

Table of Contents



Intro

Modular Instrument Design



Some General Utilities

Uniform Code for Files on
Different Computers
figure lists
Plotting **All** the data



nddata objects

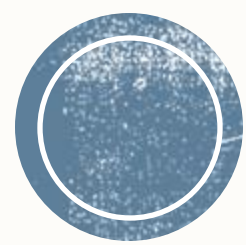
PySpecData nddata objects
PySpecData nddata objects
PySpecData nddata objects
Standard numpy arrays (ndarray
objects)
PySpecData nddata objects
Automatic Propagation of Error
Automatic Relabeling of axes
“Smart” Determination of
Dimensionality
Aliasing + Axis Registration
Methods
Methods



Some examples

Phasing Echo-like data
Fitting T_1 data
Some noisy RM data
Determining the transfer
function from the pulse
reflection

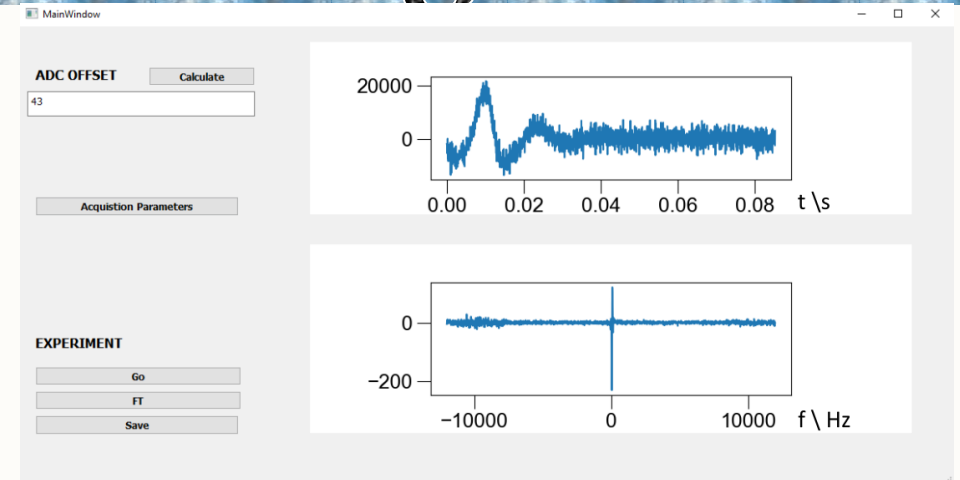
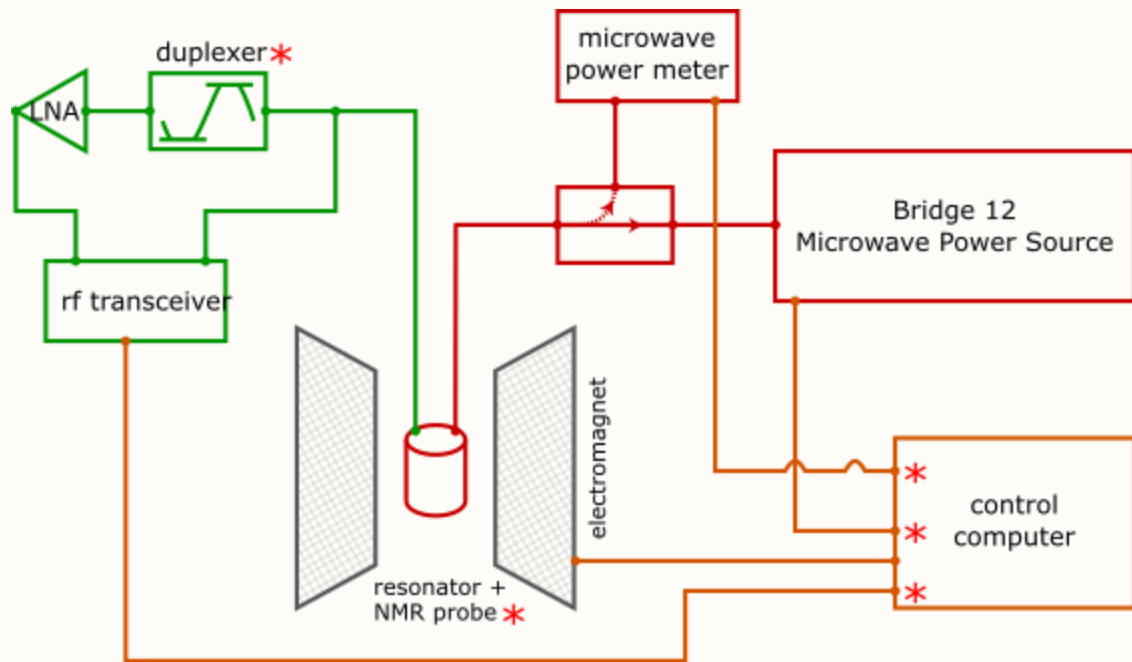




Intro



Modular Instrument Design



```
print("\nLOADING PULSE PROG...\n")
if phase_cycling:
    SpinCore_pp.load([
        ('marker', 'start', 1),
        ('phase_reset', 1),
        ('delay_TTL', deblank),
        ('pulse_TTL', p90, 'ph1', r_[0, 1, 2, 3]),
        ('delay', tau),
        ('delay_TTL', deblank),
        ('pulse_TTL', 2.0*p90, 'ph2', r_[0, 2]),
        ('delay', deadtime),
        ('acquire', acq_time),
        ('delay', repetition),
        ('jump_to', 'start')
    ])
```



pySpecData → our library (starting in Songi's lab, carried to Freed lab, continuing with it at Syracuse) → <http://github.com/jmfrancklab/pyspecdata>





Some General Utilities



Uniform Code for Files on Different Computers

Designed for integration with different computers:

```
d = find_file('200212_IR_3_30dBm', exp_type='test_equip')
```

(I have configured a data directory on my computer, and there is a subdirectory called `test_equip` that lives inside this → look for a file named `200212_IR_3_30dBm` inside that subdirectory).

Automatic filetype recognition.

If file is missing, generates a message on how to download the data:

```
Traceback (most recent call last):
  File "fitdata_test.py", line 82, in
<module>
    directory = getDATADIR(exp_type =
'test_equip' ))
  File
"c:\users\johnf\notebook\pyspecdata\pyspecda
ta\core.py", line 6629, in __init__
    check_only=True, directory=directory)
  File
"c:\users\johnf\notebook\pyspecdata\pyspecda
ta\core.py", line 1042, in h5nodebypath
    +errmsg)
AttributeError: You're checking for a node
in a file (200212_IR_3_30dBm.h5) that does
not exist
I can't find 200212_IR_3_30dBm.h5 in
C:\Users\johnf\exp_data\test_equip\, so I'm
going to search for it in your rclone remotes
checking remote g_syr:
You should be able to retrieve this file
with:
rclone copy -v --include
'200212_IR_3_30dBm.h5'
g_syr:exp_data/test_equip
C:\\Users\\johnf\\exp_data\\test_equip
```



