.
- spread: *float*Standard deviation of the peak position distribution, in Hz.

### Returns

• ridge: *2darray*Matrix of the ridge.

# 3.5 FIT package

Functions for performing fits.

# 3.5.1 fit.CostFunc class

Class that groups several ways to compute the target of the minimization in a fitting procedure. It includes the classic squared sum of the residuals, as well as some other non-quadratic cost functions. Let x be the residuals and s the chosen threshold value. Then the objective value R is computed as:

$$R = \sum_{i} if(x_i)$$

where f(x) can be chosen between the following options:

• Quadratic:

$$f(x) = x^2$$

• Truncated Quadratic:

$$f(x) = \begin{cases} x^2 & \text{if } |x| < s \\ s^2 & \text{otherwise} \end{cases}$$

• Huber function:

$$f(x) = \begin{cases} x^2 & \text{if } |x| < s \\ 2s|x| - s^2 & \text{otherwise} \end{cases}$$

• Asymmetric Truncated Quadratic:

$$f(x) = \begin{cases} x^2 & \text{if } x < s \\ s^2 & \text{otherwise} \end{cases}$$

• Asymmetric Huber function:

$$f(x) = \begin{cases} x^2 & \text{if } x < s \\ 2sx - s^2 & \text{otherwise} \end{cases}$$

# Attributes:

- method: function

  Function to be used for the computation of the objective value. It must take as input the array of the residuals and the threshold, no matter if the latter is actually used or not.
- s: float
  Threshold value

# Methods:

Initialize the method according to your choice, then stores the threshold value in the attribute 's'. Allowed choices are:

- 'q': Quadratic
- 'tq': Truncated Quadratic
- 'huber': Huber function
- 'atq': Asymmetric Truncated Quadratic
- 'ahuber': Asymmetric Huber function

#### Parameters:

- $\bullet$  method: str Label for the method selection
- s: float
  Threshold value

# call (self, x)

Computes the objective value according to the chosen method and the residuals array x.

# Parameters:

• x: 1darray
Array of the residuals

# Returns:

• R: float
Computed objective value

# asymm huber(r, s)

Linear behaviour above s, penalizes negative entries

# asymm truncated quadratic(r, s)

Constant behaviour above s, penalizes negative entries

# huber(r, s)

Linear behaviour above s

# method selector(self, method)

Performs the selection of the method according to the identifier string.

Parameters:
-------------

• method: str Method label

# Returns:

• f: function Selected model

 $squared\_sum(r,\,s{=}0)$ 

Quadratic everywhere

 $truncated\_quadratic(r, s)$ 

Constant behaviour above s

# 3.5.2 fit.LR(y, x=None)

Performs a linear regression of y with a model  $y_c = mx + q$ .

# Parameters:

- y: 1darray
  Data to be fitted
- ullet x: 1darray Independent variable. If None, the point indexes are used.

# Returns:

- y\_c: 1darray Fitted trend
- values: tuple (m, q)

# 3.5.3 fit.LSP(y, x, n=5)

Make a polynomial fit on the experimental data y by solving the linear system

$$\vec{y} = \mathbb{T}\vec{c}$$

where  $\mathbb{T}$  is the Vandermonde matrix of the x-scale and c is the set of coefficients that minimize the problem in the least-squares sense.

### Parameters:

- y: 1darray Experimental data
- x: 1darray
  Independent variable (better if normalized)
- ullet n: int Order of the polynomion + 1, i.e. number of coefficients

# Returns:

• c: *1darray* Set of minimized coefficients

# 3.5.4 fit.SINC ObjFunc

class

Computes the objective function as explained in 'M. Sawall et al., *Journal of Magnetic Resonance* 289 (2018), 132-141'.

The cost function is computed as:

$$f(d) = \sum_{i=1}^{3} \gamma_i g_i(d|\varepsilon_i)$$

where d is the real part of the NMR spectrum.

# Attributes:

- gamma1: *float*Weighting factor for function g1
- gamma2: float
  Weighting factor for function g2
- gamma3: float
  Weighting factor for function g3
- e1: *float*Tolerance value for function g1
- e2: *float*Tolerance value for function g2

# Methods:

 $\_\_init\_\_(self, \, gamma1 = 10, \, gamma2 = 0.01, \, gamma3 = 0, \, e1 = 0, \, e2 = 0)$ 

Initialize the coefficients used to weigh the objective function.

### Parameters:

- gamma1: float
  Weighting factor for function g1
- gamma2: float
  Weighting factor for function g2
- gamma3: *float*Weighting factor for function g3
- e1: *float*Tolerance value for function g1
- e2: float
  Tolerance value for function g2

 $\_\_call\_\_(self,\,d)$ 

Computes the objective function f as explained in the paper

# g1(d, e1=0)

Penalty function for negative entries of the spectrum

# Parameters:

- d: *1darray* Spectrum
- e1: *float*Tolerance for negative entries

# g2(d, e2=0)

Regularization function that favours the smallest integral.

# Parameters:

- d: *1darray* Spectrum
- e2: *float*Tolerance for ideal baseline

# **g3(d)**

Regularization function for the smoothing.

# Parameters:

• d: *1darray* Spectrum

# 3.5.5 fit.SINC\_phase(data, gamma1=10, gamma2=0.01, gamma3=0, e1=0, e2=0, \*\*fit kws)

Perform automatic phase correction according to the SINC algorithm, as described in 'M. Sawall et al., *Journal of Magnetic Resonance* 289 (2018), 132-141'. The fitting method defaults to 'least\_squares'.

### Parameters:

- data: *1darray* Spectrum to phase-correct
- gamma1: floatWeighting factor for function  $g_1$ : non-negativity constraint
- gamma2: floatWeighting factor for function  $g_2$ : smallest-integral constraint
- gamma3: floatWeighting factor for function  $g_3$ : smoothing constraint
- e1: floatTolerance factor for function  $g_1$ : adjustment for noise
- e2: floatTolerance factor for function  $g_2$ : adjustment for non-ideal baseline
- fit\_kws: keyworded arguments
  Additional parameters for the fit function. See lmfit.Minimizer.minimize for details. Do not use 'leastsq' because the cost function returns a scalar value!

### Returns:

- p0: float
  Fitted zero-order phase correction angle, in degrees
- p1: float
  Fitted first-order phase correction angle, in degrees

# 3.5.6 fit.Voigt Fit

class

This class offers an 'interface' to fit a 1D NMR spectrum.

#### Attributes:

- ppm\_scale: 1darray Self-explanatory
- S: 1darray
  Spectrum to fit. Only real part
- t\_AQ: 1darray acquisition timescale of the spectrum
- SFO1: *float*Larmor frequency of the nucleus
- olp: float
  Pulse carrier frequency
- nuc : str or None Nucleus. Used to write the X\_scale of the plot.
- input\_file : str filename of the input file
- output\_file : str filename of the output file
- log\_file : str filename of the log file
- limits: tuple borders of the fitting window
- Vi : 2darray array with the values of the signals used as initial guess
- Ci : *1darray* coefficients of the baseline polynomion as initial guess
- Vf: 2darray array with the values of the signals after the fit
- Cf: 1darray coefficients of the baseline polynomion after the fit
- s\_labels : *list* legend entries for the single signals.

### Methods:

\_\_init\_\_(self, ppm\_scale, S, t\_AQ, SFO1, o1p, nuc=None)

Initialize the class with common values.

### Parameters:

- ppm\_scale: *1darray* ppm scale of the spectrum
- S: *1darray*Spectrum to be fitted
- t\_AQ: 1darray
  Acquisition timescale
- SFO1: float
  Larmor frequency of the observed nucleus, in MHz
- o1p: float
  Carrier position, in ppm
- $\bullet$  nuc: str Observed nucleus. Used to customize the x-scale of the figures.

dofit(self, input file=None, output file=None, log file='fit.log', on cplex=True, \*\*kwargs)

Performs the fit using fit.voigt fit with self.Vi and self.Ci as initial parameters.

#### Parameters:

- input\_file: strPath to the input file. If None, self.input\_file is used.
- output\_file: strPath to the output file. If None, self.output\_file is used.
- log\_file: strPath to the log file.
- on\_cplex: bool Choose if to fit on the complex data (True) or on the real data (False).
- kwargs: keyworded parameters See fit.voigt fit.

get\_fit\_lines(self, what='fit')

Calculates the components, total fit curve and baseline used for iguess or fit. Calls fit.calc\_fit\_lines to do it.

### Parameters:

• what: str 'iguess' or 'fit'. self.Vi and self.Ci or self.Vf and self.Cf are read, respectively.

#### **Returns:**

• signals: *list of 1darray*Components used for the fit

• Total: *1darray*Sum of all the signals

 $\bullet$  baseline: 1darray

Baseline.

# iguess(self, input file=None, limits=None)

Reads, or computes, the initial guess for the fit. The window limits can be set interactively or specified through the parameter 'limits'.

### Parameters:

- input\_file: str or None
  Path to the input file. If None, the default name 'ppm1:ppm2.inp' is set.
- limits: tuple or None (lim\_sx, lim\_dx) in ppm. If None and input\_file does not exist yet, they are set interactively.

# load fit(self, output file=None)

Reads the output file of a fit and stores the values in self.Vf, self.Cf and self.limits. self.output\_file is also replaced with output\_file.

### **Parameters:**

• output\_file: strPath to the output file to be read

# plot(self, what='fit', s\_labels=None, \*\*kwargs)

Makes a figure of either the initial guess or the fit. Calls figures.fitfigure.

# Parameters:

- what: str 'iguess' or 'fit'. self.Vi and self.Ci or self.Vf and self.Cf are read, respectively.
- s\_labels: *list*Labels for the components that will appear in the legend of the figure.
- kwargs: keyworded parameters
  See figures.fitfigure for the additional parameters.

