



Tutorial on Fermi-LAT data analysis the case of 3c454.3

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Material from P.Bruel, M.Razzano, S.Buson and R.Desiante

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Outline

- Overview of the Fermi Large AreaTelescope
 - How it works
 - LAT data
 - LAT performance
- Fermi Science Tools
 - General Introduction
- Maximum Likelihood Overview
 - Source modeling
- One study case:
 - 3c454.3: analysis tutorial



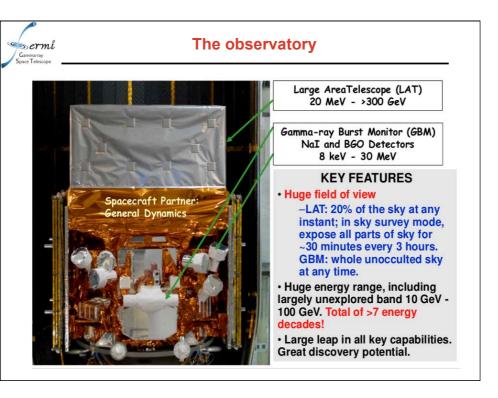
Organization

Today

- · Introduction to the LAT data analysis
- · LAT data introduction
- LAT data exploration
- · LAT data preparation
- Preparation of Sky models

Tomorrow

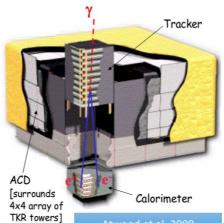
- Likelihood analysis of LAT data
- Hands-on on an Extragalactic source
 - · General LAT analysis with gtlike
 - · Light curve with aperture photometry
 - SED (possibly)





How the LAT works

- Precision Si-strip Tracker (TKR) Measure the photon direction; gamma ID.
- Hodoscopic Csl Calorimeter (CAL) Measure the photon energy; image the shower.
- Segmented Anticoincidence <u>Detector (ACD)</u> Reject background of charged cosmic rays; segmentation removes self-veto effects at high energy.
- <u>Electronics System</u> Includes flexible, robust hardware trigger and software filters.



Atwood et al. 2009

Systems work together to identify and measure the flux of cosmic gamma rays with energy 20 MeV - >300 GeV.

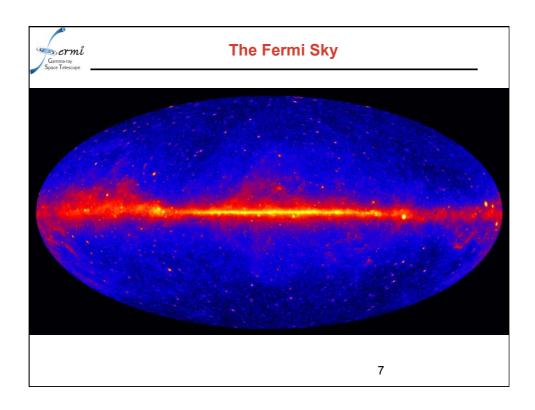
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Operating Mode

- Primary observing mode is Sky Survey
 - Full sky every 2 orbits (3 hours)
 - Uniform exposure, with each region viewed for ~30 minutes every 2 orbits
 - Best serves majority of science, facilitates multiwavelength observation planning
 - Exposure intervals commensurate with typical instrument integration times for sources
 - EGRET sensitivity reached in days
 - Pointed observations when appropriate (selected by peer review in later years) with automatic earth avoidance selectable. Target of Opportunity pointing.
 - · Autonomous repoints for onboard GRB detections in any mode.

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What do you need for the analysis

•Data ...of course!

LAT DATA ARE PUBLIC!!

- LAT detected events
- -Spacecraft related stuff
- -Extras (Backgrounds, catalog sources, timing..)