figure lists

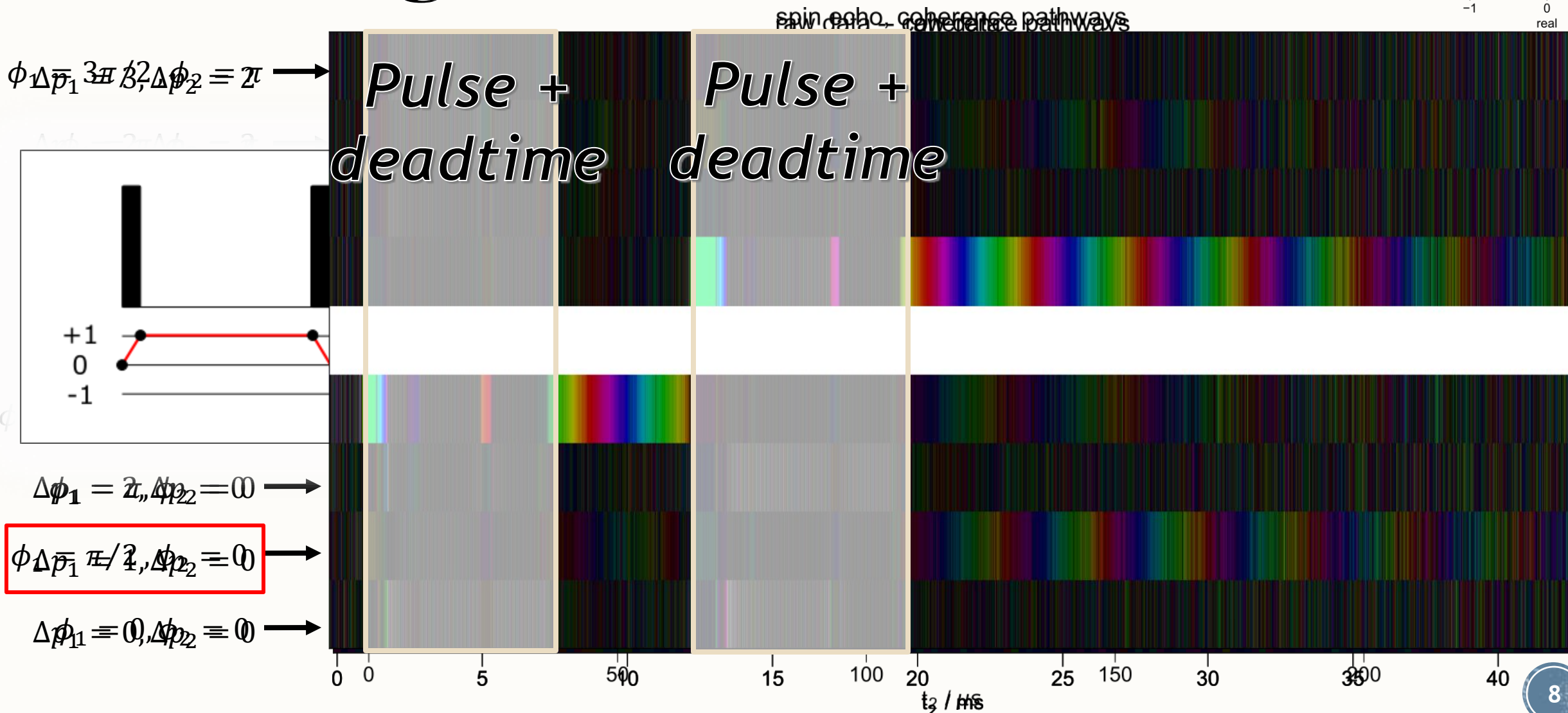
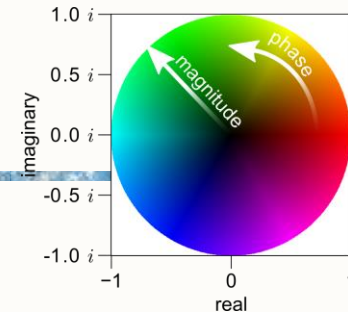
Add figures (and text) to a figure list:

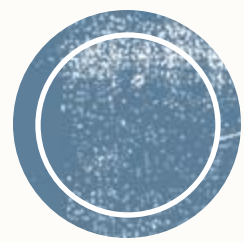
- keep track of units of different datasets added to the figure
- assign names (usually matching titles) to the plots
- use the same code to generate pop-up windows, or drop it into a PDF (latex) lab notebook

% pylab inline



Plotting all the data

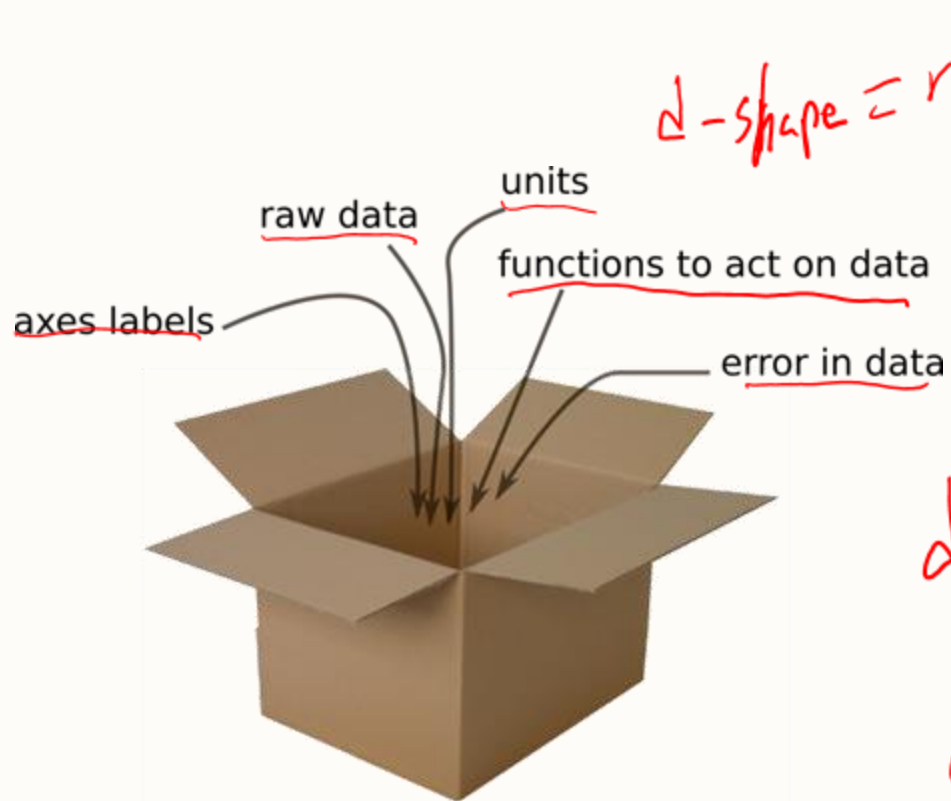




nddata objects



PySpecData nddata objects



$$d_shape = ndshape(d)$$

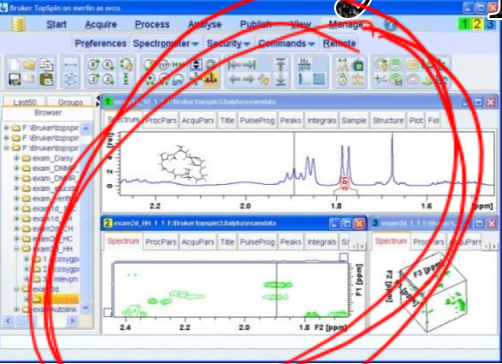
Store all the information we might want for a plot (and more) in one “container” (object) → call our container nddata.

$$d_shape += [(1, 10)]$$
$$d_combined = d_shape + ndshape(d)$$

$$d = nddata(\sim)$$



PySpecData ndarray objects



A screenshot of a database viewer showing a table of NMR data. The table has columns for 'Index', 'Name', 'Value', and 'Unit'. The data is organized into rows, with some rows highlighted in blue. The table contains numerical values and units, representing NMR data.