# 3.5.7 fit.ax\_histogram(ax, data0, nbins=100, density=True, f\_lims=None, xlabel=None, x symm=False, barcolor='tab:blue', fontsize=10)

Computes an histogram of 'data' and tries to fit it with a gaussian lineshape. The parameters of the gaussian function are calculated analytically directly from 'data' using 'scipy.stats.norm'

# Parameters:

- ax : matplotlib.subplot Object panel of the figure where to put the histogram
- data0 : ndarray the data to be binned
- nbins : *int* number of bins to be calculated
- density : *bool*True for normalize data
- f\_lims: tuple or None limits for the x axis of the figure
- xlabel : str or None
  Text to be displayed under the x axis
- x\_symm : bool set it to True to make symmetric x-axis with respect to 0
- barcolor: strColor of the bins
- fontsize: *float*Biggest fontsize in the figure

### Returns:

- m : float Mean of data
- s : *float* Standard deviation of data.

# 3.5.8 fit.bin data(data0, nbins=100, density=True, x symm=False)

Computes the histogram of data, sampling it into nbins bins. If data is multidimensional, it is flattened.

### Parameters:

• data: *ndarray* the data to be binned

• nbins: *int* number of bins to be calculated

• density : bool

True for normalize data

• x\_symm : bool set it to True to make symmetric x-axis with respect to 0

# Returns:

• hist: *1darray*The bin intensity

• bin\_scale: *1darray*Scale built with the mean value of the bin widths.

## $3.5.9 \quad \text{fit.build} \quad 2D\_\text{sgn}(\text{parameters, acqus, N=None, procs=None})$

Create a 2D signal according to the final parameters returned by fit.gen\_iguess\_2D. Process it according to procs.

## Parameters:

- parameters: list or 2darray sequence of the parameters: u1, u2, fwhm1, fwhm2, I, x\_g. Multiple components are allowed
- acqus: dict
  2D-like acqus dictionary containing the acquisition timescales (keys: t1 and t2)
- N: tuple of int Zero-filling values (F1, F2). Read only if procs is None
- procs: *dict*2D-like procs dictionary.

## Returns:

• peak: *2darray*rr part of the generated signal

## 3.5.10 fit.build baseline(ppm scale, C, L=None)

Builds the baseline calculating the polynomion with the given coefficients, and summing up to the right position.

## Parameters:

- ppm\_scale: *1darray* ppm scale of the spectrum
- C: *list*Baseline coefficients. No baseline corresponds to False.
- L: *list*List of window regions. If it is None, the baseline is built on the whole ppm\_scale

## Returns:

• baseline: *1darray* Self-explanatory.

## 3.5.11 fit.calc fit lines(ppm scale, limits, t AQ, SFO1, o1p, N, V, C=False)

Given the values extracted from a fit input/output file, calculates the signals, the total fit function, and the baseline.

## Parameters:

- ppm\_scale: 1darray PPM scale of the spectrum
- limits: *tuple* (left, right) in ppm
- t\_AQ: 1darray
  Acquisition timescale
- SFO1: *float*Larmor frequency of the nucleus /ppm
- o1p: float
  Pulse carrier frequency /ppm
- N: *int* Size of the final spectrum.
- V: 2darray
  Matrix containing the values to build the signals.
- C: *1darray*Baseline polynomion coefficients. False to not use the baseline

- sgn: *list*Voigt signals built using V
- Total: *1darray* sum of all the arrays in sgn
- baseline: *1darray*Polynomion built using C as coefficients. False if C is False.

## 3.5.12 fit.dic2mat(dic, peak names, ns, A=None)

This is used to make the matrix of the parameters starting from a dictionary like the one produced by *lmfit*. The column of the total intensity is not added, unless the parameter A is passed. In this case, the third column (which is the one with the relative intesities) is corrected using the function misc.molfrac.

## Parameters:

• dic: dict input dictionary

• peak\_names : *list* list of the parameter entries to be looked for

• ns: int number of signals to unpack

• A: float or None Total intensity.

#### **Returns:**

• V: 2darray
Matrix containing the parameters.

## 3.5.13 fit.fit\_int(y, y\_c)

Computes the optimal intensity and intercept of a linear model in the least squares sense. Let y be the experimental data and  $y_c$  the model, and let  $\langle w \rangle$  the mean of variable w. Then:

$$A = \frac{\langle y y_c \rangle - \langle y \rangle \langle y_c \rangle}{\langle y_c^2 \rangle - \langle y_c \rangle^2}$$

$$\langle y \rangle \langle y^2 \rangle - \langle y y_c \rangle \langle y^2 \rangle$$

$$q = \frac{\langle y \rangle \langle y_c^2 \rangle - \langle y y_c \rangle \langle y_c \rangle}{\langle y_c^2 \rangle - \langle y_c \rangle^2}$$

## Parameters:

- y: 1darray Experimental data
- y\_c: 1darray Model data

- A: float Optimized intensity
- q: float Optimized intercept

## 3.5.14 fit.gaussian\_fit(x, y)

Fit y with a gaussian function, built using x as independent variable. The gaussian function is built with  $sim.f_gaussian$ .

## Parameters:

- x : 1darray x-scale
- y: 1darray data to be fitted

- ullet u: float mean
- s: *float* standard deviation
- ullet A: float Integral

## 3.5.15 fit.gen iguess(x, experimental, param, model, model args=[])

GUI for the interactive setup of a Parameters object to be used in a fitting procedure. Once you initialized the Parameters object with the name of the parameters and a dummy value, you are allowed to set the value, minimum, maximum and vary status through the textboxes given in the right column, and see their effects in real time. Upon closure of the figure, the Parameters object with the updated entries is returned. A maximum of 18 parameters will fit the figure.

## Parameters:

- x: 1darray Independent variable
- experimental: *1darray*The objective values you are trying to fit
- param: *lmfit.Parameters Object* Initialized parameters object
- model: function

  Function to be used for the generation of the fit model. param must be the first argument.
- model\_args: list
   List of args to be passed to model, after param

#### **Returns:**

• param: *lmfit.Parameters Object* Updated Parameters Object

## $3.5.16 \quad \text{fit.gen\_iguess\_2D(ppm\_f1, ppm\_f2, tr1, tr2, u1, u2, acqus, fwhm0=100 procs=None)}$

Generate the initial guess for the fit of a 2D signal. The employes model is the one of a 2D Voigt signal, acquired with the States-TPPI scheme in the indirect dimension (i.e. sim.t\_2DVoigt). The program allows for the inclusion of up to 10 components for the signal, in order to improve the fit. The acqus dictionary must contain the following keys:

- t1: acquisition timescale in the indirect dimension (States)
- t2: acquisition timescale in the direct dimension
- SF01: Larmor frequency of the nucleus in the indirect dimension
- SF02: Larmor frequency of the nucleus in the direct dimension
- olp: carrier position in the indirect dimension /ppm
- o2p: carrier position in the direct dimension /ppm

The signals will be processed according to the values in the procs dictionary, if given; otherwise, they will be just zero-filled up to the data size (i.e. (len(ppm\_f1), len(ppm\_f2))).

#### Parameters:

- ppm\_f1: *1darray* ppm scale for the indirect dimension
- ppm\_f2: *1darray* ppm scale for the direct dimension
- tr1: 1darray
  Trace of the original 2D peak in the indirect dimension
- tr2: *1darray*Trace of the original 2D peak in the direct dimension
- u1: float Chemical shift of the original 2D peak in the indirect dimension /ppm
- u2: float Chemical shift of the original 2D peak in the direct dimension /ppm
- acqus: *dict*Dictionary of acquisition parameters
- fwhm0: float
  Initial value for FWHM in both dimensions
- procs: *dict*Dictionary of processing parameters

- final\_parameters: 2darray Matrix of dimension (# signals, 6) that contains, for each row:  $\nu_1({\rm Hz}),~\nu_2({\rm Hz}),~\Gamma_1({\rm Hz}),~\Gamma_2({\rm Hz}),~A,~x_g$
- fit\_interval: tuple of tuple Fitting window. ( (left\_f1, right\_f1), (left\_f2, right\_f2) )

## 3.5.17 fit.get region(ppmscale, S, rev=True)

Interactively select the spectral region to be fitted. Returns the border ppm values.

## Parameters:

- ppmscale: *1darray*The ppm scale of the spectrum
- S: *1darray*The spectrum to be trimmed
- rev: bool Choose if to reverse ppm scale and data (True) or not (False).

- left: *float*Left border of the selected spectral window
- right: *float*Right border of the selected spectral window

3.5.18 fit.histogram(data, nbins=100, density=True, f\_lims=None, xla-bel=None, x\_symm=False, barcolor='tab:blue', fontsize=10, name=Nonext='png', dpi=600)

Computes an histogram of 'data' and tries to fit it with a gaussian lineshape. The parameters of the gaussian function are calculated analytically directly from 'data' using 'scipy.stats.norm'

#### **Parameters:**

• data : ndarray the data to be binned

• nbins : *int* number of bins to be calculated

• density : *bool*True for normalize data

• f\_lims: tuple or None limits for the x axis of the figure

• xlabel : str or None

Text to be displayed under the x axis

• x\_symm : bool set it to True to make symmetric x-axis with respect to 0

• barcolor: strColor of the bins

• fontsize: *float*Biggest fontsize in the figure

• name: str name for the figure to be saved

• ext: strFormat of the image

• dpi: *int*Resolution of the image in dots per inches

#### **Returns:**

• m : float Mean of data

• s : float Standard deviation of data.