•Fermi Science Tools

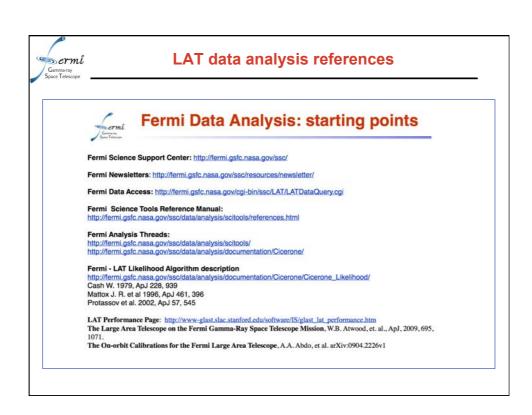
– http://fermi.gsfc.nasa.gov/ssc/data/analysis/software/

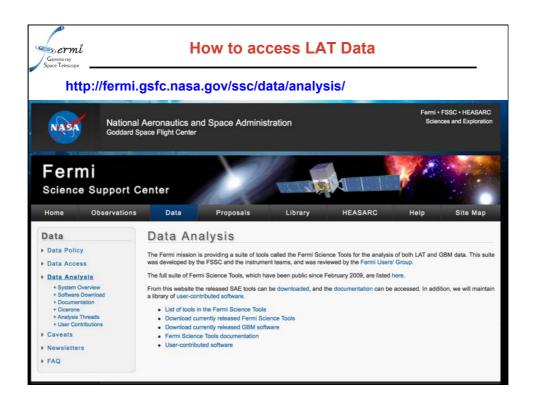
Other ancillary tools

- ftools, HEASOFT, DS9 etc..
- http://heasarc.gsfc.nasa.gov/docs/software.html

·Lots and lots of scripts!

- Fermi tools are already scriptable
- You can also use your favourite scripting language... but ..
- Science Tools can be imported as Python modules!