Index	Name	Value	Unit
1	1H	10.00000000	ppm
2	1H	9.99999999	ppm
3	1H	9.99999999	ppm
4	1H	9.99999999	ppm
5	1H	9.99999999	ppm
6	1H	9.99999999	ppm
7	1H	9.99999999	ppm
8	1H	9.99999999	ppm
9	1H	9.99999999	ppm
10	1H	9.99999999	ppm
11	1H	9.99999999	ppm
12	1H	9.99999999	ppm
13	1H	9.99999999	ppm
14	1H	9.99999999	ppm
15	1H	9.99999999	ppm
16	1H	9.99999999	ppm
17	1H	9.99999999	ppm
18	1H	9.99999999	ppm
19	1H	9.99999999	ppm
20	1H	9.99999999	ppm
21	1H	9.99999999	ppm
22	1H	9.99999999	ppm
23	1H	9.99999999	ppm
24	1H	9.99999999	ppm
25	1H	9.99999999	ppm
26	1H	9.99999999	ppm
27	1H	9.99999999	ppm
28	1H	9.99999999	ppm
29	1H	9.99999999	ppm
30	1H	9.99999999	ppm

Provides an opportunity to structure the data.

database

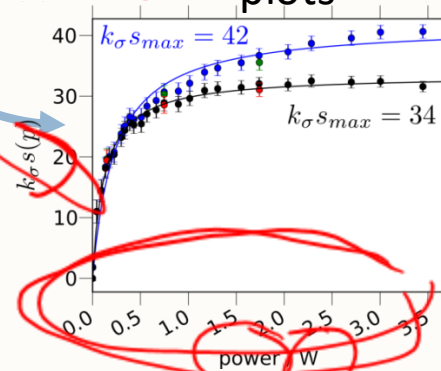
"Nd-data"



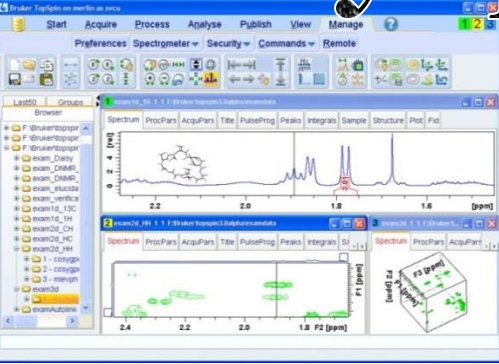
d

$plot(d)$

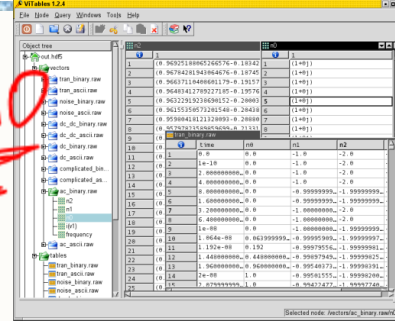
plots



PySpecData nddata objects



SC, gain = 10



Also → facilitate manipulation and acquisition of data

Technically:

- Methods ↵
- Operator overloading ↵
- Properties ↵

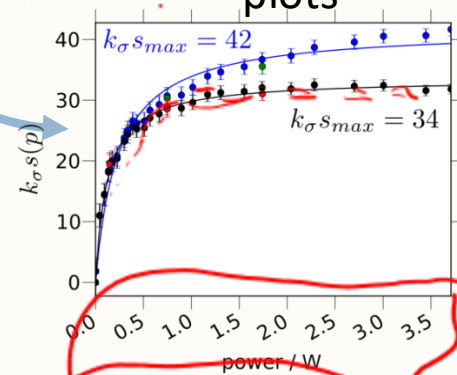
proprietary data formats

"No-data"

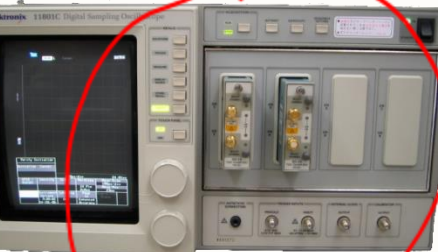
database

$$C = a + b$$

plots



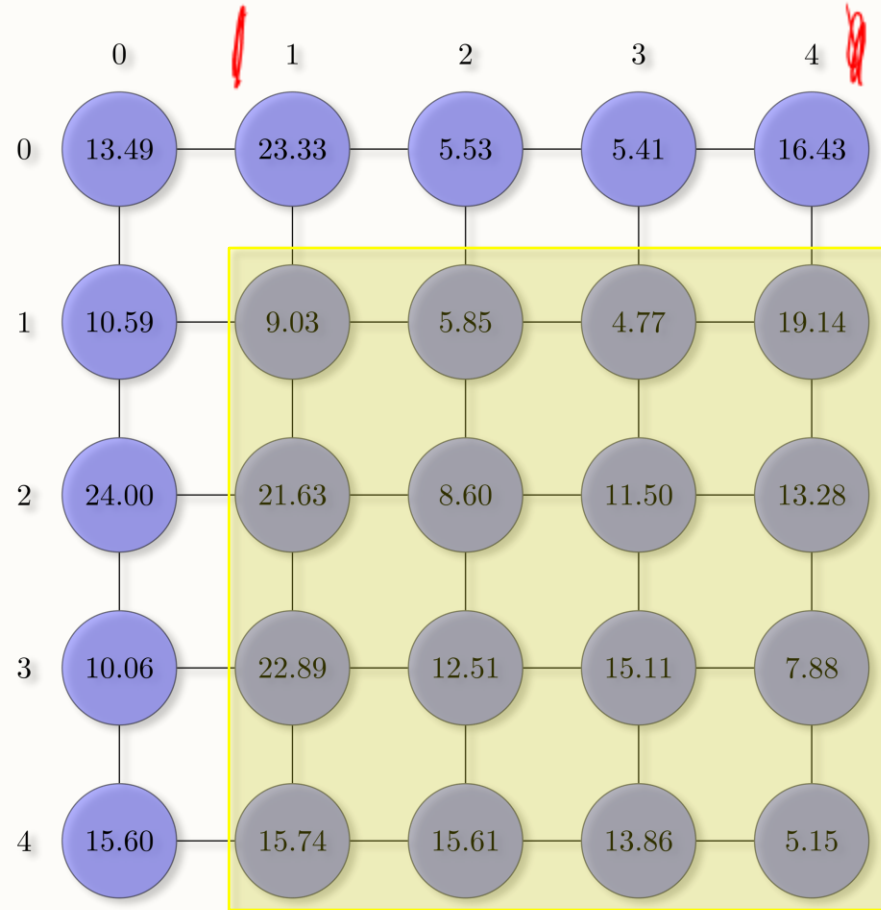
~~Fourier transform,~~
~~error propagation,~~
~~curve fit, etc.~~



instruments



Standard numpy arrays (ndarray objects)