## 3.5.19 fit.integrate(ppm0, data0, X label='\$\delta\,\$F1 /ppm')

Allows interactive integration of a NMR spectrum through a dedicated GUI. Returns the values as a dictionary, where the keys are the selected regions truncated to the 2nd decimal figure. The returned dictionary contains some predefined keys, as follows:

- total: total integrated area
- ref\_pos: location of the reference peak (ppm1 : ppm2)
- ref\_int: absolute integral of the reference peak
- ref\_val: for how many nuclei the reference peak integrates

The absolute integral of the x-th peak,  $I_x$ , must be calculated according to the formula:

$$I_x = I_x^{\text{relative}} \cdot \frac{\texttt{ref\_int}}{\texttt{ref\_val}}$$

## **Parameters:**

- ppm: 1darray
  PPM scale of the spectrum
- data: *1darray* Spectrum to be integrated.
- X\_label: str Label of the x-axis

## Returns:

• f\_vals: dict
Dictionary containing the values of the integrated peaks.

## 3.5.20 fit.integrate\_2D(ppm\_f1, ppm\_f2, data, SFO1, SFO2, fwhm\_1=200, fwhm\_2=200, utol\_1=0.5, utol\_2=0.5, plot\_result=False)

Function to select and integrate 2D peaks of a spectrum, using dedicated GUIs. Calls integral\_2D to do the dirty job.

## Parameters:

- ppm\_f1: 1darray
  PPM scale of the indirect dimension
- ppm\_f2: *1darray* PPM scale of the direct dimension
- data: *2darray* real part of the spectrum
- SFO1: float
  Larmor frequency of the nucleus in the indirect dimension
- SFO2: float
  Larmor frequency of the nucleus in the direct dimension
- fwhm\_1: float Starting FWHM /Hz in the indirect dimension
- fwhm\_2: float Starting FWHM /Hz in the direct dimension
- utol\_1: float
  Allowed tolerance for u\_1 during the fit. (u\_1-utol\_1, u\_1+utol\_1)
- utol\_2: float
  Allowed tolerance for u\_2 during the fit. (u\_2-utol\_2, u\_2+utol\_2)
- plot\_result: bool

  True to show how the program fitted the traces.

## Returns:

• I: dict
Computed integrals. The keys are '<ppm f1>:<ppm f2>' with 2 decimal figures.

## 3.5.21 fit.interactive smoothing(x, y, cmap='RdBu')

Interpolate the given data with a 3rd-degree spline. Type the desired smoothing factor in the box and see the outcome directly on the figure. When the panel is closed, the smoothed function is returned.

## Parameters:

- x: 1darray
  Scale of the data
- y: *1darray*Data to be smoothed
- $\bullet$  cmap: strName of the colormap to be used to represent the weights

- sx: *1darray*Location of the spline points
- sy: *1darray* Smoothed y
- s\_f: float
  Employed smoothing factor for the spline
- weights: *1darray* Weights vector

## 3.5.22 fit.join par(filenames, ppm scale, joined name=None)

Load a series of parameters fit files. Join them together, returning a unique array of signal parameters, a list of coefficients for the baseline, and a list of tuples for the regions. Also, uses the coefficients and the regions to directly build the baseline according to the ppm windows.

## **Parameters:**

• filenames: *list*List of directories of the input files.

• ppm\_scale: 1darray ppm scale of the spectrum. Used to build the baseline

• joined\_name: str or None

If it is not None, concatenates the files in the list filenames and saves them in a single file named joined\_name.

## Returns:

• V: 2darray
Array of joined signal parameters

• C: *list*Baseline polynomion coefficients. No baseline corresponds to False.

• L: *list*List of window regions.

• baseline: *1darray*Baseline built from C and L.

## 3.5.23 fit.make\_iguess(S, ppm\_scale, t\_AQ, limits=None, SFO1=701.125, o1p=0, rev=True, name='i guess.inp')

Compute the initial guess for the quantitative fit of 1D NMR spectrum in an interactive manner. When the panel is closed, the values are saved in a file.

## Parameters:

- S: 1darray
  Spectrum to be fitted
- ppm\_scale : 1darray Self-explanatory
- t\_AQ : 1darray
  Acquisition timescale
- limits: tuple or None
  Trim limits for the spectrum (left, right). If None, the whole spectrum is used.
- SFO1 : float Larmor frequency /MHz
- o1p: float pulse carrier frequency /ppm
- rev: bool choose if you want to reverse the x-axis scale (True) or not (False).
- $\bullet$  name: str name of the file where to save the parameters

- V\_f: 2darray matrix (# signals, parameters)
- C\_f: 1darray or False
  Coefficients of the polynomion to be used as baseline correction. If the 'baseline' checkbox in the interactive figure panel is not checked, C\_f is False.

## 3.5.24 fit.make\_signal(t, u, s, k, x\_g, phi, A, SFO1=701.125, o1p=0, N=None)

Generates a Voigt signal using sim.t\_voigt on the basis of the specified parameters. Then, makes the Fourier transform and returns it.

## Parameters:

- t : ndarray acquisition timescale
- u: float chemical shift, in ppm
- s: *float* full-width at half-maximum, in hertz
- k: float relative intensity
- x\_g: *float* fraction of gaussianity
- phi: float phase of the signal, in degrees
- A: float total intensity
- SFO1: float Larmor frequency, in MHz
- o1p: float pulse carrier frequency, in ppm
- N: int or None length of the final signal array. If None, the signal is not zero-filled before to be transformed.

## Returns:

• sgn: *1darray* generated signal in the frequency domain (just the real part).

## 3.5.25 fit.peak pick(ppm f1, ppm f2, data, coord filename='coord.tmp')

Make interactive peak-picking on 2D spectra. The position of the selected signals are saved in coord\_filename. If coord\_filename already exists, the new signals are appended at its bottom: nothing is overwritten. Calls misc.select\_traces for the selection.

## Parameters:

- ppm\_f1: *1darray* ppm scale for the indirect dimension
- ppm\_f2: *1darray* ppm scale for the direct dimension
- data: 2darray
  Spectrum to peak-pick. The dimension should match the scale sizes.
- coord\_filename: str
  Path to the file where to save the peak coordinates

## Returns:

• coord: *list*List of (u2, u1) for each peak

## 3.5.26 fit.polyn basl(y, n=5, method='tq', s=0.2, itermax=10000)

Fit the baseline of a spectrum with a low-order polynomion using a non-quadratic objective function. Let y be an array of N points. The polynomion is generated on a normalized scale that goes from -1 to 1 in N steps, and the coefficients are initialized with the ordinary least squares fit. Then, the guess is refined using the objective function of choice employing the trust-region reflective least-squares algorithm.

## Parameters:

- y: 1darray
  Experimental data
- n: intOrder of the polynomion + 1, i.e. number of coefficients
- method: str Objective function of choice. 'q': quadratic, 'tq': truncated quadratic, 'huber': Huber, 'atq': asymmetric truncated quadratic, 'ahuber': asymmetric huber
- s: float
  Relative threshold value for the non-quadratic behaviour of the objective function
- itermax: *int*Number of maximum iterations

- px: *1darray*Fitted polynomion
- c: *list*Set of coefficients of the polynomion

## 3.5.27 fit.print\_par(V, C, limits=[None,None])

Prints on screen the same thing that fit.write\_par writes in a file.

## Parameters:

- V : 2darray matrix (# signals, parameters)
- C: 1darray or False
  Coefficients of the polynomion to be used as baseline correction. If it is False, they are not printed.
- limits: tuple or None
  Trim limits for the spectrum (left, right).

## 3.5.28 fit.read par(filename)

Reads the input file of the fit and returns the values.

## Parameters:

• filename: str directory and name of the input file to be read

- V: 2darray matrix (# signals, parameters)
- C: 1darray or False

  Coefficients of the polynomion to be used as baseline correction. If the corresponding flag is not found in the file, C\_f is set to False.
- limits: tuple
  Trim limits for the spectrum (left, right).

## 3.5.29 fit.smooth\_spl(x, y, s\_f=1, size=0, weights=None)

Fit the input data with a 3rd-order spline, given the smoothing factor to be applied.

## Parameters:

- x: 1darray
  Location of the experimental points
- y: 1darray Input data to be fitted
- s\_f: float Smoothing factor of the spline. 0=best straight line, 1=native spline.
- size: *int*Size of the spline. If size is 0, the same dimension as y is chosen.
- weights: *1darray*Array of weights of the spline points. **None** assign the same weight to all points in x.

- x\_s: 1darray Location of the spline data points.
- y\_s: 1darray Spline that fits the data.

## 3.5.30 fit.test residuals(R, nbins=100, density=False)

Test the residuals of a fit to see if it was reliable. Returns two values: SYSDEV and Q\_G.

SYSDEV is inspired by Svergun's Gnom, and it tells if there are systematic deviations basing on the number of sign changes in the residual. Optimal value must be 1. Values greater than 1 show positive bias, values lesser than 1 show negative bias.

Let R be an array of N points, and let  $N_s$  be the number of sign changes in R. Then:

$$\mathtt{SYSDEV} = \frac{N_s}{N/2}$$

Q\_G is to see the discrepancy between a gaussian function built with the mean and standard deviation of the residuals and the gaussian built fitting the histogram of the residuals. Values go from 0 (worst case) to 1 (best case).

Let x be the scale of the bin edges, and let  $G_T(x)$  and  $G_F(x)$  be the theoretical and the fitted gaussians respectively.

$$G_T(x) = \frac{A_T}{\sqrt{2\pi}\sigma_T} \exp\left[-\frac{1}{2} \left(\frac{x - \mu_T}{\sigma_T}\right)^2\right] \qquad G_F(x) = \frac{A_F}{\sqrt{2\pi}\sigma_F} \exp\left[-\frac{1}{2} \left(\frac{x - \mu_F}{\sigma_F}\right)^2\right]$$

Then:

$$Q_{G} = \frac{\int |G_{T}(x) - G_{F}(x)| dx}{\int G_{T}(x) dx + \int G_{F}(x) dx}$$

The numerator of this fraction expresses the area of the superposition region of the two gaussians. This value is then related to the worst possible case, i.e. when the two gaussians are totally disjointed.

