

# **Introduction to SHERPA**

The Modeling and Fitting Tool  
of the CIAO Software System

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## What can you do in Sherpa?

- Standard PHA based analysis.
- Model data in many spectral bands simultaneously, e.g optical/X-rays.
- Access ATOMDB and GUIDE/ISIS for grating data analysis.
- Fit radial profiles.
- Simulate 1D data.
- Model 2D image data, e.g. fit surface brightness of the extended source.
- Get normalization of your PSF, while fitting the data with 1D/2D PSF.
- Use the PSF as a convolution kernel in the 2D image analysis (FFT or sliding cell).
- Convolution using the TCD library kernel.
- Use of exposure maps in the image analysis.
- Joint-mode data: spatial-spectral, spatial-timing
- Use scripts based on Sherpa only commands.
- Use S-lang on command line and in S-lang based scripts.
- Use your own models with User Models and S-lang user models.

## Standard PHA based analysis:

- **Source data:**
  - can be modeled in energy/wavelength space.
  - multiple data sets can be modeled with the same or different models in one Sherpa session.
  - data can be filtered on the command line, or from filter file.
- **Instrument responses (RMF/ARF):**
  - are entered independently from the source data.
  - one set of instrument responses can be read once and applied to multiple data sets.
  - several instrument responses and use in analysis of one source model or multiple data sets.
  - Future: multiple response files can be used in one source model expression.
- **Background files:**
  - are entered independently from the source data.
  - multiple background files can be used for one data set, e.g. grating analysis
  - the same background can be applied to multiple data sets.
  - background can be modeled independently of the source data, and have its separate instrument responses.
  - background can be modeled simultaneously with the source data.
  - background can be subtracted from the source data (subtract/unsubtract).

# Main **SHERPA** Components

- Data Input/Output.
- Visualization through ChIPS and ds9
- Model library and model language.
- Statistics and Error Analysis.
- Optimization Methods.

## Data Input/Output

- General use of data type and dimensionality.
- Supported types of files: ASCII, FITS binary tables and Images, PHA types I & II, IRAF IMH and QPOE files.
- Sherpa:
  - groups the data if appropriate;
  - treats integer, float or double precision data;
  - supports data of arbitrary dimensionality
- I/O interface through Data Model and Varmm
- Filtering while reading the data.
- Input data on the command line in two ways.

```
sherpa> data "image.fits[150:300,160:310]"
sherpa> show
Current Data Files:
Data 1: image.fits[150:300,160:310] fits.
Total Size: 22801 bins (or pixels)
Dimensions: 2
Size: 151 x 151
Total counts (or values): 20711 cts
```

or

```
sherpa> mydata=readfile("image.fits[150:300,160:310]")
sherpa> print(mydata)
filename          = image.fits
path              = /data
filter            = [150:300,160:310]
naxes             = 2
transform         = TAN
datatype          = Real4
pixels            = Float_Type[151,151]
crval             = Double_Type[2]
crpix             = Double_Type[2]
crdelt            = Double_Type[2]
sherpa> print(mydata.crval[0])
278.386
sherpa> print(mydata.crval[1])
-10.5899
```

# MODELS

- Three main type of models:

Source

Background

Instrument

- Model library consists of several models (plus XSPEC v.11) which can be used to define a **source** or **background** model.
- There are different types of instrument models:

RSP

PsfFromTCD

PsfFromFile

- **Instrument** models are **convolved** with **Source** and **Background** models before the model predicted data is compared with the observed data.

- Instrument and Background models are **NOT** required. Source models **have to be defined** for fitting.



RSP[rsp]

RMF file name:

ARF file name:

EEARF file name:

PSFfromTCD[psffromtcd]

Param	Type	Value	Min	Max
-----	-----	-----	---	---
1numCuts	frozen	1	1	1
2convTyp	frozen	1	1	2
3 xsize	frozen	4	1	1024
4 ysize	frozen	4	1	1024
5 nsigma	frozen	2	0.0100	100
6funcTyp	frozen	1	0	7
7 norm	frozen	1	0	3.4028e+38
8 dim	frozen	1	0	7

The Function Type is: Gaussian.