Say this is called `d`, then

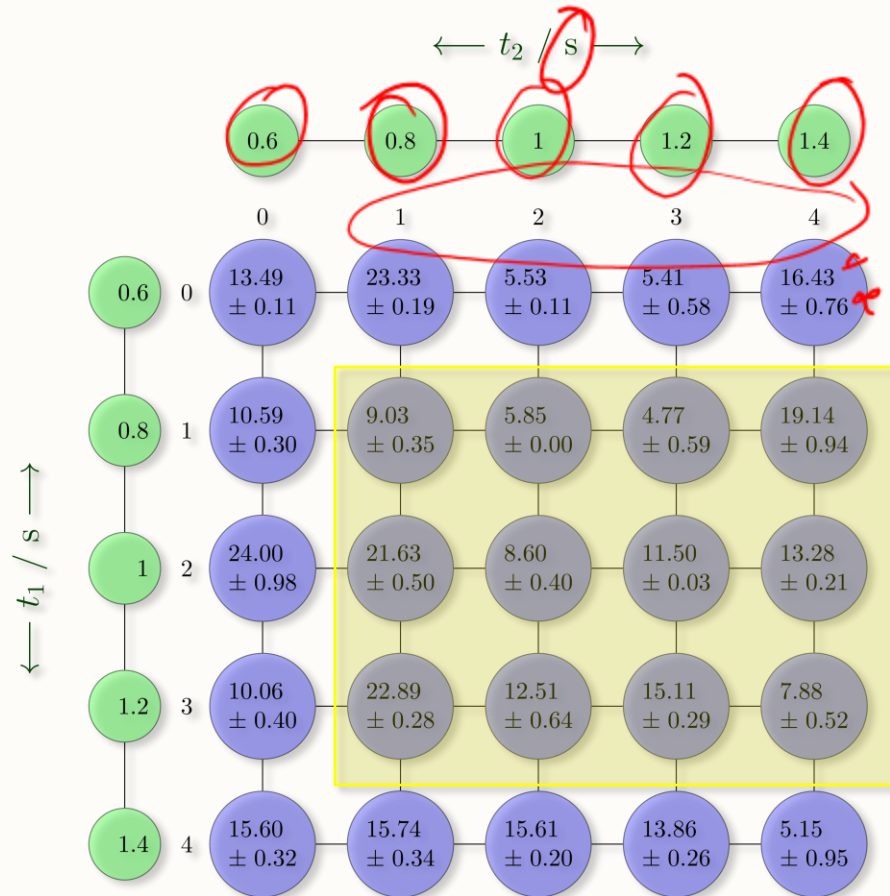
```
d_sliced = d[1:, 1:4]
```

slices out relevant data - e.g. for:

- frequency filtering
- selecting portions of time axis that actually contain data



PySpecData nddata objects



Named axes help avoid confusion:

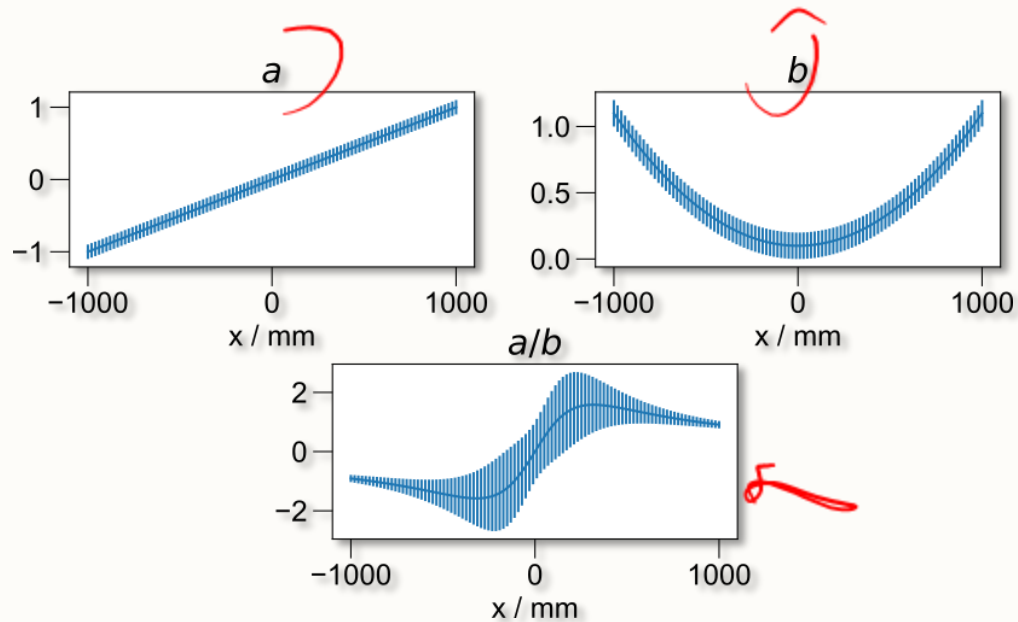
```
d_sliced =  
d['t1', 1:4]['t2', 1:]
```

We can slice by the *frequency* rather than the *index number*:

```
d_sliced =  
d['t1'; (0.8, 1.2)]['t2': (0.8,  
None)]
```



