

National Aeronautics and Space Administration



Fermi

Gamma-ray Space Telescope

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Gamma-ray Space Telescope

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Advanced Likelihood

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Goals

- **Quality checks on spectral fitting of point sources**
 - **Major gotchas**
 - **Reminder of simple checks**
 - **Upper Limits**
- **Binned vs. Unbinned likelihood**

Major gotchas

- Flux/Spectral analysis depends critically on calculating the proper exposure

selection

livelime

response/exposure

minimization

gtselect **gtmktime**

gtlcube

gtexpmap

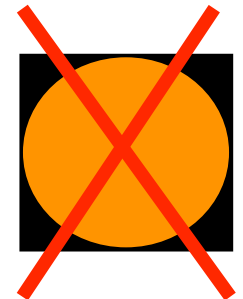
gtlike

(gtbin)

(gtsrcmaps)

(gtlike binned)

- Examples of things that can screw this up
 - fselect, fcopy**
 - these do not update the header keywords used in the exposure calculation
 - Mismatch of data selection and IRF set**
 - Use the diffuse class IRFs with the diffuse class event selection
 - Mismatch of ROI selection (gtselect) and data cube (gtbin) in binned likelihood analysis**



Major Gotchas II

- **Mismatch of calculated diffuse response and model diffuse components**
 - Use the recommended diffuse models with the data (includes precalculated diffuse response values for each photon for those specific models)
 - Diffuse response for experts
 - *gtdiffrsp* calculates the diffuse response values
 - Use unique names in the input xml model for different diffuse model templates
 - Example: If you come up with a new version of the Galactic diffuse template, don't call it "GAL_v02"
- The currently recommended isotropic template is only appropriate for use with the diffuse event class

Likelihood output - simple checks

*Did the fit work and does it make sense?
Reading the tea leaves of glike output*

- **Did the minimization converge?**
- **Are the number of predicted photons reasonable?**
- **Do the parameter values make sense?**
 - **Are values hitting limits?**
 - **Is there a source with an extremely soft spectrum or hard spectrum?**
- **Do the parameter errors make sense?**
 - **Too small? Were enough parameters left free?**
 - **Larger than the parameter values? Is the source significant?**
- **Consider the above for the target source and field sources**
- **All of the above become more critical for faint sources, complex regions, time-binned flux light curves...**

Gtlike bits

Convergence

```
Minuit did successfully converge.
# of function calls: 2401
minimum function Value: 2808753.9585
minimum edm: 0.74929079
```

Failed Convergence

```
WARNING: FunctionMinimum is invalid.

Exception encountered while minimizing objective function:
Minuit abnormal termination. No convergence?
.....!
Computing TS values for each source (40 total)
.....
```

Example Source Results

Source A: 2 free parameters

```
Integral: 0.000706819 +/- 0.017999
Index: -4.87644 +/- 1.0237
LowerLimit: 100
UpperLimit: 100000
Npred: 0.116799
ROI distance: 10.1342
TS value: -0.00617604
```

Galactic Diffuse: 2 free parameters

```
Prefactor: 0.984611 +/- 0.00884279
Index: 0.0296784 +/- 0.00345898
Scale: 100
Npred: 175955
```

Source B: 2 free parameters

```
Integral: 0.479765 +/- 0.156542
Index: -2.37132 +/- 0.12927
LowerLimit: 100
UpperLimit: 100000
Npred: 458.434
ROI distance: 3.26662
TS value: 25.9222
```

Crude Fit Quality Info

```
WARNING: Fit may be bad in range [100, 199.526] (MeV)
WARNING: Fit may be bad in range [281.838, 398.107] (MeV)
WARNING: Fit may be bad in range [2238.72, 4466.84] (MeV)
WARNING: Fit may be bad in range [25118.9, 35481.3] (MeV)
```