#### Parameters:

• R: *1darray*Array of the residuals

• nbins: int number of bins of the histogram, which is also the length of x.

• density: *bool*True to normalize the histogram, False otherwise.

#### Returns:

• SYSDEV : float Read full caption

• Q\_G: float Read full caption 3.5.31 fit.voigt\_fit(S, ppm\_scale, V, C, t\_AQ, limits=None, SFO1=701.125, o1p=0, utol=0.5, vary\_phi=False, vary\_xg=True, hist\_name=None, write out='fit.out', test\_res=True)

Fits an NMR spectrum with a set of signals, whose parameters are specifed in the V matrix. There is the possibility to use a baseline through the parameter C. The signals are computed in the time domain and then Fourier transformed.

## Parameters:

- S: 1darray
  Spectrum to be fitted
- ppm\_scale : 1darray Self-explanatory
- V : 2darray matrix (# signals, parameters)
- C: 1darray or False
  Coefficients of the polynomion to be used as baseline correction. If it is False, the baseline correction is not used.
- t\_AQ : 1darray
  Acquisition timescale
- limits: tuple or None
  Trim limits for the spectrum (left, right). If None, the whole spectrum is used.
- SFO1: float
  Larmor frequency /MHz
- o1p: float pulse carrier frequency /ppm
- utol: float tolerance for the chemical shift. The peak center can move in the range  $[\mu utol, \mu + utol]$ .
- vary\_xg: bool

  If it is False, the parameter x\_g cannot be varied during the fitting procedure. Useful when fitting with pure Gaussians or pure Lorentzians.
- vary\_basl: bool

  If it is False, the baseline is kept fixed at the initial parameters.

- C\_f: 1darray or False
  Coefficients of the polynomion to be used as baseline correction, or just False if not used.
- V\_f: 2darray matrix (# signals, parameters) after the fit
- result : *lmfit.fit\_result Object* container of all information on the fit

 $3.5.32 \quad fit.voigt\_fit\_2D(x\_scale, y\_scale, data, parameters, lim\_f1, lim\_f2, \\ acqus, N=None, procs=None, utol=(1,1), s1tol=(0,500), s2tol=(0,500), \\ vary\_xg=False, logfile=None)$ 

Function that performs the fit of a 2D peak using multiple components. The program reads a parameter matrix, that contains:

u1 /ppm, u2 /ppm, fwhm1 /Hz, fwhm2 /Hz, I /a.u., x\_g

in each row. The number of rows corresponds to the number of components used for the computation of the final signal. The function returns the analogue version of the parameters matrix, but with the optimized values.

## **Parameters:**

- x\_scale: 1darray ppm f2 of the spectrum, full
- y\_scale: 1darray ppm\_f1 of the spectrum, full
- data: 2darray spectrum, full
- parameters: 1darray or 2darray
  Matrix (# signals, 6). Read main caption.
- lim\_f2: tuple
  Trimming limits for x scale
- lim\_f1: tuple
  Trimming limits for y scale
- acqus: *dict*Dictionary of acquisition parameters.
- N: tuple of ints len(y\_scale), len(x\_scale). Used only if procs is None
- procs: *dict*Dictionary of processing parameters.
- utol: tuple of floats

  Tolerance for the chemical shifts (utol\_f1, utol\_f2). Values will be set to  $u_1 \pm utol_f1$ ,  $u_2 \pm utol_f2$ .
- s1tol: tuple of floats
  Range of variations for the fwhm in f1, in Hz
- s2tol: tuple of floats
  Range of variations for the fwhm in f2, in Hz
- vary\_xg: boolChoose if to fix the  $x_q$  value or not
- logfile: str or None
  Path to a file where to write the fit information. If it is None, they will be printed into standard output.

## Returns:

• out\_parameters: 2darray parameters, but with the optimized values.

# 3.5.33 fit.write\_log(input\_file, output\_file, limits, V\_i, C\_i, V\_f, C\_f, result, runtime, test\_res=True, log\_file='fit.log')

Write a log file with all the information of the fit.

## Parameters:

- input\_file: str Location and filename of the input file
- output\_file: strLocation and filename of the output file
- limits: tuple
  Delimiters of the spectral region that was fitted. (left, right)
- V\_i: 2darray
  Initial parameters of the fit
- C\_i: 1darray or False
  Coefficients of the starting polynomion used for baseline correction. If False, they are not written.
- V\_f: 2darray Final parameters of the fit
- C\_f: 1darray or False
  Coefficients of the final polynomion used for baseline correction. If False, they are not written.
- result:  $lmfit.FitResult\ Object$ Object returned by lmfit after the fit.
- runtime: datetime.datetime Object
  Time taken for the fit
- test\_res: bool
  Choose if to test the residual with the fit.test\_residual function (True) or not (False)
- log\_file: str Filename of the log file to be saved.

## 3.5.34 fit.write par(V, C, limits, filename='i guess.inp')

Write the parameters of the fit, whether they are input or output.

## Parameters:

- V : 2darray matrix (# signals, parameters)
- C: 1darray or False
  Coefficients of the polynomion to be used as baseline correction. If it is False, they will not appear in the written file.
- limits: *tuple*Trim limits for the spectrum (left, right).
- filename: str or TextIO Object directory and name of the file to be written, or directly a file opened with open and writing rights.

## 3.6 SPECTRA package

All the classes in the Spectra module are automatically imported together with klassez itself. Refer to the examples reported in the *User guide* section to understand how to use them, or use the functions help(), vars(), dir() to get detailed info on how they exactly work.

## 3.6.1 Spectra.Pseudo 2D

class

Subclass of Spectrum\_2D to simulate and handle pseudo-2D experiments. Basically, they share more or less the same attributes, but some methods were adapted in order to suit well with a not-Fourier-transformed indirect dimension.

#### Attributes:

- datadir: *str*Path to the input file/dataset directory
- filename: strBase of the name of the file, without extensions
- fid: 2darray
  FID. For simulated data, this must be explicitly set!
- acqus: *dict*Dictionary of acqusition parameters
- ngdic: dict Created only if it is an experimental spectrum. Generated by nmrglue.bruker.read, contains all the information on the spectrometer and on the spectrum.
- procs: dict
  Dictionary of processing parameters
- S: 2darray
  Complex spectrum
- rr: 2darray Real part F2, real part F1
- ii: 2darray Imaginary part F2, imaginary part F1
- freq\_f1: 1darray
  Indeces of the experiments, works as placeholder
- freq\_f2: 1darray
  Frequency scale of the direct dimension, in Hz
- ppm\_f1: *1darray*Indeces of the experiments, works as placeholder
- ppm\_f2: *1darray* ppm scale of the direct dimension
- trf1: dict
  Projections of the indirect dimension, as 1darrays. Keys: 'ppm f2' where they were taken

• trf2: dict

Projections of the direct dimension, as 1darrays. Keys: 'ppm\_f1' where they were taken

• Trf1: dict

Projections of the indirect dimension, as pSpectrum\_1D objects. Keys: 'ppm\_f2' where they were taken

• Trf2: dict

Projections of the direct dimension, as pSpectrum\_1D objects. Keys: 'ppm\_f1' where they were taken

• integrals: dict

Dictionary where to save the regions and values of the integrals.

## Methods:

\_\_init\_\_(self, in\_file, fid=None, pv=False, isexp=True)

Initialize the class.

#### **Parameters:**

- in\_file: str
  path to file to read, or to the folder of the spectrum
- fid: 2darray or None Array that replaces self.fid.
- pv: bool

True if you want to use pseudo-voigt lineshapes for simulation, False for Voigt

• isexp: bool

True if this is an experimental dataset, False if it is simulated

## adjph(self, expno=0, p0=None, p1=None, pv=None, update=True)

Adjusts the phases of the spectrum according to the given parameters, or interactively if they are left as default.

#### Parameters:

- expno: *int*Index of the experiment (python numbering) to use in the interactive panel
- p0: float or None 0-th order phase correction /°
- p1: float or None 1-st order phase correction /°
- pv: float or None 1-st order pivot /ppm
- update: bool

Choose if to upload the procs dictionary or not

## align(self, lims=None, u off=0.5, ref idx=0)

Aligns the spectrum to a reference signal in the reference spectrum (default: first one).

#### **Parameters:**

- lims: tuple or None
  Reference signal region, in ppm. If None, you can select it interactively.
- u\_off: *float*Maximum displacement allowed, in ppm
- ref\_idx: int
  Index of the spectrum to be used as a reference (python numbering)

## basl(self, from procs=False, phase=True)

Apply baseline correction to the whole pseudo-2D by subtracting self.baseline from self.S. Then, self.S is unpacked in self.rr and self.ii.

## **Parameters:**

- from procs: bool

  If True, computes the baseline using the polynomion model reading self.procs['basl\_c'] as coefficients
- phase: bool
  Choose if to apply the same phase correction of the spectrum to the baseline. This should be done if the baseline was computed before the phase adjustment!

## cal(self, offset=[None, None], isHz=False, update=True)

Calibration of the ppm and frequency scales according to a given value, or interactively. In this latter case, a reference peak must be chosen. Calls processing.calibration

## Parameters:

- offset: tuple (scale shift F1, scale shift F2)
- isHz: tuple of bool

  True if offset is in frequency units, False if offset is in ppm
- update: *bool*Choose if to update the procs dictionary or not

## calf1(self, value=None, isHz=False)

Calibrates the ppm and frequency scale of the indirect dimension according to a given value, or interactively. Calls self.cal on F1 only.