Automatic Propagation of Error

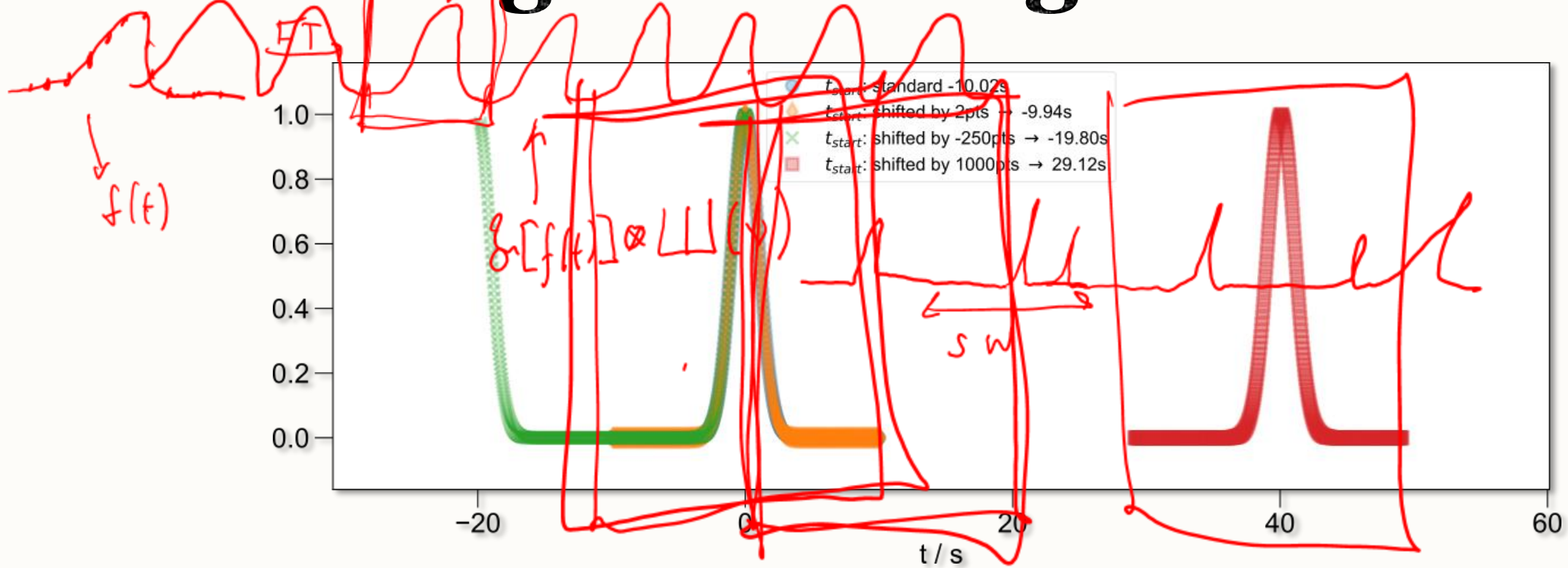


```
from pyspecdata import *
fl = figlist var()
a = nddata(r[-1:1:1j], 'x').set_units('x', 'm')
a.set_error(0.1)
b = a.C
b.run(lambda t: t**2+0.1)
b.set_error(0.1)
print(b.get_error('x'))
fl.next('$a$',
figsize=(4.5,2))
fl.plot(a)
fl.next('$b$',
figsize=(4.5,2))
fl.plot(b)
fl.next('aob',
figsize=(4.5,2))
fl.plot(a/b)
title('$a/b$')
fl.show()
```





Aliasing + Axis Registration



Powerful tools, such as looking at different “aliases”:

```
forplot.ft_clear_startpoints('t',  
t=new_startpoint, f='current')
```

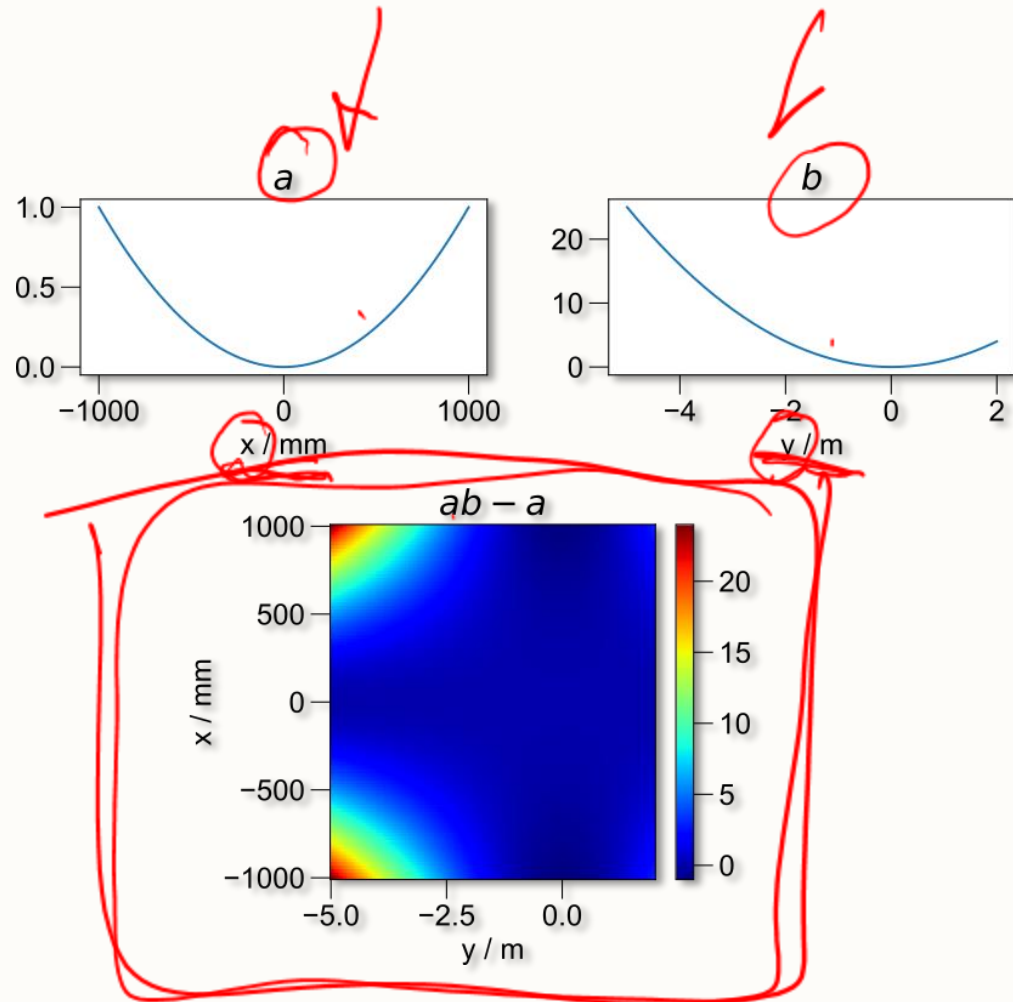
More specifically, we can relabel our axes and interpolate onto a new axis:

```
s.setaxis('t2', lambda x: x-peak_center)  
s.register_axis({'t2':0})
```

(first-order phase shift, but actually reflected in the data)



“Smart” Determination of Dimensionality



```
from pyspecdata import *
fl = figlist var()
a = nddata(r[-1:1:100j], 'x').set_units('x', 'm')
a.run(lambda t: t**2)
b = nddata(r[-5:2:200j], 'y').set_units('y', 'm')
b.run(lambda t: t**2)
fl.next('$a$',
figsize=(4.5, 2))
fl.plot(a)
fl.next('$b$',
figsize=(4.5, 2))
fl.plot(b)
fl.next('$ab-a$',
figsize=(4.5, 4))
fl.image(a*b-a)
fl.show()
```



Methods

ph1 0 1 2 3 0 1 2 3 0 1 2 3 0 1 2 3
ph2 0 0 0 0 1 1 1 1 2 2 2 2 3 3 3 3

These are all *methods* → they appear to be *inside* the data, e.g.:

```
d = d.copy()
```

Manipulating attributes of the data:

- copy
- copy_props → copy the “other info”
- get_units → units of axis
- replicate_units
- set_units
- setaxis → manipulate the axis
 - can take a function as an argument

```
s.setaxis('t2', lambda x: x - peak_center)
```
- set_prop → properties in a generic “other info” dictionary
- copyaxes
- get_error
- get_plot_color
- get_prop → opposite of “set_prop”

