

PyPWA Tutorial – General Shell Fitting/Simulation - Version 2

Step-by-step guide to general fitting/simulation using PyPWA

(November 2015)

M. Jones and C. Salgado
Norfolk State University
and

The Thomas Jefferson National Accelerator Facility

All software used by GeneralShell PyPWA can be downloaded from <https://github.com/JeffersonLab/PyPWA/releases>.

Please take a look at the documentation at <https://pypwa.jlab.org/>

The usual problem to be solved by this software is the following: A model for the scattering amplitude (and therefore “intensity” = amplitude x amplitude*) that depends on some parameters (**params[]**) and some kinematical variables (**kVars[]**) is to be fitted to the data, therefore finding the parameters.

After the fitted parameters are obtained, we “simulate data” using the fitted parameters in the intensity formula and check that “simulated data” reproduces the kinematic properties observed in the “real data”.

NOTE: *It is advised that you read the general documentation in pypwa.jlab.org for more details.*

Below we detail step-by-step instructions for (fitting/simulations):

Previous Requirements (You'll need your own software for these steps):

1) Analyze your data to select the signal and create a file with all your signal events properties, such that if the events are characterized by the variables {x1...xn}, each line of a text file will have the structure:
x1=0.33,x2=0.0456,...,xn=0.78

This file can be named **data_events.txt**.

We also allow for the use of a Q factor (i.e. the probability for each event to be a signal, ie. $Q = \text{signal} / (\text{signal} + \text{background})$), to be included in the fit. Just create a file named **QFactor.txt** with the Q values of all events, one per line, written in the same order that the events are entered.

2) Run a full Monte Carlo simulation (generate+geant(detector + reconstruction + analysis) using a flat phase-space generator (you can or can't include your t distribution there or some other “weight” here). Create two (three) files, one with the generated and another with the accepted events, formatted as the data: for example, **gen_events.txt** with the generated events and **acc_events.txt** with all the events obtained after the full simulation. You may also need (if you want to simulate data) a pass/fail file (**events.pf**) with 0 (non-accepted) and 1 (accepted) events, a line for each event (txt).

You are ready to start using the PyPWA general fitting/simulation software.

Fitting and simulating on your desktop (these will be also work in the farm, scripts for the batch farm are being developed):

FITTING

[1] Create a directory to do your work.

Download tar (version 2) and run (from <https://github.com/JeffersonLab/PyPWA/>):

```
pip install PyPWA-2.0.0b0.tar.gz
```

(this will setup your software)

running in the ifarm (without previliges to write in system /bin) you need

```
/u/apps/anaconda/bin/pip install PyPWA-2.0.0-Beta0.0.2.tar.gz --user --install-option="--install-scripts={path to home folder}/bin"
```

[2] run **GeneralFitting -wc**

it generates two files `Example.py` where you define the function to be fitted.
`Example.yml` configuration file for the fits

[3] Edit those two files and rename them according to your analysis.

(note that if you run the simulation right after, the new configuration will overwrite your Example files so you must rename them now)

```
for example renam Example.py → test.py  
Example.yml → test.yml
```

For example (see full example code below):

```
def intFn
```

```
    To define the function  $I(x_1...x_n, a_1, ...a_m)$  (called here intensity)  
    that you will be fitted to the data. Each data event is defined by n  
    variables  $x_n$ , and the function depends on m parameters  $a_m$  that will  
    be the output of the fit .
```

[4] run **GeneralFitting test.yml**

This run your fits and get the Minuit results

Results are in files: **Vvalues.npy** and **minuitCov3.npy**

SIMULATION

[1] run **GeneralSimulator -wc**

produces also two file **Example.py**
Example.yml

The Example.py may be the same file you use before (i.e. test.py) if you just are trying to simulate with the same function you have fitted.

[2] edit those two files

[3] run **GeneralSimulator Example.yml**

Running the simulator produces a file with weights (i.e. weights.txt)

[4] run **python gamMasker.py [arguments]**

```
usage: gamMasker.py [-h] [-pf ACCEPTANCE_MASK] [-pfOut ACCEPTED_OUT]
                  [-w WEIGHTED_MASK] [-wOut WEIGHTED_OUT] [-b]
                  [-bOut BOTH_OUT] [-c]
                  file
```

A tool for producing gamp files from phase space using pre-calculated mask files.

positional arguments:

file The full filepath/name of the gamp or text file to be masked.

optional arguments:

-h, --help show this help message and exit
-pf ACCEPTANCE_MASK, --acceptance_mask ACCEPTANCE_MASK
 The full filepath/name of the pf acceptance (.txt) file to use.
-pfOut ACCEPTED_OUT, --accepted_out ACCEPTED_OUT
 The full filepath/name of the accepted output file.
-w WEIGHTED_MASK, --weighted_mask WEIGHTED_MASK
 The full filepath/name of the weight mask (.npy) file to use.
-wOut WEIGHTED_OUT, --weighted_out WEIGHTED_OUT
 The full filepath/name of the weighted output file.
-b, --both_masks Use both the acceptance and weighted masks.
-bOut BOTH_OUT, --both_out BOTH_OUT
 The full filepath/name of the accepted and weighted output file.
-c, --custom_mask Use a custom mask.