Source C: 0 free parameters

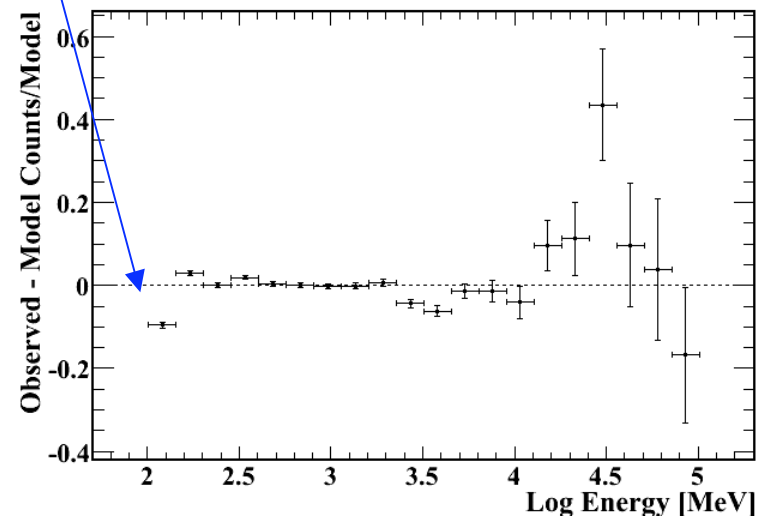
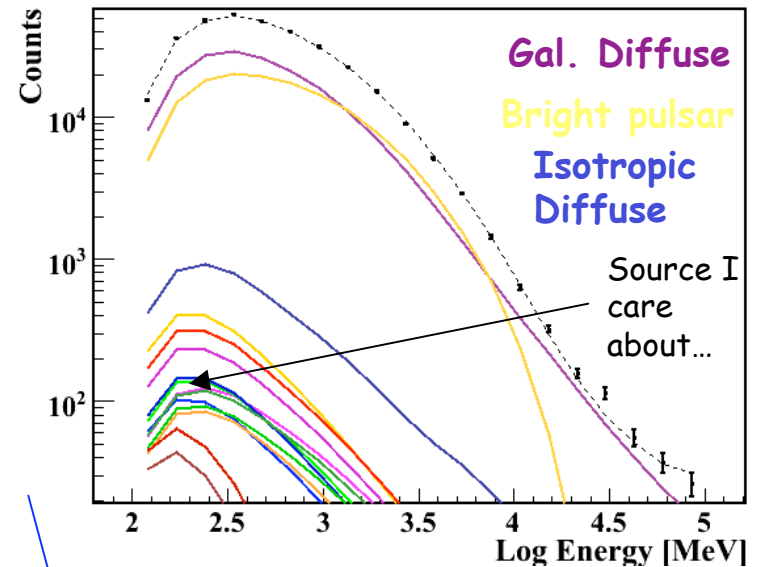
```
Integral: 6.28448
Index: -2.33404
LowerLimit: 100
UpperLimit: 100000
Npred: 315.177
```

```
Total number of observed counts: 325124
Total number of model events: 325093
```

```
-log(Likelihood): 2808753.988
```

Spectral Residuals

- **Unbinned analysis produces predicted counts and residuals as a function of energy.**
 - Example: a long integration near the Galactic plane and a bright pulsar
- **Discrepancy at low energy common**
 - **Likelihood uses true energy**
- **Discrepancies strongly tied to diffuse model for most analysis**
 - **Diffuse mediates cross talk between target source and nearby neighbors**
 - **Consider relative source strength**
 - **Test impact of model choices and selections on target source**



Likelihood - ROI selection

How big?

- **Big enough to constrain model components - source of interest, diffuse emission, nearby sources**
- **Small enough to avoid significant zenith cut loss to livetime**
 - **Practical advantage! less photons and less sources => less calculations for unbinned analysis**
 - **Analysis disadvantage! likelihood is an inclusive modeling strategy**
- **Recommendations**
 - **~10 deg for isolated point source ($E > 100$ MeV)**
 - **Larger regions (15-20 deg) benefit confused sources, aid in separating diffuse at low energy, improve error estimates**
- **Test it!**
 - **Are fit results reliable for different ROI radii?**
 - **What is the impact on GTIs?**

Likelihood Model - sources

What should be included?

- **All sources that contribute photons to the selected region**
 - **Bright source list sources within ~10 deg of the ROI boundary - accommodates tail of low energy PSF**
 - **Same goes for 1FGL catalog sources once available**
- **Galactic diffuse model**
- **Isotropic diffuse model**
 - **Important for all parts of the sky...provides a home for residual instrument effects (cosmic rays)**

This is a starting point. Adapt to find what works best for your region and source.

Likelihood Model - spectra

What spectral shape?

- **Power laws are simple and well defined**
 - **For faint sources, difficult to justify more parameters**
- **BUT lots of LAT sources are not simple power laws... some tips to help motivate other spectral forms**
 - **Bright pulsars?**
 - **Try simple exponentially cutoff power laws to improve fits for the pulsar itself *and for nearby sources***
 - **Visually inspect energy-dependent ROI selections**
 - **Do power-law fit parameters vary significantly for different minimum energy selections or fits in separate energy bins?**
- **Confirm: Most accurate and unbiased way to determine spectral parameters and errors is by testing the hypothesis using the likelihood analysis**

Likelihood - reality checks

Is anything missing?

- **Visual inspection of count maps and residuals**
- **Test Statistic maps (for unbinned analysis)**
 - *gttsmap* - Tests hypothesis of additional point source over a spatial grid
 - **Very Calculation Intensive**
 - try small regions (5 deg) and large grid spacing (0.5 deg)
 - Discrepancies may be additional source or component, or could be deficiencies in the diffuse model in some regions
 - **Warning: *gttsmap* is not ideal for localization, use *gtfindsrc***
- **Predicted and residual count maps (for binned analysis)**
 - **Profiles, radial density, energy dependence**

Likelihood - checking results

Is the result consistent for a different analysis?