#### Parameters:

• value: float or None scale shift value

• isHz: bool

True if offset is in frequency units, False if offset is in ppm

## calf2(self, value=None, isHz=False)

Calibrates the ppm and frequency scale of the direct dimension according to a given value, or interactively. Calls self.cal on F2 only

#### Parameters:

• value: float or None scale shift value

• isHz: bool

True if offset is in frequency units, False if offset is in ppm

## convdta(self, scaling=1)

Calls processing.convdta

## eae(self)

Calls processing.EAE to shuffle the data and make a States-like FID. Sets self.eaeflag to 0.

## integrate(self, which=0, lims=None)

Integrate the spectrum with a dedicated GUI. Calls processing.integral on each experiment, then saves the results in self.integrals. Therefore, the entries of self.integrals are sequences! If lims is not given, calls fit.integrate on the trace to select the regions to integrate.

## Parameters:

• which: int

Experiment index to show in interactive panel

 $\bullet$  lims: tuple

Region of the spectrum to integrate (ppm1, ppm2)

## inv process(self)

Performs the inverse processing of the spectrum according to the given parameters. Overwrites the S attribute!! Calls inv xfb.

## mc(self)

Computes the magnitude of the spectrum on self.S. Then, updates rr, ri, ir, ii.

## plot(self, Neg=True, lvl0=0.2, Y label=")

Plots the real part of the spectrum as a 2D contour plot.

#### Parameters:

- Neg: bool
  Plot (True) or not (False) the negative contours.
- lvl0: *float* Starting contour value.
- Y\_label: str Custom label for vertical axis.

## plot md(self, which=None, lims=None)

Plot a number of experiments, superimposed.

#### Parameters:

- which: str or None
  List of experiment indexes, so that eval(which) is meaningful. None plots all of them
- lims: tuple
  Region of the spectrum to show (ppm1, ppm2)

## plot stacked(self, which=None, lims=None)

Plot a number of experiments, stacked.

#### Parameters:

- which: str or None List of experiment indexes, so that eval(which) is meaningful. None plots all of them.
- lims: tuple
  Region of the spectrum to show (ppm1, ppm2)

## process(self)

Process only the direct dimension. Calls processing.fp on each transient. The parameters are read from the process dictionary

## projf1(self, a, b=None)

Calculates the sum trace of the indirect dimension, from a to b in F2. Store the trace in the dictionary trf1 and as 1D spectrum in Trf1. The key is 'a' or 'a:b' Updates the Trf1[label].freq and Trf1[label].ppm with self.freq f1 and self.ppm f1 respectively.

#### Parameters:

- a: *float* ppm F2 value where to extract the trace.
- b: float or None.

  If it is None, extract the trace in a. Else, sum from a to b in F2.

## projf2(self, a, b=None)

Calculates the sum trace of the direct dimension, from a to b in F1. Store the trace in the dictionary trf2 and as 1D spectrum in Trf2. The key is 'a' or 'a:b'

## Parameters:

- a: *float* ppm F1 value where to extract the trace.
- b: float or None.

  If it is None, extract the trace in a. Else, sum from a to b in F1.

## qfil(self, which=None, u=None, s=None)

Gaussian filter to suppress signals. Tries to read self.procs['qfil'], which is { 'u': u, 's': s } Otherwise, these are set interactively by processing.interactive\_qfil and then added to self.procs. Calls processing.qfil

## Parameters:

- which: int or None Index of the F2 trace to be used for interactive\_qfil. If None, a suitable trace can be selected using misc.select\_traces.
- u: float
  Position /ppm
- s: float
  Width (standard deviation) /ppm

## read\_procs(self, other\_dir=None)

Reads the procs dictionary from a file named 'filename.procs' in the same directory of the input file.

#### Parameters:

• other\_dir: str or None
Different location for the procs dictionary to look into. If None, self.datadir is used instead.
W! Do not put the trailing slash!

## **Returns:**

• procs: dict
Dictionary of processing parameters

## rpbc(self, ref exp=0, \*\*rpbc kws)

Computes the phase angles and the baseline using processing.rpbc on a reference spectrum taken from self.S. Then applies the phase correction and subtracts the baseline, automatically, to all experiments of the pseudo-2D. The procs dictionary is then updated and saved. The polynomial baseline is computed according to the given coefficients and stored in self.baseline

#### Parameters:

- ref\_exp: int
  Index of the reference experiment on which to apply the algorithm
- rpbc\_kws: keyworded arguments See processing.RPBC for details.

## write acqus(self, other dir=None)

Write the acqus dictionary in a file named 'filename.acqus'. Calls misc.write acqus 1D

## Parameters:

• other\_dir: str or None
Different location for the acqus dictionary to write into. If None, self.datadir is used instead.

## write integrals(self, filename='integrals.dat')

Write the integrals in a file named filename.

## Parameters:

• filename: str name of the file where to write the integrals.

## write procs(self, other dir=None)

Writes the actual procs dictionary in a file named 'filename.procs' in the same directory of the input file.

#### Parameters:

• other\_dir: str or None
Different location for the procs dictionary to write into. If None, self.datadir is used instead.
W! Do not put the trailing slash!

## write ser(ser, acqus, path=None)

Writes a real/complex array in binary format. Calls misc.write\_ser. Be sure that acqus contains the BYTORDA and DTYPA keys. See misc.write\_ser to understand the meaning of these values.

## Parameters:

- ser: *ndarray*Array that you want to convert in binary format.
- acqus: dict Dictionary of acquisition parameters. It must contain BYTORDA and DTYPA.
- path: strPath where to save the binary file.

## xf1(self)

Process only the indirect dimension. Transposes the spectrum in hypermode or normally if FnMODE!= QF, then calls for processing.fp using self.procs[keys][0], finally transposes it back. The result is stored in self.S, then self.rr and self.ii are written. freq\_fl and ppm\_fl are assigned with the indexes of the transients.

## xf2(self)

Process only the direct dimension. Calls processing.fp using procs[keys][1] The result is stored in self.S, then self.rr and self.ii are written. freq\_f1 and ppm\_f1 are assigned with the indexes of the transients.

class

# 3.6.2 Spectra.Spectrum 1D

# Attributes:

 $\bullet$  datadir: str

Class: 1D NMR spectrum

Path to the input file/dataset directory

 $\bullet$  filename: str

Base of the name of the file, without extensions

• fid: 1darray

FID

• acqus: dict

Dictionary of acquiition parameters

 $\bullet$  ngdic: dict

Created only if it is an experimental spectrum. Generated by nmrglue.bruker.read, contains all the information on the spectrometer and on the spectrum.

• procs: dict

Dictionary of processing parameters

• S: 1darray

Complex spectrum

• r: 1darray

Real part of the spectrum

• i: 1darray

Imaginary part of the spectrum

• freq: 1darray

Frequency scale of the spectrum, in Hz

• ppm: 1darray

ppm scale of the spectrum

• F: fit. Voigt Fit object

Used for deconvolution. See fit. Voigt fit.

• baseline: 1darray

Baseline of the spectrum.

• integrals: dict

Dictionary where to save the regions and values of the integrals.

#### Methods:

\_\_init\_\_(self, in\_file, pv=False, isexp=True, spect='bruker')

Initialize the class. Simulation of the dataset (i.e. isexp=False) employs sim.sim 1D.

- in\_file: str
  path to file to read, or to the folder of the spectrum
- pv: bool

  True if you want to use pseudo-voigt lineshapes for simulation, False for Voigt
- isexp: bool

  True if this is an experimental dataset, False if it is simulated
- spect: str Data file format. Allowed: 'bruker', 'varian', 'magritek', 'oxford'

# acme(self, \*\*method\_kws)

Automatic phase correction based on entropy minimization It calculates the phase angles using the algorithm specified in method, then calls self.adjph with those values.

#### Parameters:

• method\_kws: keyworded arguments
Additional parameters for the chosen method.

## adjph(self, p0=None, p1=None, pv=None, update=True)

Adjusts the phases of the spectrum according to the given parameters, or interactively if they are left as default. Calls for processing.ps

#### Parameters:

- p0: float or None 0-th order phase correction /°
- p1: float or None 1-st order phase correction /°
- pv: float or None 1-st order pivot /ppm
- update: bool
  Choose if you want to update the procs dictionary or not

# baseline\_correction(self, basl\_file='spectrum.basl', winlim=None)

Correct the baseline of the spectrum, according to a pre-existing file or interactively. Calls processing.baseline\_correction or processing.load\_baseline

- basl\_file: str
  Path to the baseline file. If it already exists, the baseline will be built according to this file;
  otherwise this will be the destination file of the baseline.
- winlim: tuple or None Limits of the baseline. If it is None, it will be interactively set. If basl\_file exists, it will be read from there. Else, (ppm1, ppm2).

# basl(self, from procs=False, phase=True)

Apply the baseline correction by subtracting self.baseline from self.S. Then, self.S is unpacked in self.r and self.i

#### Parameters:

- from\_procs: bool

  If True, computes the baseline using the polynomion model reading self.procs['basl\_c'] as coefficients
- phase: bool
  Choose if to apply the same phase correction of the spectrum to the baseline. This should be done if the baseline was computed before the phase adjustment!

## cal(self, offset=None, isHz=False, update=True)

Calibrates the ppm and frequency scale according to a given value, or interactively. Calls processing.calibration

#### Parameters:

- offset: float or None scale shift value
- isHz: bool

  True if offset is in frequency units, False if offset is in ppm
- update: bool Choose if to update the procs dictionary or not

## convdta(self, scaling=1)

Call processing.convdta using self.acqus['GRPDLY']

## integrate(self, lims=None)

Integrate the spectrum with a dedicated GUI. Calls fit.integrate and writes in self.integrals with keys [ppm1:ppm2]

• lims: tuple
Integrates from lims[0] to lims[1]. If it is None, calls for interactive integration.