- ~~hdf5 write~~
- like → “ones-like”
- labels → set multiple axis labels w/ a dictionary
- name → (of the data) used for HDF5 + plotting
- set_error

Reshaping:

- ~~chunk~~ → split dimension
- ~~chunk_auto~~ → split dimension based on record array
- item → single element to float or int
- rename → dimension label
- reorder
- smoosh → opposite of “chunk”
- squeeze → remove dimensions of size 1
- extend → zero padding

`detach('ph1', ['ph2', 'ph1'], [4, 4])`



Methods

Utility Functions (benefited by naming):

- `argmax, argmin` → yields axis values (frequency/time)
- `contiguous` → takes a logical function, yields axis values (e.g. peak picking)
- `dot`
- `fromaxis` → generate an `nddata` from an axis
- `integrate` → uses x axis for numerical integration
- `interp, invinterp`
- `polyfit`
- *(overloaded)*: `+` `-` `/` `exp` and trig functions
 - error propagation
 - automatic reordering

General utilities:

- `add_noise` → for simulation
- `cdf` → cumulative dist. func.
- `circshift`
- `cropped_log`
- `mean, mean_all_but, mean_nopop, mean_weighted`
- `run, run_nopop, runcopy` → apply numpy functions in-place
- `sum, sum_nopop, diff`
- `sort`
- `to_ppm`



Methods

Plotting:

- contour
- histogram
- human_units
- matrices_3d
- mayavi_surf
- meshplot
- oldtimey
- unify_axis
- plot_labels
- waterfall
- set_plot_color

Fourier:

- convolve

- fourier_shear
- ft
- ft_clear_startpoints
- ftshift
- get_ft_prop
- ift
- register_axis
- set_ft_prop



Methods

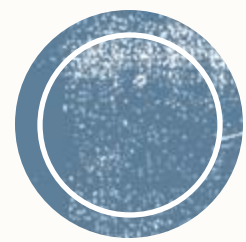
More internal/development:

- aligndata → match shapes
- axn → (internal) number of named axis
- fld → generate dict of info about axes
- getaxis → ndarray for axis of a given dimension
- mkd ~nddata.popdim ~nddata.set_to

More complicated manipulations:

- extend_for_shear
- inhomog_coords
- linear_shear
- nnls → ILT
- secsy_transform
- secsy_transform_manual
- shear