- **Iteration**
 - Consistent results for the best fit parameters?
 - Tip: *gtlike sfile=best_fit_model.xml*
- **Data selection tests**
 - Minimum energy selection?
 - ROI selection? (Keep in mind this also effects good time selection in combination with zenith cut)
 - Consistency in distinct energy bins (catalog analysis)
 - Agreement using front or back events (requires use of appropriate IRFs, diffuse response, and isotropic model for each)
 - Time selections?
- **Fit and Minimization choices**
 - Impact of starting parameter values in the model?
 - Fit tolerance? (converging to true minimum?)
 - Effects of optimizer?

Likelihood: Upper Limits I

- For low Test Statistic (TS $\lesssim 9$), there is a python tool for determining the flux upper limit
- Profile method is used
 - Scan in flux to find value that gives $\Delta(\text{Log Likelihood}) = 1.35$ (2.71/2)

Upper Limits I

- **Set up the observation - same as for an unbinned analysis**
 - **For this example selecting energy 3000-10000 MeV**

```
>>> from UnbinnedAnalysis import *  
>>> from UpperLimits import *  
>>> infile='events_15_3000-10000_z105_gti.fits'  
>>> scfile='L090823195543E0D2F37E30_SC00.fits'  
>>> expmap='expmap.fits'  
>>> expcube='lrcube.fits'  
>>> obs=UnbinnedObs(infile,scfile,expMap=expmap,  
expCube=expcube,irfs='P6_V3_DIFFUSE')
```

Upper Limits II

- **Make sure the xml model matches your selection**
 - **Adjust to fit power law in your energy band - use PowerLaw2 spectral model and adjust upper and lower limits to match your selection**

Model xml file

```
<source name="3C 273" type="PointSource">  
<spectrum type="PowerLaw2">  
<parameter free="1" max="1000.0" min="0.001" name="Integral"  
scale="1e-09" value="10"/>  
<parameter free="0" max="-1.0" min="-5.0" name="Index"  
scale="1.0" value="-2.5"/>  
<parameter free="0" max="300000" min="20" name="LowerLimit"  
scale="1" value="3000"/>  
<parameter free="0" max="300000" min="20" name="UpperLimit"  
scale="1" value="10000"/>  
</spectrum>
```

Upper Limits III

- **Run the unbinned likelihood analysis**

```
>>> analysisEBand=UnbinnedAnalysis(obs,'model_3000-  
10000.xml',optimizer='MINUIT')  
  
>>> loglikeEBand=analysisEBand.fit()  
  
>>> print loglikeEBand  
8151.66172988  
  
>>> analysisEBand.Ts('3C 279')  
92.896105960517161  
  
>>> analysisEBand.Ts('3C 273')  
24.10538584394817  
  
>>> loglikeEBand.writeXml('fit_3000-10000.xml')
```


Upper Limits IV

- **Generate the upper limit for selected energy range**

```
>>> ul=UpperLimit(analysisEBand,'3C 273')
>>> results=ul.compute(emin=3000,emax=10000)
0 0.85992510336 -4.87233419335e-05 8.60017840996e-101 1.0238844754
0.0739214054393 1.02399489505e-092 1.18784384745 0.264851439945
1.1879719491e-093 1.35180321949 0.544267079746 1.35194900315e-094
1.51576259153 0.893093107955 1.51592605721e-095 1.70298920163 1.35927962311
1.70317285854e-09(1.7014539204951863e-09, 1.7012704489365027)
>>> print results
1.70e-09 ph/cm^2/s for emin=3000.0, emax=10000.0,delta(logLike)=1.35
>>> print results.parvalues
[0.85992510335991024, 1.0238844754025653, 1.1878438474452204,
1.3518032194878755, 1.5157625915305306, 1.7029892016282362]
>>> print results.dlogLike
[-4.8723341933509801e-05, 0.07392140543925052, 0.26485143994614191,
0.54426707974380406, 0.89309310795124475, 1.359279623112343]
```

Binned vs. Unbinned Likelihood

- **Unbinned:** Treats each photon independently (position, energy)
 - Best theoretical performance
 - More sensitive - important for faint sources
 - Best option for low statistics scenarios (e.g. flux light curves)
 - Drawbacks:
 - Not for use with spatially extended sources
 - Difficult to diagnose problems in individual source fit
- **Binned:** Treats the data in bins of position and energy. Minimal criteria - photons > bins
 - Less computationally intensive than unbinned
 - Handles templates for extended sources
 - Allows nice diagnostics of fit (source maps, spatial profiles, energy dependent comparisons of prediction and model)
 - Drawback: At highest energies, can run into low statistics even for long integrations

**Use of both allows consistency check
(if both can be reasonably used)**

Summary

- **Lots of ways to use the tools to evaluate spectral fitting and to validate results**
 - **Consistency is key**
 - **Analysis Cookbook provides basic starting points. Cicerone documentation provides deeper insight into the likelihood technique**
 - **The First Catalog paper provides detailed examples of spectral fitting with the science tools**