## inv process(self)

Performs the inverse processing of the spectrum according to the given parameters. Overwrites the S attribute!! Calls processing.inv\_fp

## mc(self)

Calculates the magnitude of the spectrum and overwrites self.S, self.r, self.i

# plot(self, name=None, ext='png', dpi=600)

Plots the real part of the spectrum.

## Parameters:

• name: strFilename for the figure. If None, it is shown instead.

• ext: strFormat of the image

• dpi: *int*Resolution of the image in dots per inches

## process(self, interactive=False)

Performs the processing of the FID. The parameters are read from self.procs. Calls processing.interactive\_fp or processing.fp using self.acqus and self.procs Writes the result is self.S, then unpacks it in self.r and self.i Calculates frequency and ppm scales. Also initializes self.F with fit.Voigt Fit class using the current parameters

## **Parameters:**

• interactive: bool

True if you want to open the interactive panel, False to read the parameters from self.procs.

## qfil(self, u=None, s=None)

Gaussian filter to suppress signals. Tries to read self.procs['qfil'], which is { 'u': u, 's': s } Otherwise, these are set interactively by processing.interactive\_qfil and then added to self.procs. Calls processing.qfil

- u: *float*Position of the filter /ppm
- s: float
  Width (standard deviation) of the filter /ppm

# read procs(self, other dir=None)

Reads the procs dictionary from a file named 'filename.procs' in the same directory of the input file.

#### **Parameters:**

• other\_dir: str or None
Different location for the procs dictionary to look into. If None, self.datadir is used instead.
W! Do not put the trailing slash!

#### **Returns:**

• procs: *dict*Dictionary of processing parameters

# rpbc(self, \*\*rpbc kws)

Computes the phase angles and the baseline using processing.RPBC on self.S. Then applies the phase correction and subtracts the baseline, automatically. The procs dictionary is then updated and saved. The polynomial baseline is computed according to the given coefficients and stored in self.baseline

#### Parameters:

• rpbc\_kws: keyworded arguments See processing.RPBC for details.

# write acqus(self, other dir=None)

Write the acqus dictionary in a file named 'filename.acqus'. Calls misc.write acqus 1D

## Parameters:

• other\_dir: str or None Different location for the acqus dictionary to write into. If None, self.datadir is used instead.

# write\_integrals(self, other\_dir=None)

Write the integrals in a file named '{self.filename}.int'.

• other\_dir: str or None Different location for the integrals file to write into. If None, self.datadir is used instead.

## write procs(self, other dir=None)

Writes the actual procs dictionary in a file named 'filename.procs' in the same directory of the input file.

#### Parameters:

• other\_dir: str or None Different location for the procs dictionary to write into. If None, self.datadir is used instead. W! Do not put the trailing slash!

# write\_ser(ser, acqus, path=None)

Writes a real/complex array in binary format. Calls misc.write\_ser. Be sure that acqus contains the BYTORDA and DTYPA keys. See misc.write\_ser to understand the meaning of these values.

- ser: *ndarray*Array that you want to convert in binary format.
- acqus: dict
  Dictionary of acquisition parameters. It must contain BYTORDA and DTYPA.
- path: strPath where to save the binary file.

class

# 3.6.3 Spectra.Spectrum 2D

Class: 2D NMR spectrum

Attributes:

• datadir: str

Path to the input file/dataset directory

 $\bullet$  filename: str

Base of the name of the file, without extensions

• fid: 2darray

FID

• acqus: dict

Dictionary of acqusition parameters

• ngdic: dict

Created only if it is an experimental spectrum. Generated by nmrglue.bruker.read, contains all the information on the spectrometer and on the spectrum.

• procs: dict

Dictionary of processing parameters

• eaeflag: int

If FnMODE is Echo-Antiecho, keeps track of the manipulation of the data so to not repeat the same process twice

• S: 2darray

Complex (or hypercomplex, depending on FnMODE) spectrum

• rr: 2darray

Real part F2, real part F1

• ii: 2darray

Imaginary part F2, imaginary part F1

• ir: 2darray

Real part F2, imaginary part F1. Only exist if F1 is acquired in phase-sensitive mode

• ri: 2darray

Imaginary part F2, real part F1. Only exist if F1 is acquired in phase-sensitive mode

• freq f1: 1darray

Frequency scale of the indirect dimension, in Hz

• freq f2: 1darray

Frequency scale of the direct dimension, in Hz

• ppm f1: 1darray

ppm scale of the indirect dimension

• ppm f2: 1darray

ppm scale of the direct dimension

• trf1: dict

Projections of the indirect dimension, as 1darrays. Keys: 'ppm f2' where they were taken

- trf2: dict
  Projections of the direct dimension, as 1darrays. Keys: 'ppm f1' where they were taken
- Trf1: dict
  Projections of the indirect dimension, as pSpectrum\_1D objects. Keys: 'ppm\_f2' where they
  were taken
- Trf2: dict Projections of the direct dimension, as pSpectrum\_1D objects. Keys: 'ppm\_f1' where they were taken
- integrals: *dict*Dictionary where to save the regions and values of the integrals.

#### Methods:

\_\_init\_\_(self, in\_file, pv=False, isexp=True, is\_pseudo=False)

Initialize the class.

#### Parameters:

- in\_file: str path to file to read, or to the folder of the spectrum
- pv: bool

  True if you want to use pseudo-voigt lineshapes for simulation, False for Voigt
- isexp: bool

  True if this is an experimental dataset, False if it is simulated
- is\_pseudo: bool True if it is a pseudo-2D. Legacy option

# adjph(self, p01=None, p11=None, pv1=None, p02=None, p12=None, pv2=None, update=True)

Adjusts the phases of the spectrum according to the given parameters, or interactively if they are left as default. The non-interactive workflow is to apply processing.ps on F2, transpose according to FnMODE, apply processing.ps on F1, transpose back. If FnMODE is 'No', the phase correction is applied only on F2, as it should be done in a pseudo-2D experiment. Once self.S was updated and unpacked, the phase values are added to the procs dictionary to keep track of multiple phase adjustments.

- p01: float or None 0-th order phase correction /° of the indirect dimension
- p11: float or None 1-st order phase correction /° of the indirect dimension
- pv1: float or None 1-st order pivot /ppm of the indirect dimension

- p02: float or None 0-th order phase correction /° of the direct dimension
- p12: float or None 1-st order phase correction /° of the direct dimension
- pv2: float or None 1-st order pivot /ppm of the direct dimension
- update: bool Choose if to update the procs dictionary or not

## cal(self, offset=[None, None], isHz=False, update=True)

Calibration of the ppm and frequency scales according to a given value, or interactively. In this latter case, a reference peak must be chosen. Calls processing.calibration

#### Parameters:

- offset: tuple (scale shift F1, scale shift F2)
- isHz: tuple of bool

  True if offset is in frequency units, False if offset is in ppm
- update: bool
  Choose if to update the procs dictionary or not

## calf1(self, value=None, isHz=False)

Calibrates the ppm and frequency scale of the indirect dimension according to a given value, or interactively. Calls self.cal on F1 only.

## Parameters:

- value: float or None scale shift value
- isHz: bool

  True if offset is in frequency units, False if offset is in ppm

## calf2(self, value=None, isHz=False)

Calibrates the ppm and frequency scale of the direct dimension according to a given value, or interactively. Calls self.cal on F2 only

- value: float or None scale shift value
- isHz: bool

  True if offset is in frequency units, False if offset is in ppm

## convdta(self, scaling=1)

Calls processing.convdta to compensate for the group delay. It does not always work, depends on TopSpin version and planets alignment.

#### Parameters:

• scaling: *float*Scaling factor for processing convolta.

## eae(self)

Calls processing.EAE to shuffle the data and make a States-like FID. Sets self.eaeflag to 0.

## integrate(self, \*\*kwargs)

Integrates the spectrum with a dedicated GUI. Calls fit.integrate 2D

#### Parameters:

• kwargs: keyworded arguments
Additional parameters for fit.integrate 2D

## inv process(self)

Performs the inverse processing of the spectrum according to the given parameters. Overwrites the S attribute!! Calls inv xfb.

#### mc(self)

Computes the magnitude of the spectrum on self.S. Then, updates rr, ri, ir, ii.

## plot(self, Neg=True, lvl0=0.2)

Plots the real part of the spectrum. Use the mouse scroll to adjust the contour starting level.

#### Parameters:

- Neg: bool
  Plot (True) or not (False) the negative contours.
- lvl0: float
  Starting contour value with respect to the maximum of the spectrum

## process(self, interactive=False, \*\*int kwargs)

Performs the full processing of the FID on both dimensions. The parameters are read from self.procs. If FnMODE is Echo-Antiecho and you did not call self.eae before, the FID is converted to States with processing.EAE before to start. If interactive is True, calls processing.interactive\_xfb with int\_kwargs, else calls processing.xfb. The complex/hypercomplex spectrum is stored in self.S, then unpacked into self.rr, self.rr, self.ir, self.ii. If FnMODE is Echo-Antiecho, a phase correction of -90 degrees is applied on the indirect dimension.

- interactive: bool

  True if you want to open the interactive panel, False to read the parameters from self.procs.
- int\_kwargs: keyworded arguments
  Additional parameters for processing.interactive\_xfb, if interactive=True.

