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1 Introduction

The quantum fitter is developed to help academics implement different fit function with ease. The whole project is mainly based on [lmfit](#), which is already convenient to use. The lmfit has two occasionally use classes, one is [Model](#), and the other is [Parameter](#), respectively control model functions and parameter properties.

Before start, please make sure python 3.8+ and the packages listed are correctly installed in your python:

```
1 numpy - 1.20.1
2 matplotlib - 3.3.4
3 scipy - 1.6.2
4 lmfit - 1.0.2
```

Lower version of python and packages are not guaranteed for running the code, but it's worth a try.

After all the preparation, you can now implement

```
1 import quantum_fitter as qf
2 import numpy as np
```

then with x_data and y_data as numpy array

```
1 t5 = qf.Qfit(x_data, y_data, 'GaussianModel')
2 t5.do_fit()
3 t5.pretty_print()
```

to start your journey.

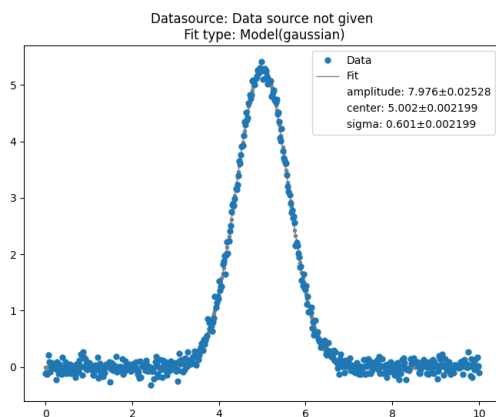


Figure 1: Pretty print for random generate Gaussian distribution

2 Functions

2.1 Qfit

```
1 class Qfit(data_x, data_y, model=None, params_init=None, method='leastsq')
```

The data have to be numpy array. Model parameter can be either function or build-in model string (see Appendix A). It's also possible to pass in a set or list of model, the Qfit will add the model up.

The params_init can be either dictionary with name and value (Recommends) or list along with sequence from your own function.

Method list can be found in [here](#) in the minimize method.

```
1 @property
2 def params(init_dict: dict)
```

Set and get initial parameters.

```
1 def set_params(self, name: str, value: float = None, vary: bool = True,
    minimum=None, maximum=None, expression=None, brute_step=None)
```

Set individual parameters' properties. See lmfit's [Parameter](#) to get more info.

```
1 def add_models(self, model, merge: str = '+')
```

Do models operation on existing functions. Merge can be +, -, *, /.

```
1 def wash(self, method='savgol', **kwargs):
```

To do filter on data (currently only [savgol](#) is implemented).

```
1 def do_fit()
```

Start fitting with current setting.

```
1 def pretty_print(self, plot_settings: dict=None)
```

Do print. Plot setting contains:

- **x_label**: Define the x label
- **y_label**: Define the y label
- **plot_title**: Define plot title
- **fit_color**: Define fit curve color
- **data_color**: Define data curve color
- **fig_size**: Define figure size
- **x_lim**: Define the x limit of the figure
- **y_lim**
- **show_fig**: Show figure or not. Default is None (which means show). Type in anything will halt this action.
- **return_fig**: Return fig, ax or not. Default is None, type in anything will return the variables.

A plot setting example:

```
1 plot_settings = {
2     'x_label': 'Time (us)',
3     'y_label': 'Voltage (mV)',
4     'plot_title': 'datasource',
5     'fit_color': 'C4',
6     'fig_size': (8, 6),
7 }
```

```
1 def pdf_print(self, file_dir, filename, plot_settings=None):
```

Output PDF to local storage.

```
1 def fit_params(name: str = None)
```

Return fit parameter's value or dict.

```
1 def err_params(name: str = None):
```

Return fit parameter's stderr value or dict.

```
1 def fit_values()
```

Return best fitting y values.

2.2 Others

```
1 def params(name: str)
```

Return build-in model's parameters.

A gaussian fit example:

```
1 import quantum_fitter as qf
2 import numpy as np
3
4 def gaussian(x, amp, cen, wid):
5     return (amp / (np.sqrt(2*np.pi) * wid)) * np.exp(-(x-cen)**2 / (2*wid
6         **2))
7
8 # Generate random number from gaussian distribution
9 x = np.linspace(0, 10, 500)
10 y = gaussian(x, 8, 5, 0.6) + np.random.randn(500) * 0.1
11
12 plot_set = {
13     'x_label': 'Time (us)',
14     'y_label': 'Voltage (mV)',
15     'plot_title': 'datasource',
16     'fit_color': 'C4',
17     'fig_size': (8, 6),
18 }
19
20 gm = qf.QFit(x, y, 'GaussianModel')
21 gm.do_fit()
22 gm.pretty_print(plot_set)
```

A Build-in models

Click [here to see](#) all the build-in models.

GaussianModel

LorentzianModel

SplitLorentzianModel

VoigtModel

PseudoVoigtModel

MoffatModel

Pearson7Model

StudentsTModel

BreitWignerModel

LognormalModel

DampedOscillatorModel

DampedHarmonicOscillatorModel

ExponentialGaussianModel

SkewedGaussianModel

SkewedVoigtModel

ThermalDistributionModel

DoniachModel