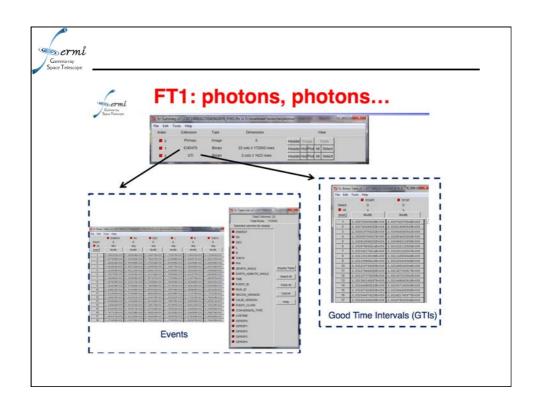


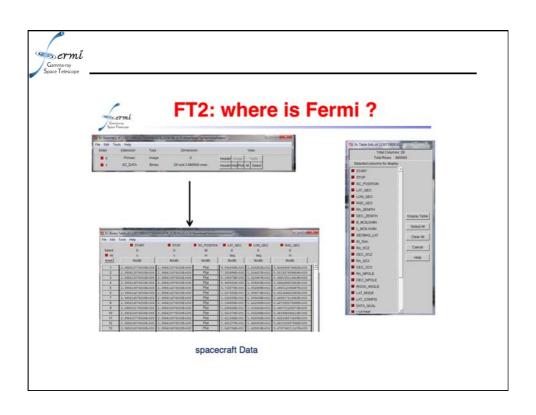


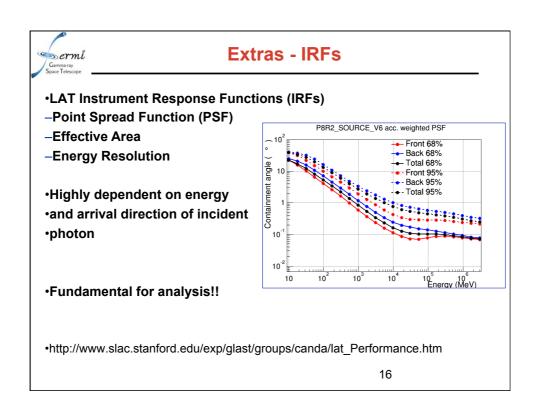


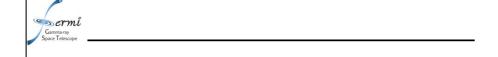
LAT Data

- •LAT data products can be downloaded by the FSSC website
- **-LAT Data server** http://fermi.gsfc.nasa.gov/cgi-bin/ssc/LAT/LATDataQuery.cgi
- -Archive of weekly files
- -ftp://legacy.gsfc.nasa.gov/fermi/data/
- •Two main data products (stored in FITS format)
- -Events file (FT1)
- -i. e. "what the LAT sees"
 - (photons, their energy, coordinates, time, event classes etc..)
- -Spacecraft files (FT2)
- -i. e. "where the LAT is"
 - (position, angles..)

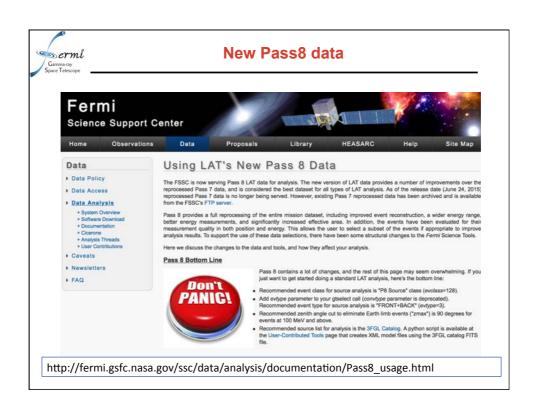








Pass8 (?)...





Pass 8 introduction

What does Pass mean?

- · Each pass corresponds to a version of the Fermi LAT data
- · It implies a whole package:
 - · Intrument simulation
 - · Reconstruction code
 - · Event selection
 - Instrument Response Functions (IRFs)
 - · Systematic uncertainties
 - Isotropic template (which includes the cosmic-ray residual background)
 - And sometimes more (Galactic diffuse model, Earth limb template, Sun+Moon template)
- It's only when we have validated the whole package that we can release it to the public.



Pass8 introduction

From Pass 6 to Pass 8

- Pass 6 (launch time)
 - · Pass 6 reconstruction
 - Pass 6 selection
 - · Based on pre-launch instrument simulation
 - First data revealed the issue of out-of-time pile-up (aka ghosts)
 - New: instrument simulation with ghosts -> correct IRFs
- Pass 7
 - Pass 6 reconstruction
 - New: Pass 7 selection optimized with simulations with ghosts
- Pass 8
 - New: improved instrument simulation
 - New: Pass 8 reconstruction, as ghost-proof as possible
 - · New: Pass 8 selection