## projf1(self, a, b=None)

Calculates the sum trace of the indirect dimension, from a ppm to b ppm in F2. Store the trace in the dictionary trf1 and as 1D spectrum in Trf1. The key is 'a' or 'a:b' Calls misc.get\_trace on self.rr with column=True

#### Parameters:

- a: float ppm F2 value where to extract the trace.
- b: float or None.

  If it is None, extract the trace in a. Else, sum from a to b in F2.

## projf2(self, a, b=None)

Calculates the sum trace of the direct dimension, from a ppm to b ppm in F1. Store the trace in the dictionary trf2 and as 1D spectrum in Trf2. The key is 'a' or 'a:b' Calls misc.get\_trace on self.rr with column=False

#### Parameters:

- a: *float* ppm F1 value where to extract the trace.
- b: float or None.

  If it is None, extract the trace in a. Else, sum from a to b in F1.

## qfil(self, which=None, u=None, s=None)

Gaussian filter to suppress signals. Tries to read self.procs['qfil'], which is { 'u': u, 's': s } Otherwise, these are set interactively by processing.interactive\_qfil and then added to self.procs. Calls processing.qfil

- which: int or None Index of the F2 trace to be used for interactive\_qfil. If None, a suitable trace can be selected using misc.select traces.
- u: float
  Position /ppm
- s: float
  Width (standard deviation) /ppm

# read procs(self, other dir=None)

Reads the procs dictionary from a file named 'filename.procs' in the same directory of the input file.

#### Parameters:

• other\_dir: str or None
Different location for the procs dictionary to look into. If None, self.datadir is used instead.
W! Do not put the trailing slash!

## **Returns:**

• procs: *dict*Dictionary of processing parameters

## write acqus(self, other dir=None)

Write the acqus dictionary in a file named 'filename.acqus'. Calls misc.write acqus 1D

#### Parameters:

• other\_dir: str or None

Different location for the acqus dictionary to write into. If None, self.datadir is used instead.

# write\_integrals(self, other\_dir=None)

Write the integrals in a file named '{self.filename}.int'.

#### Parameters:

• other\_dir: str or None
Different location for the integrals file to write into. If None, self.datadir is used instead.

## write procs(self, other dir=None)

Writes the actual procs dictionary in a file named 'filename.procs' in the same directory of the input file.

#### Parameters:

• other\_dir: str or None
Different location for the procs dictionary to write into. If None, self.datadir is used instead.
W! Do not put the trailing slash!

## write ser(ser, acqus, path=None)

Writes a real/complex array in binary format. Calls misc.write\_ser. Be sure that acqus contains the BYTORDA and DTYPA keys. See misc.write\_ser to understand the meaning of these values.

• ser: *ndarray*Array that you want to convert in binary format.

• acqus: dict Dictionary of acquisition parameters. It must contain BYTORDA and DTYPA.

• path: strPath where to save the binary file.

## xf1(self)

Process only the indirect dimension. Transposes the spectrum in hypermode or normally if FnMODE!= QF, then calls for processing.fp using self.procs[keys][0], finally transposes it back. The result is stored in self.S, then self.rr and self.ii are written. freq\_fl and ppm\_fl are assigned with the indexes of the transients.

## xf2(self)

Process only the direct dimension. Calls processing.fp using procs[keys][1] The result is stored in self.S, then self.rr and self.ii are written. freq\_f1 and ppm\_f1 are assigned with the indexes of the transients.

# 3.6.4 Spectra.pSpectrum 1D

class

Subclass of Spectrum\_1D that allows to handle processed 1D NMR spectra. Useful when dealing with traces of 2D spectra. Shares the same attributes with Spectrum 1D.

#### Attributes:

• datadir: *str*Path to the input file/dataset directory

• filename: strBase of the name of the file, without extensions

• acqus: *dict*Dictionary of acqusition parameters

• ngdic: dict Created only if it is an experimental spectrum. Generated by nmrglue.bruker.read, contains all the information on the spectrometer and on the spectrum.

• procs: *dict*Dictionary of processing parameters

• S: *1darray*Complex spectrum

• r: 1darray
Real part of the spectrum

• i: *1darray*Imaginary part of the spectrum

• freq: *1darray*Frequency scale of the spectrum, in Hz

• ppm: *1darray* ppm scale of the spectrum

• F: fit. Voigt\_Fit object Used for deconvolution. See fit. Voigt fit.

• baseline: *1darray*Baseline of the spectrum.

• integrals: *dict*Dictionary where to save the regions and values of the integrals.

#### Methods:

\_\_init\_\_(self, in\_file, acqus=None, procs=None, istrace=False, filename='T')

Initialize the class.

- in\_file: str or 1darray
  If istrace is True, in\_file is the NMR spectrum. Else, it is the directory of the processed data.
- acqus: dict or None

  If istrace is True, you must supply the associated 'acqus' dictionary. Else, it is not necessary as it is read from the input directory
- procs: dict or None
  You can pass the dictionary of processing parameters, if you want. Otherwise, it is initialized with standard values.
- istrace: bool

  Declare the object as trace extracted from a 2D (True) or as true experimental spectrum (False)
- filename: strIf istrace is True, this will be the filename of self.acqus and self.procs

## acme(self, \*\*method kws)

Automatic phase correction based on entropy minimization It calculates the phase angles using the algorithm specified in method, then calls self-adjph with those values.

#### Parameters:

• method\_kws: keyworded arguments
Additional parameters for the chosen method.

## adjph(self, p0=None, p1=None, pv=None, update=True)

Adjusts the phases of the spectrum according to the given parameters, or interactively if they are left as default. Calls for processing.ps

#### Parameters:

- p0: float or None 0-th order phase correction /°
- p1: float or None 1-st order phase correction /°
- pv: float or None 1-st order pivot /ppm
- update: bool
  Choose if you want to update the procs dictionary or not

## baseline correction(self, basl file='spectrum.basl', winlim=None)

Correct the baseline of the spectrum, according to a pre-existing file or interactively. Calls processing baseline correction or processing load baseline

- basl\_file: str
  Path to the baseline file. If it already exists, the baseline will be built according to this file;
  otherwise this will be the destination file of the baseline.
- winlim: tuple or None Limits of the baseline. If it is None, it will be interactively set. If basl\_file exists, it will be read from there. Else, (ppm1, ppm2).

# basl(self, from procs=False, phase=True)

Apply the baseline correction by subtracting self.baseline from self.S. Then, self.S is unpacked in self.r and self.i

#### Parameters:

- from\_procs: bool

  If True, computes the baseline using the polynomion model reading self.procs['basl\_c'] as coefficients
- phase: bool
  Choose if to apply the same phase correction of the spectrum to the baseline. This should be done if the baseline was computed before the phase adjustment!

## cal(self, offset=None, isHz=False, update=True)

Calibrates the ppm and frequency scale according to a given value, or interactively. Calls processing.calibration

#### Parameters:

- offset: float or None scale shift value
- isHz: bool

  True if offset is in frequency units, False if offset is in ppm
- update: bool
  Choose if to update the procs dictionary or not

## convdta(self, scaling=1)

Call processing.convdta using self.acqus['GRPDLY']

## integrate(self, lims=None)

Integrate the spectrum with a dedicated GUI. Calls fit.integrate and writes in self.integrals with keys [ppm1:ppm2]

• lims: tuple
Integrates from lims[0] to lims[1]. If it is None, calls for interactive integration.

## inv process(self)

Performs the inverse processing of the spectrum according to the given parameters. Overwrites the S attribute!! Calls processing.inv\_fp

## mc(self)

Calculates the magnitude of the spectrum and overwrites self.S, self.r, self.i

# plot(self, name=None, ext='png', dpi=600)

Plots the real part of the spectrum.

#### Parameters:

• name: *str*Filename for the figure. If None, it is shown instead.

• ext: strFormat of the image

• dpi: *int*Resolution of the image in dots per inches

## process(self, interactive=False)

Performs the processing of the FID. The parameters are read from self.procs. Calls processing.interactive\_fp or processing.fp using self.acqus and self.procs Writes the result is self.S, then unpacks it in self.r and self.i Calculates frequency and ppm scales. Also initializes self.F with fit.Voigt Fit class using the current parameters

## **Parameters:**

• interactive: bool

True if you want to open the interactive panel, False to read the parameters from self.procs.

## qfil(self, u=None, s=None)

Gaussian filter to suppress signals. Tries to read self.procs['qfil'], which is { 'u': u, 's': s } Otherwise, these are set interactively by processing.interactive\_qfil and then added to self.procs. Calls processing.qfil

- u: *float*Position of the filter /ppm
- s: float
  Width (standard deviation) of the filter /ppm

# read procs(self, other dir=None)

Reads the procs dictionary from a file named 'filename.procs' in the same directory of the input file.

#### **Parameters:**

• other\_dir: str or None
Different location for the procs dictionary to look into. If None, self.datadir is used instead.
W! Do not put the trailing slash!

#### **Returns:**

• procs: dict
Dictionary of processing parameters

# rpbc(self, \*\*rpbc kws)

Computes the phase angles and the baseline using processing.RPBC on self.S. Then applies the phase correction and subtracts the baseline, automatically. The procs dictionary is then updated and saved. The polynomial baseline is computed according to the given coefficients and stored in self.baseline

#### Parameters:

• rpbc\_kws: keyworded arguments See processing.RPBC for details.

# write acqus(self, other dir=None)

Write the acqus dictionary in a file named 'filename.acqus'. Calls misc.write acqus 1D

## Parameters:

• other\_dir: str or None Different location for the acqus dictionary to write into. If None, self.datadir is used instead.

## write\_integrals(self, other\_dir=None)

Write the integrals in a file named '{self.filename}.int'.

• other\_dir: str or None Different location for the integrals file to write into. If None, self.datadir is used instead.

## write procs(self, other dir=None)

Writes the actual procs dictionary in a file named 'filename.procs' in the same directory of the input file.