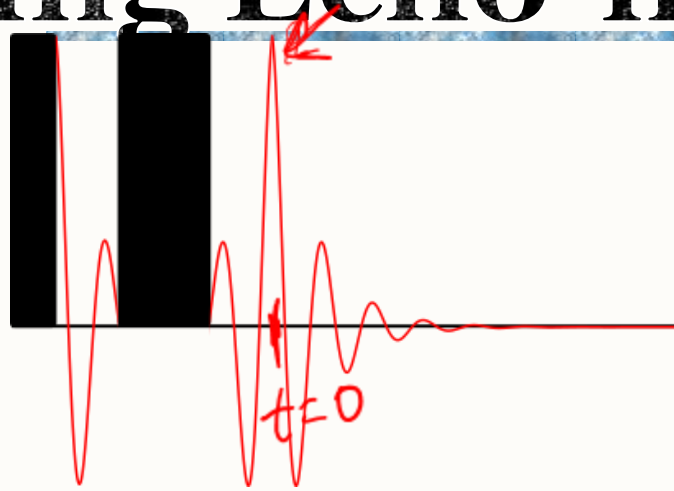




Some examples

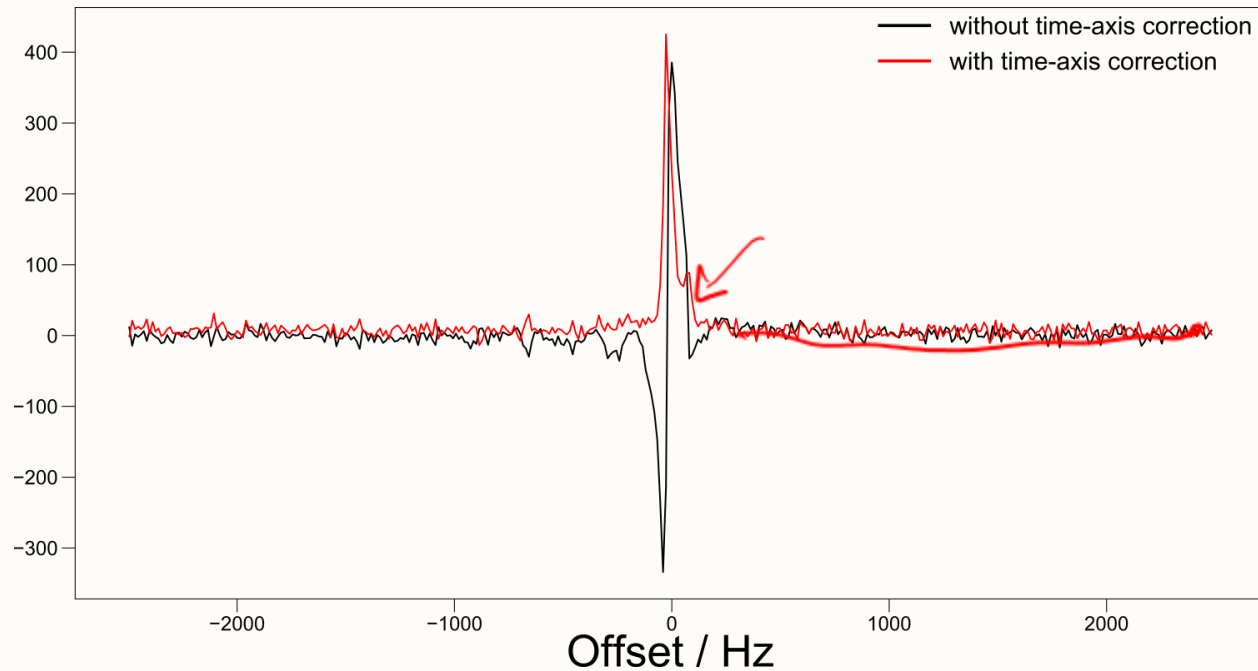


Phasing Echo-like data

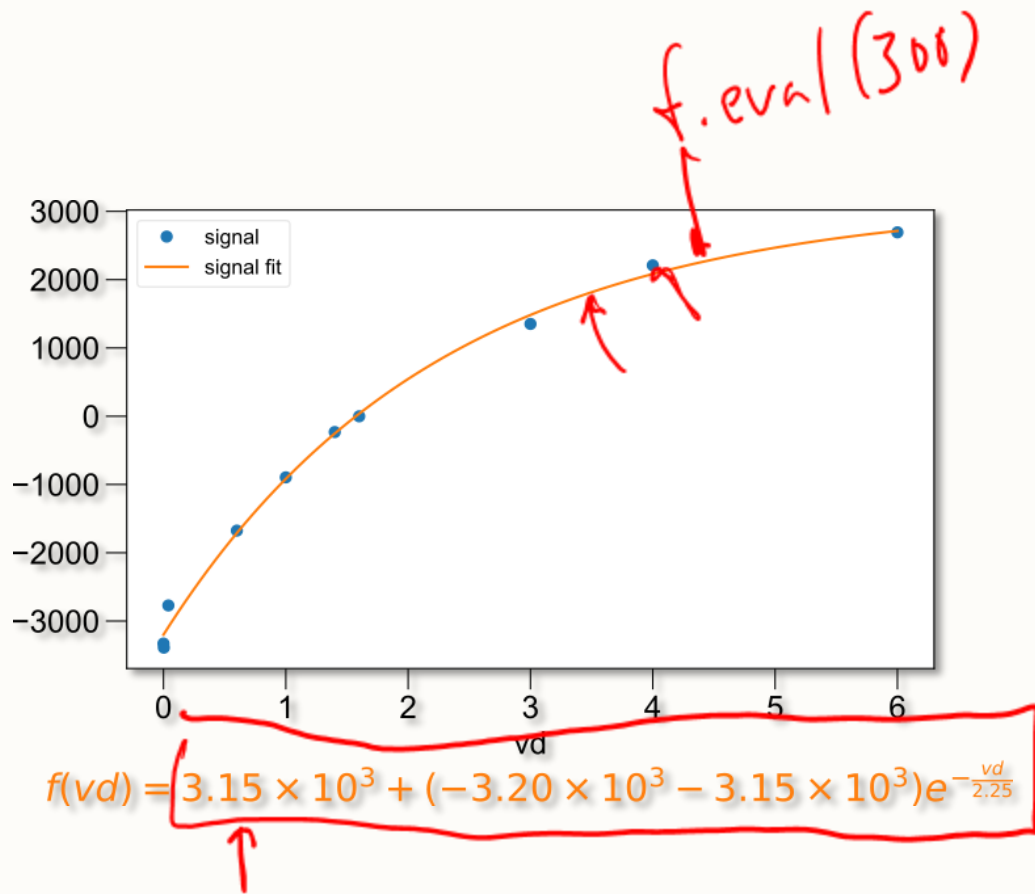


accomplished with:

```
s.setaxis('t2', lambda x: x -  
            peak_center)  
s.register_axis({'t2': 0})
```



Fitting T_1 data



def fitfunc(p, x):
 return p[0] + (p[1] - p[2])
 class fitdata inherits from nddata
 this is python for:

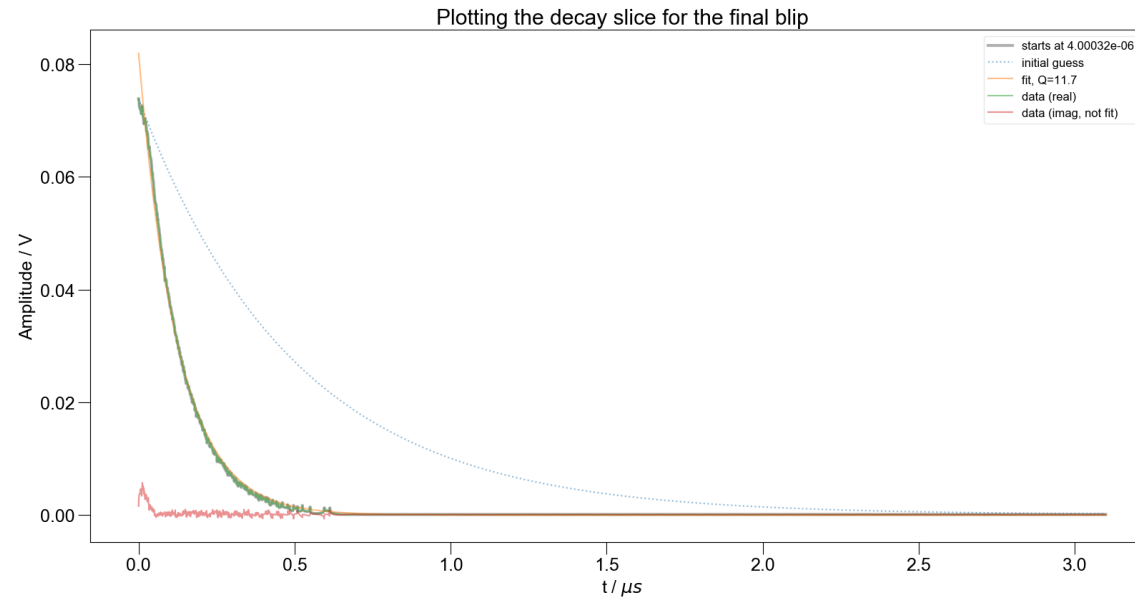
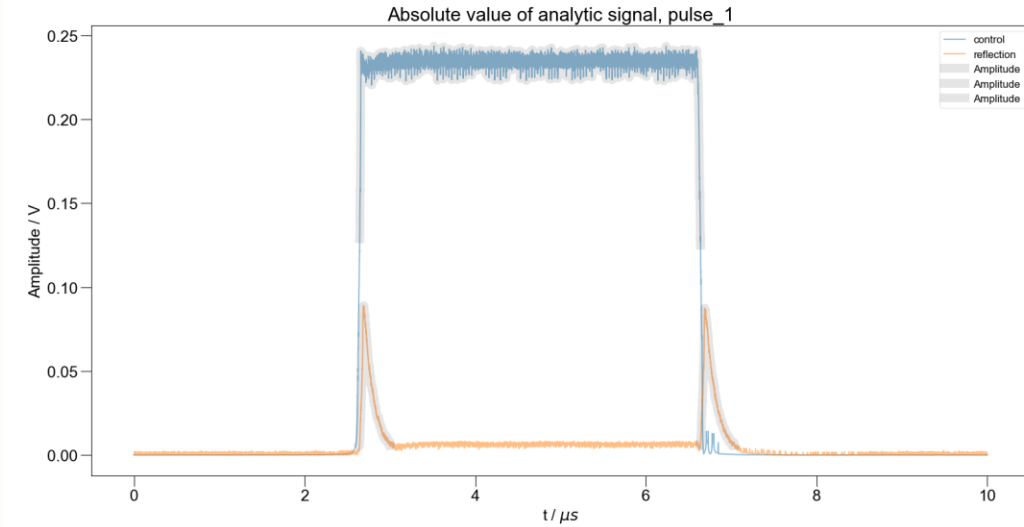
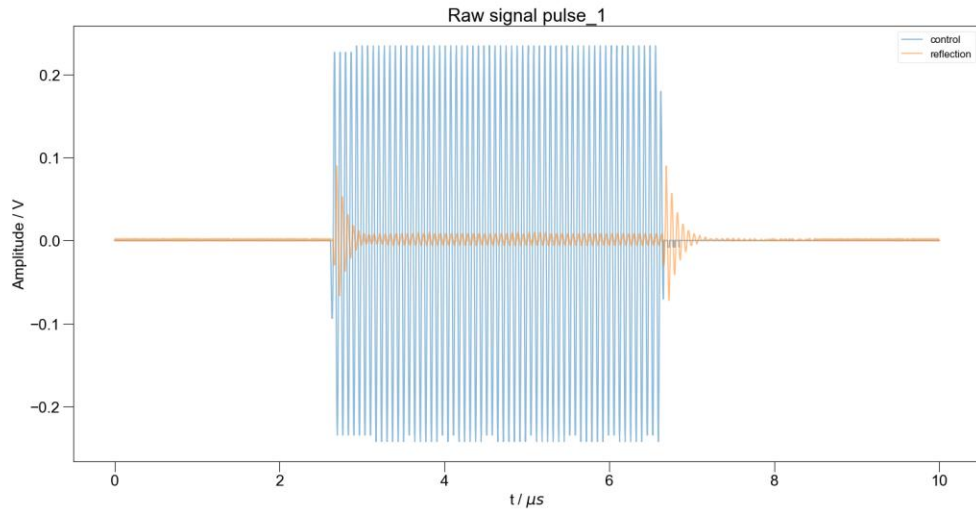
$f(vd) = M_{inf} + (M_0 - M_{inf}) e^{-\frac{vd}{T_1}}$
 f = fitdata(s_sliced)
 M0, Mi, T1, vd = sympy.symbols("M_0 M_inf T_1 vd",
 real=True)
 f.functional_form = Mi + (M0 - Mi) * sympy.exp(-vd/T1)