Pass8 Introduction

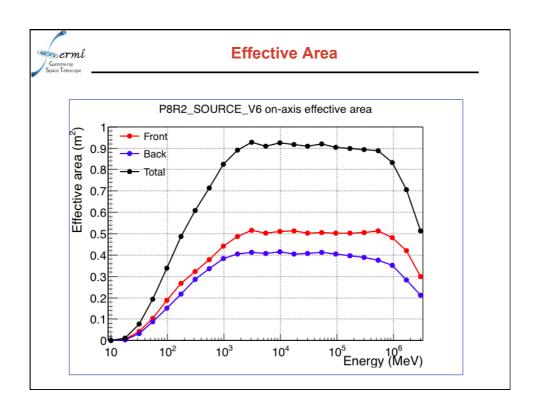
Pass 8 improvements

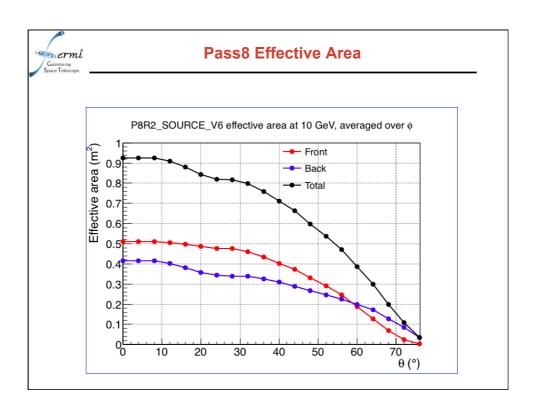
- · Ghost handling
- Tracker: ignoring ghost hits
 ACD: partial deghosting
 Calorimeter: clustering and cluster classification
- Improved direction measurement
 - · Tree-based track finder
- · Improved energy measurement
- Extension of the energy range: from ~10 MeV to ~3 TeV
- · Improved track/ACD matching information
 - · Using the uncertainty of the tracker direction
- Improved event selection
 Using the ROOT TMVA package (tmva.sourceforge.net)
- · Additional sub-classes of events

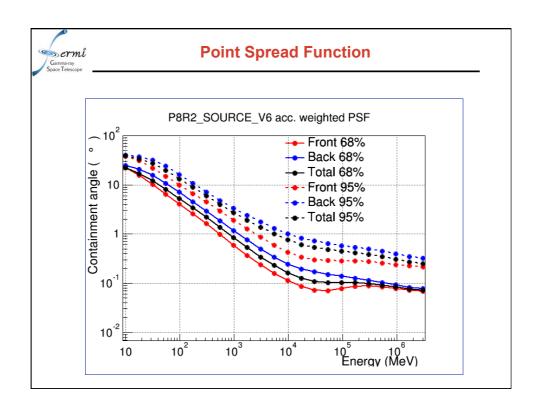


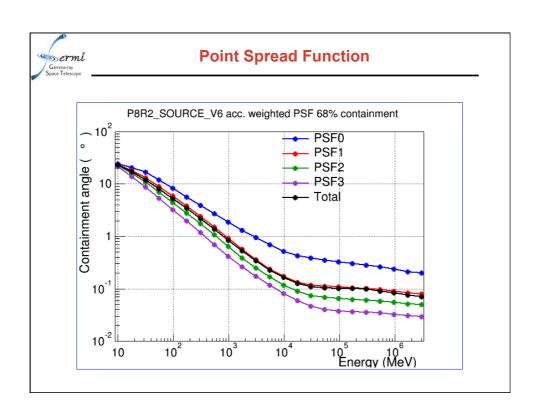
Pass8 performance

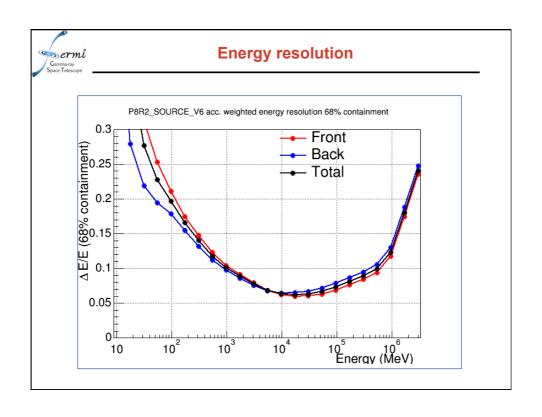
http://www.slac.stanford.edu/exp/glast/groups/canda/lat_Performance.htm