=====
=====>> **EXAMPLE**-----

file : test.py (with definition of function intFn)

```
-----  
import numpy,sys, os  
import math
```

```
def intFn(kVars,params): #You can change both the variable names and function name  
    the_size = len(kVars.values()[0]) #You can change the variable name here, or
```

```

set the length of values by hand
values = numpy.zeros(shape=the_size) # needed
for x in range(the_size):           # needed

    tDist= params['A1']
    wConst = (3.0/(4.0*math.pi))
    theta = math.acos(kVars["ctAD"][x])

    WUN = wConst*(0.5*(1-0.1786)+0.5*(3*0.17861)*math.cos(theta)**2-
math.sqrt(2.0)*0.02653*math.sin(2*theta)*math.cos(kVars['phiAD'][x])
+0.03021*(math.sin(theta))**2*math.cos(2*kVars['phiAD'][x]))

    # B. Kubis paper

    z = math.sqrt(kVars['x'][x]**2+kVars['y'][x]**2)
    #phi = math.atan2(kVars['y'][x],kVars['x'][x])

    FB = 1. + 2*params['A2']*z
    #+ 2*params['A3']*(z**(1.5))*math.sin(3*phi)
    #+2*params['A4']*z**2+2*params['A5']*(z**(5/2))*math.sin(3*phi)

    PP = kVars['P'][x]
    if PP <= 0.0:
        PP = .00001

    values[x] = tDist*WUN*PP*FB
return values

def the_setup(): #This function can be renamed, but will not be sent any
arguments.
    #This function will be ran once before the data is Minuit begins.

    Pass

```

file test.yml

Likelihood Information:

```

Generated Length : 1000000    #Number of Generated events
Function's Location : test.py  #The python file that has the functions in it
Processing Name : intFn      #The name of the processing function
Setup Name : the_setup      #The name of the setup function, called only once
before fitting
Data Information:
Data Location :
/volatile/clas/clasg12/salgado/Andres_omega/energybins/totaldat_events.txt #The
location of the data

```

```

Accepted Monte Carlo Location:
/volatile/clas/clasg12/salgado/Andres_omega/Fits/simulation/FLAT_values.txt #The
location of the Accepted Monte Carlo
QFactor List Location :
/volatile/clas/clasg12/salgado/Andres_omega/energybins/totalQF_events.txt #The
location of the Qfactors
Minuit's Settings:
Minuit's Initial Settings :
{'A1':1.E+4, 'fix_A1':False, 'error_A1':0.1, 'limit_A1':[0,10.E+12],
'A2':0.1, 'fix_A2':False, 'error_A2':0.01}
# 'A3':0.1, 'fix_A3':False, 'error_A3':0.01,
# 'A4':0.01, 'fix_A4':False, 'error_A4':0.01,
# 'A5':0.01, 'fix_A5':False, 'error_A5':0.01}

Minuit's Parameters: [ A1, A2] #The name of the Parameters passed to Minuit
Minuit's Strategy : 1
Minuit's Set Up: 0.5
Minuit's ncall: 1000
General Settings:
Number of Threads: 4 #Number of threads to use. Set to one for debug
Use QFactor: True #Boolean, using Qfactor or not

```

Then you run the fitting with:

GeneralFitting test.yml

you will get something like

```

-----
Parsing files into memory.

Loading users function.

Starting minimization.

*****
*                               *
*                               *
*****

-713205.759386
-713294.0393
-713117.445367
-713484.412273
-712926.766456
-713784.603913
-712623.502057
-713388.851401
-713022.327294
=====
-939196.612064
-1024410.19375

```

-1119061.50277

....

-1233198.54486
-1233198.53324
-1233198.53376
-1233198.54475
-1233198.54485
-1233197.4007

Elapsed Time: 311 sec Call: 135/1000 (2.3037 sec/call).

Est. time to maxcall: 1992.70 sec.

FCN=-1.233199e+06 FROM MIGRAD STATUS=VALID 135 CALLS

EDM=1.765887e-06 ERROR MATRIX ACCURATE POSDEF

NO.	NAME	VALUE	ERROR	FIXED
0	A1	1.969182e+09	1.730527e+07	
1	A2	3.009784e-01	1.148702e-02	

fval = -1233198.5452729429 | total call = 135 | ncalls = 135
edm = 1.7658866337391053e-06 (Goal: 5e-06) | up = 0.5

Valid	Valid Param	Accurate Covar	Posdef	Made Posdef
True	True	True	True	False
Hesse Fail	Has Cov	Above EDM		Reach callim
False	True	False	''	False

Name	Value	Para Err	Err-	Err+	Limit-	Limit+
0 A1 =	1.969E+09	1.731E+07			0	1E+13
1 A2 =	0.301	0.01149				

The values of the fit are saved in Vvalues.npy and the covariant matrix in minuitCov3.npy

=====

For simulation

I want to simulate events weighted by the amplitudes fitted before. Generate a phase-space MC in .txt (same variables as you have in your fit) or gamp (4-momenta information).

File Example.yml

Simulator Information:

Function's Location : test.py #The python file that has the functions in it

Processing Name : intFn #The name of the processing function

Setup Name : the_setup #The name of the setup function, called only once before fitting

Parameters : { A1: 1.969E+09, A2: 0.301}

Number of Threads: 4

Data Information:

Monte Carlo Location :

/volatile/clas/clasg12/salgado/Andres_omega/Fits/simulation/gen_events_small.txt

#The location of the MC

Save Location : weights.txt #Where you want to save the weights

run **GeneralSimulator -Example.yml**

this produce a file of weights : **weights.txt** with 0 (fail) and 1 (pass)

then we run the Masker:

for txt files:

python gampMasker.py -pf events.pf -pfOut pfEV.txt -w weights.txt -wOut wnEV.txt -b -bOut bEV.txt gen_events.txt

three files are generated with:

pfEV.txt : generated events weighted by acceptance only

wnEV.txt : generated events weighted by the amplitude function

bEV.txt : generated events weighted by amplitude and acceptance ("simulated data")

You can also mask a gamp file using:

python gampMasker.py -pf events.pf -pfOut pfEV.gamp -w weights.txt -wOut wnEV.gamp -b -bOut bEV.txt gen_events.gamp