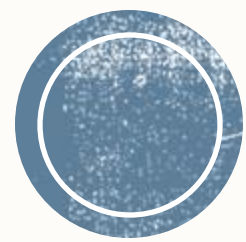
f.fit()

$$A e^{-R_2 t}$$



Determining the transfer function from the pulse reflection

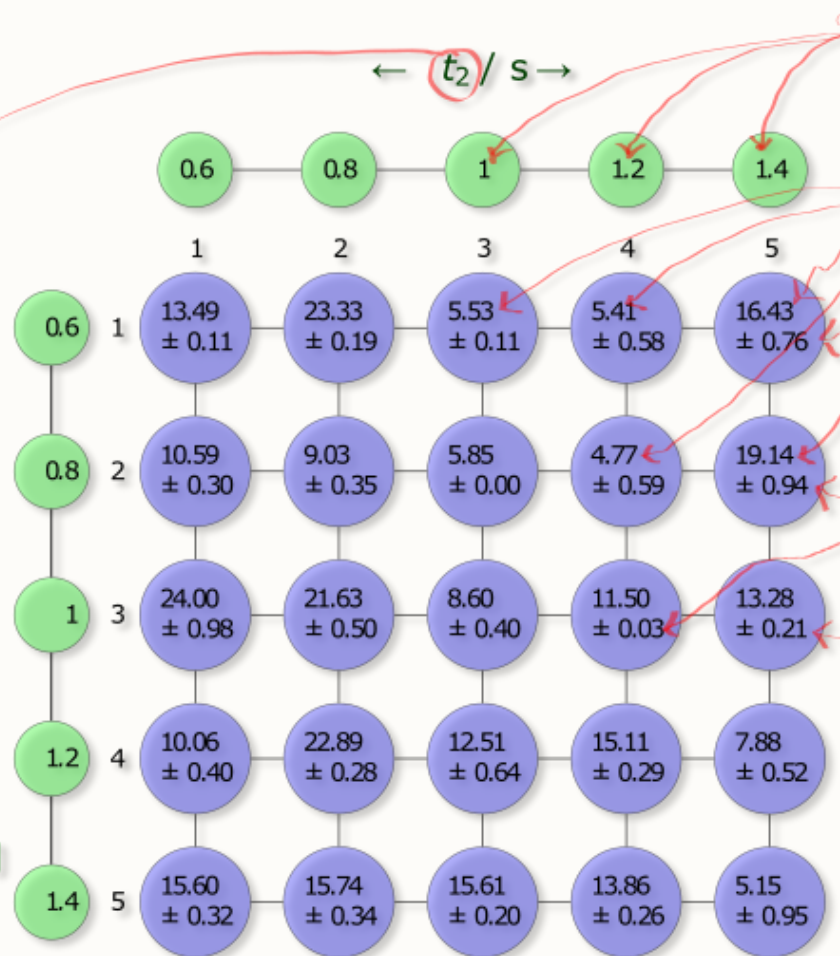




Future Directions



for these examples, say the instance of nddata is called "d"



green bracketed gives the type (ndarray is a standard numpy array)

proposed changes:

d.t2 to get the axis object, where "t2" in d.dimlabels
d.t2 will be same as d.axes["t2"]
d.t2 slices, indexes, and setslice like a normal numpy array
setting d.t2 will automatically determine d.t2.scalefunc and dx
addition, multiplication, division, log/exp/sin are overloaded
when multiplied by an nddata, d.t2 will behave like an nddata (d.fromaxis('t2'))
d.t2 will have attributes/properties -- only items in brown would be modifiable:
d.t2.units
d.t2.scalefunc ("linear" would be a scalar, for log spacing, the function log10, or else None)
d.t2.domain ("orig" vs. "trans" -- is t2 the original or transformed domain)
d.t2.status ("orig" vs. "trans" -- current status of the data)
d.t2.error
d.t2.dx (dwell time or frequency spacing or None if nonlinear)
d.t2.aliases (other names: ["t_2", "t"]) -- here d.t would be same as d.t2
d.t2.fancy_alias (for plotting)
d.t2.inv (d.nu2 -- simple name of the conjugate axis)
d.t2.transform name of method used to transform ("ft", "ift", "nnls", "nus", etc.)
also d.t2.do_transform()

d.getaxis('t2') [ndarray]

d.data [ndarray]

d.get_error() [ndarray]

d.set_error(r_[0.11, 0.19...0.95])

proposed changes:

d.error
-- e.g.: d.error = x_[0.11, 0.19...0.95] OR error_array = d.error

d.dimlabels [list of strings]

- Make axes into an object (see black text above)

- Clean up fitdata (symbolic math)

- Separate methods into categories

- Object-oriented figure list

d.set_units('t1', 's') [string]

d.get_units('t1')

see proposed changes top left

Figure list upgrades

Take advantage of “figure lists” while avoiding some of the disadvantages:

Build the new system around a list of *visualization objects*.

In the simplest example, a visualization object corresponds to e.g. a single matplotlib axis; but it can also correspond to multiple axes, as when we want to plot multi-dimensional data, or when we want to show the projections of 2D data.

By breaking the plotting process into several methods, we can automate the process, while also keeping it maximally flexible.

From the end-user’s perspective, plotting is achieved by

1. optionally modifying an existing plotting object class to get different properties than the default: the default classes available are image, plot, contour, and surface
2. creating instances of the plotting classes inside the figure list, in order of appearance (note that sometimes, it will make sense to first initialize blank plots, and drop data into them if the order of plotting and data generation are different)
3. adding nddata objects (or numpy objects, which are converted to nddata objects) to the plotting instances (quite literally)

```
with figlist_var('filename.pdf')  
as fl:  
    pl_first = fl.next()  
    pl_first += d_rough.real
```

4. if interactive plotting is employed, the nddata themselves are updated, and then the figurelist update method is called.