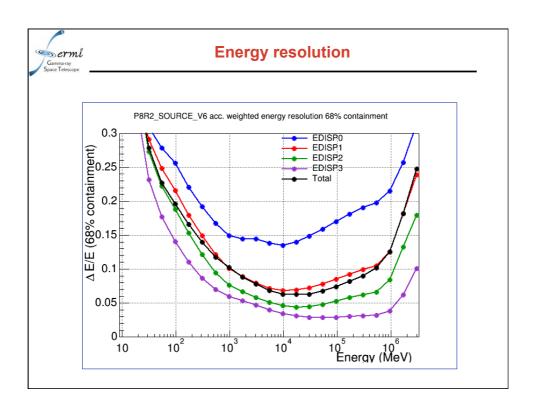








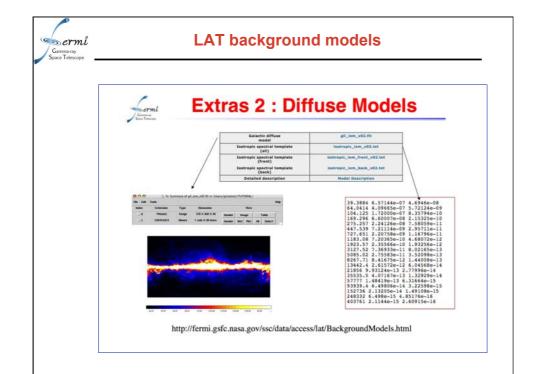


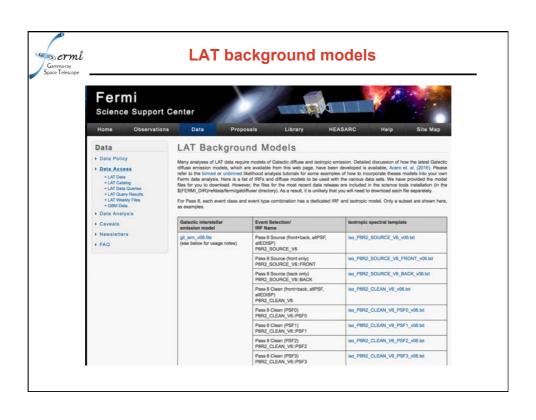


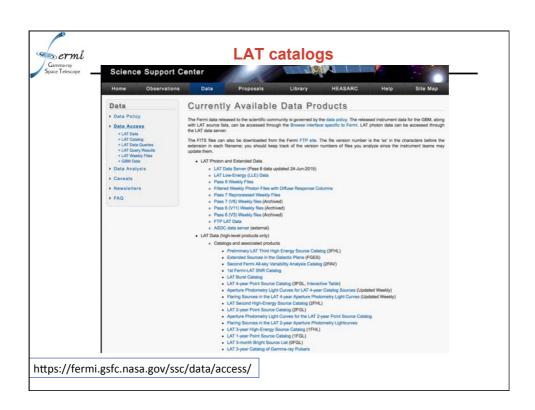


Extras

- Diffuse models (.txt & FITS files)
- -To correctly take into accounts the galactic and extagalactic backgrounds
- -http://fermi.gsfc.nasa.gov/ssc/data/access/lat/BackgroundModels.html
- Source Catalogs (3FGL, 3FHL, PSRs, GRB, SNR ...)
- · Region of Interest model definition (stored in XML files)
 - More on this later ...
- Choose the proper combination of
 - event classes based on the probability of being a photon and background level
 - TRANSIENT (for very short events)
 - SOURCE (suggested for source analysis)
 - CLEAN and ULTRACLEAN (lowest particle contamination for diffuse sources analysis)
 - Event types based on conversion point or PSF or Energy reconstruction









Event types

P8R2 Event Type Name	Event Type Partition	Event Type Value (evtype)	
FRONT	Conversion Type	1	
BACK	Conversion Type	2	
PSF0	PSF	4	
PSF1	PSF	8	
PSF2	PSF	16	
PSF3	PSF	32	
EDISP0	EDISP	64	
EDISP1	EDISP	128	
EDISP2	EDISP	256	
EDISP3	EDISP	512	



Event classes

P8R2 IRF name	Event Class (evclass)	Class Hierarchy	Photon File	Extended File	
P8R2_ULTRACLEANVETO_V6	1024	Standard	х	х	
P8R2_ULTRACLEAN_V6	512	Standard X		х	
P8R2_CLEAN_V6	256	Standard	х	х	
P8R2_SOURCE_V6	128	Standard	х	х	
P8R2_TRANSIENT010_V6	64	Standard		х	
P8R2_TRANSIENT020_V6	16	Standard		X	
P8R2_TRANSIENT010E_V6	64	Extended		х	
P8R2_TRANSIENT020E_V6	8	Extended		х	
P8R2 TRANSIENT015S V6	65536	No-ACD		х	



Recommendations

Event Selection Recommendations (P8R2)