#### Parameters:

• other\_dir: str or None
Different location for the procs dictionary to write into. If None, self.datadir is used instead.
W! Do not put the trailing slash!

# write\_ser(ser, acqus, path=None)

Writes a real/complex array in binary format. Calls misc.write\_ser. Be sure that acqus contains the BYTORDA and DTYPA keys. See misc.write\_ser to understand the meaning of these values.

- ser: *ndarray*Array that you want to convert in binary format.
- acqus: dict
  Dictionary of acquisition parameters. It must contain BYTORDA and DTYPA.
- path: strPath where to save the binary file.

# 3.6.5 Spectra.pSpectrum 2D

class

Subclass of Spectrum\_2D that allows to handle processed 2D NMR spectra. Reads the processed spectrum from Bruker.

#### Attributes:

- datadir: *str*Path to the input file/dataset directory
- filename: strBase of the name of the file, without extensions
- acqus: *dict*Dictionary of acqusition parameters
- ngdic: dict Generated by nmrglue.bruker.read, contains all the information on the spectrometer and on the spectrum.
- procs: dict
  Dictionary of processing parameters
- S: 2darray Complex (or hypercomplex, depending on FnMODE) spectrum
- rr: 2darray Real part F2, real part F1
- ii: 2darray Imaginary part F2, imaginary part F1
- ir: 2darray
  Real part F2, imaginary part F1. Only exist if F1 is acquired in phase-sensitive mode
- ri: 2darray
  Imaginary part F2, real part F1. Only exist if F1 is acquired in phase-sensitive mode
- freq\_f1: 1darray
  Frequency scale of the indirect dimension, in Hz
- freq\_f2: 1darray
  Frequency scale of the direct dimension, in Hz
- ppm\_f1: 1darray ppm scale of the indirect dimension
- ppm\_f2: 1darray ppm scale of the direct dimension
- trf1: dict Projections of the indirect dimension, as 1darrays. Keys: 'ppm\_f2' where they were taken
- trf2: dict Projections of the direct dimension, as 1darrays. Keys: 'ppm\_f1' where they were taken

- Trf1: dict
  Projections of the indirect dimension, as pSpectrum\_1D objects. Keys: 'ppm\_f2' where they
  were taken
- Trf2: dict
  Projections of the direct dimension, as pSpectrum\_1D objects. Keys: 'ppm\_f1' where they
  were taken
- integrals: *dict*Dictionary where to save the regions and values of the integrals.

#### Methods:

init (self, in file)

Initialize the class.

#### **Parameters:**

• in\_file: str Path to the spectrum. Here, the 'pdata/#' folder must be specified.

# 

Adjusts the phases of the spectrum according to the given parameters, or interactively if they are left as default. The non-interactive workflow is to apply processing.ps on F2, transpose according to FnMODE, apply processing.ps on F1, transpose back. If FnMODE is 'No', the phase correction is applied only on F2, as it should be done in a pseudo-2D experiment. Once self.S was updated and unpacked, the phase values are added to the procs dictionary to keep track of multiple phase adjustments.

- p01: float or None 0-th order phase correction /° of the indirect dimension
- p11: float or None 1-st order phase correction /° of the indirect dimension
- pv1: float or None
  1-st order pivot /ppm of the indirect dimension
- p02: float or None 0-th order phase correction /° of the direct dimension
- p12: float or None 1-st order phase correction /° of the direct dimension
- pv2: float or None
  1-st order pivot /ppm of the direct dimension
- update: bool Choose if to update the procs dictionary or not

## cal(self, offset=[None, None], isHz=False, update=True)

Calibration of the ppm and frequency scales according to a given value, or interactively. In this latter case, a reference peak must be chosen. Calls processing.calibration

#### Parameters:

- offset: tuple (scale shift F1, scale shift F2)
- isHz: tuple of bool

  True if offset is in frequency units, False if offset is in ppm
- update: bool Choose if to update the procs dictionary or not

## calf1(self, value=None, isHz=False)

Calibrates the ppm and frequency scale of the indirect dimension according to a given value, or interactively. Calls self.cal on F1 only.

#### Parameters:

- value: float or None scale shift value
- isHz: bool

  True if offset is in frequency units, False if offset is in ppm

## calf2(self, value=None, isHz=False)

Calibrates the ppm and frequency scale of the direct dimension according to a given value, or interactively. Calls self.cal on F2 only

#### Parameters:

- value: float or None scale shift value
- isHz: bool

  True if offset is in frequency units, False if offset is in ppm

## convdta(self, scaling=1)

Calls processing.convdta to compensate for the group delay. It does not always work, depends on TopSpin version and planets alignment.

## Parameters:

• scaling: *float* Scaling factor for processing convdta.

## eae(self)

Calls processing. EAE to shuffle the data and make a States-like FID. Sets self.eaeflag to 0.

## integrate(self, \*\*kwargs)

Integrates the spectrum with a dedicated GUI. Calls fit.integrate\_2D

#### Parameters:

• kwargs: keyworded arguments
Additional parameters for fit.integrate 2D

## inv process(self)

Performs the inverse processing of the spectrum according to the given parameters. Overwrites the S attribute!! Calls inv xfb.

## mc(self)

Computes the magnitude of the spectrum on self.S. Then, updates rr, ri, ir, ii.

## plot(self, Neg=True, lvl0=0.2)

Plots the real part of the spectrum. Use the mouse scroll to adjust the contour starting level.

#### Parameters:

- Neg: bool
  Plot (True) or not (False) the negative contours.
- lvl0: float
  Starting contour value with respect to the maximum of the spectrum

## process(self, interactive=False, \*\*int kwargs)

Performs the full processing of the FID on both dimensions. The parameters are read from self.procs. If FnMODE is Echo-Antiecho and you did not call self.eae before, the FID is converted to States with processing.EAE before to start. If interactive is True, calls processing.interactive\_xfb with int\_kwargs, else calls processing.xfb. The complex/hypercomplex spectrum is stored in self.S, then unpacked into self.rr, self.ri, self.ir, self.ii. If FnMODE is Echo-Antiecho, a phase correction of -90 degrees is applied on the indirect dimension.

- interactive: bool

  True if you want to open the interactive panel, False to read the parameters from self.procs.
- int\_kwargs: keyworded arguments
  Additional parameters for processing.interactive\_xfb, if interactive=True.

## projf1(self, a, b=None)

Calculates the sum trace of the indirect dimension, from a ppm to b ppm in F2. Store the trace in the dictionary trf1 and as 1D spectrum in Trf1. The key is 'a' or 'a:b' Calls misc.get\_trace on self.rr with column=True

#### Parameters:

- a: *float* ppm F2 value where to extract the trace.
- b: float or None.

  If it is None, extract the trace in a. Else, sum from a to b in F2.

## projf2(self, a, b=None)

Calculates the sum trace of the direct dimension, from a ppm to b ppm in F1. Store the trace in the dictionary trf2 and as 1D spectrum in Trf2. The key is 'a' or 'a:b' Calls misc.get\_trace on self.rr with column=False

#### Parameters:

- a: *float* ppm F1 value where to extract the trace.
- b: float or None.

  If it is None, extract the trace in a. Else, sum from a to b in F1.

## qfil(self, which=None, u=None, s=None)

Gaussian filter to suppress signals. Tries to read self.procs['qfil'], which is { 'u': u, 's': s } Otherwise, these are set interactively by processing.interactive\_qfil and then added to self.procs. Calls processing.qfil

#### Parameters:

- which: int or None Index of the F2 trace to be used for interactive\_qfil. If None, a suitable trace can be selected using misc.select traces.
- u: float
  Position /ppm
- s: float
  Width (standard deviation) /ppm

## read procs(self, other dir=None)

Reads the procs dictionary from a file named 'filename.procs' in the same directory of the input file.

• other\_dir: str or None
Different location for the procs dictionary to look into. If None, self.datadir is used instead.
W! Do not put the trailing slash!

#### **Returns:**

• procs: *dict*Dictionary of processing parameters

## write acqus(self, other dir=None)

Write the acqus dictionary in a file named 'filename.acqus'. Calls misc.write\_acqus\_1D

#### Parameters:

• other\_dir: str or None

Different location for the acqus dictionary to write into. If None, self.datadir is used instead.

# write\_integrals(self, other\_dir=None)

Write the integrals in a file named '{self.filename}.int'.

#### Parameters:

• other\_dir: str or None
Different location for the integrals file to write into. If None, self.datadir is used instead.

## write\_procs(self, other\_dir=None)

Writes the actual procs dictionary in a file named 'filename.procs' in the same directory of the input file.

#### Parameters:

• other\_dir: str or None
Different location for the procs dictionary to write into. If None, self.datadir is used instead.
W! Do not put the trailing slash!

## write ser(ser, acqus, path=None)

Writes a real/complex array in binary format. Calls misc.write\_ser. Be sure that acqus contains the BYTORDA and DTYPA keys. See misc.write\_ser to understand the meaning of these values.

• ser: *ndarray*Array that you want to convert in binary format.

• acqus: dict
Dictionary of acquisition parameters. It must contain BYTORDA and DTYPA.

• path: strPath where to save the binary file.

## xf1(self)

Process only the indirect dimension. Transposes the spectrum in hypermode or normally if FnMODE!= QF, then calls for processing.fp using self.procs[keys][0], finally transposes it back. The result is stored in self.S, then self.rr and self.ii are written. freq\_f1 and ppm\_f1 are assigned with the indexes of the transients.

## xf2(self)

Process only the direct dimension. Calls processing.fp using procs[keys][1] The result is stored in self.S, then self.rr and self.ii are written. freq\_f1 and ppm\_f1 are assigned with the indexes of the transients.