Dimension: 1

PSFfromFile[psffromFile]

Param	Type	Value	Min	Max
-----	-----	-----	---	---
1numCuts	frozen	1	1	1
2convTyp	frozen	1	1	2
3 file	string:			
4 xsize	frozen	32	1	1024
5 ysize	frozen	32	1	1024
6 xoff	frozen	0	-512	512
7 yoff	frozen	0	-512	512
8 xpos	thawed	512	1	1024
9 ypos	thawed	512	1	1024
10 norm	frozen	1	0	1000

# Model Language

- All predefined in model library models can be used in model expression to build a **source or background model**.
- Each library model can be given a **unique name** within Sherpa session.

```
sherpa> gauss1d[g1]
sherpa> source = ATTEN[att1]*BPL[b1]
att1.hcol parameter value [1e+20]
att1.heiRatio parameter value [0.1]
att1.heiiRatio parameter value [0.01]
b1.gamma1 parameter value [0]
b1.gamma2 parameter value [0]
b1.eb parameter value [100]
b1.ref parameter value [1]
b1.ampl parameter value [1]
```

- Model Parameters can be **linked** to other model parameters, arithmetic expression or other models.

```
sherpa> source = POLY[con]+gauss1d[g1]+gauss1d[g2]
sherpa> g1.ampl => 0.4*g2.ampl
```

or

```
sherpa> func = const1d[red]
sherpa> g1.pos => 0.568*func
```

- An argument of a model (e.g. energy) is defined as an expression in **Nested Models**.

Parameter Expression:

```
sherpa> Temperature = POLY
sherpa> BB.kT => Temperature
sherpa> show source
```

BB

```
bbody[BB] (integrate: on)
  Param  Type      Value      Min      Max
  -----  ----  -
  1      kT  link    varying  expression: Temperature
  2      ampl thawed      0.3      1e-20 3.4028e+38
```

Argument Expression:

```
sherpa> xenergy = SHLOG[mod]
sherpa> source = BB{xenergy}
sherpa> show source
```

BB{ xenergy }

```
bbody[BB] (integrate: on)
  Param  Type      Value      Min      Max
  -----  ----  -
  1      kT  thawed      0.3      0.1000 1000
  2      ampl thawed      0.001    1e-20 3.4028e+38
```

```
shloge[mod] (integrate: off)
```

```
  Param  Type      Value      Min      Max
  -----  ----  -
  1 offset frozen      0-3.4028e+38 3.4028e+38
  2 coeff frozen      1-3.4028e+38 3.4028e+38
  3      ampl frozen      1          0 3.4028e+38
```

- For **Joint-Mode** analysis one can apply models on each axis:

```

sherpa> DATA image.fits FITSIMAGE
sherpa> LORENTZ[SpatialAxis0](98:5:200, 70:50:90, 1:1:200)
sherpa> POWLAW1D[SpecAxis1]
sherpa> SRC = SpatialAxis0{x1}*SpecAxis1{x2}
sherpa> show source
(SpatialAxis0{ 0 } * SpecAxis1{ 1 })
lorentz1d[SpatialAxis0] (integrate: on)

```

	Param	Type	Value	Min	Max
	-----	-----	-----	---	---
1	fwhm	thawed	98	5	200
2	pos	thawed	70	50	90
3	ampl	thawed	1	1	200

```

powlaw1d[SpecAxis1] (integrate: on)

```

	Param	Type	Value	Min	Max
	-----	-----	-----	---	---
1	gamma	thawed	1.5	-10	10
2	ref	frozen	1-3.4028e+38	3.4028e+38	3.4028e+38
3	ampl	thawed	1	1e-20	3.4028e+38

# Fit Statistics in Sherpa:

A key feature of *Sherpa* is its large array of statistics appropriate for analyzing Poisson-distributed (*i.e.* counts) data.

- Statistics based on  $\chi^2$ :

- CHI GEHRELS
- CHI DVAR
- CHI MVAR
- CHI PARENT
- CHI PRIMINI

- Statistics based on the Poisson likelihood  $\mathcal{L}$ :

- CASH
- BAYES

If the data are not Poisson-distributed (*e.g.* fluxes), then alternatives include:

- least-squares fitting: setting all variances to one; or
- providing errors in an input file.

# $\chi^2$ -Based Statistics

The  $\chi^2$  statistic is

$$\chi^2 \equiv \sum_i \frac{(D_i - M_i)^2}{\sigma_i^2},$$

where

- $D_i$  represents the observed datum in bin  $i$ ;
- $M_i$  represents the predicted model counts in bin  $i$ ;  
and
- $\sigma_i^2$  represents the variance of the sampling distribution for  $D_i$ .

---

$$\chi^2 \text{ Statistic}$$
$$\sigma_i^2$$

GEHRELS:  $[1 + \sqrt{D_i + 0.75}]^2$

DVAR  $D_i$

MVAR  $M_i$

PARENT  $\frac{\sum_{i=1}^N D_i}{N}$

PRIMINI  $M_i$  from previous best-fit

# Likelihood-Based Statistics

The CASH statistic is

$$C \equiv 2 \sum_i [M_i - D_i \log M_i] \propto -2 \log \mathcal{L},$$

where

- $D_i$  represents the observed datum in bin  $i$ ;
- $M_i$  represents the predicted model counts in bin  $i$ ;  
and
- $\mathcal{L} = \prod_i \frac{M_i^{D_i}}{D_i!} \exp(-M_i)$ .