Analysis Type	Minimum Energy (emin)	Maximum Energy (emax)	Max Zenith Angle (zmax)	Event Class (evclass)	IRF Name	
Galactic Point Source Analysis	100 (MeV)	500000 (MeV)	90 (degrees)	128	P8R2_SOURCE_V6	
Off-plane Point Source Analysis	100 (MeV)	500000 (MeV)	90 (degrees)	128	P8R2_SOURCE_V6	
Burst and Transient Analysis (<200s)	100 (MeV)	500000 (MeV)	100 (degrees)	16	P8R2_TRANSIENT020_V6	
Galactic Diffuse Analysis	100 (MeV)	500000 (MeV)	90 (degrees)	128	P8R2_SOURCE_V6	
Extra-Galactic Diffuse Analysis	100 (MeV)	500000 (MeV)	90 (degrees)	1024	P8R2_ULTRACLEANVETO_V6	
Impulsive Solar Flare Analysis	100 (MeV)	500000 (MeV)	100 (degrees)	65536	P8R2_TRANSIENT015S_V6	



Caveats

Caveats About Analyzing LAT Pass 8 Data

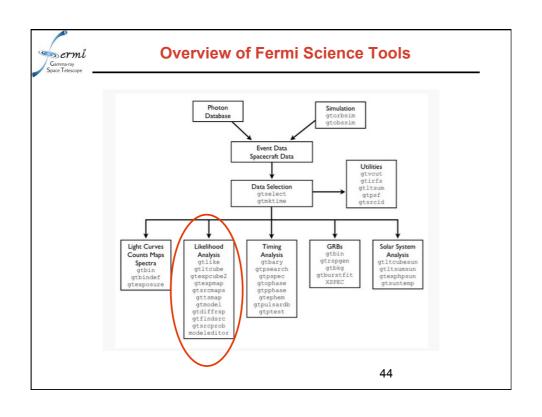
These caveats are relevant for the P8R2 version of the Pass 8 photon dataset. They are an updated version of previous sets of caveats for analysis of Pass 7 reprocessed (P7REP), Pass7 (P7_V6) and Pass 6 (P6_V3 and P6_V11) event selections and Instrument Response Functions (IRFs).

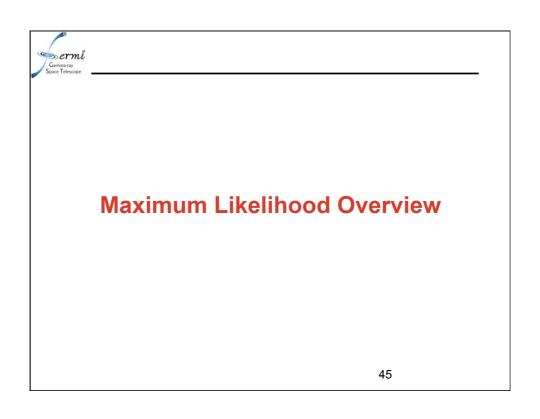
The LAT team is still working to validate all aspects of Pass 8 data and analysis. As a consequence it is expected that, in the coming year, the range of application of Pass 8 data will be increased, the tools and files will be improved and the systematic uncertainties will be decreased. These caveats will be modified accordingly.

The P8R2_V6 IRFs are defined between 5.62 MeV and 3.16 TeV but that does not mean that they have been fully validated over this whole energy range. Following the Pass 7 validation effort, the LAT team has started performing studies in order to check the consistency and precision of the instrument simulation and the instrument response representation provided by the IRFs. These studies are based on the analysis of Vela (E < 10 GeV), bright AGN (3 GeV < E < 100 GeV) and the Earth limb (E > 10 GeV). They include:

- comparing the distributions of reconstructed quantities between data and the simulation of the instrument for a given selection (e.g. SOURCE class)
- comparing the fraction of events of a loose selection (e.g. TRANSIENT020 class) that are accepted in a more stringent selection (e.g. SOURCE class) in data and as predicted by the instrument simulation or the IRFs
- comparing the fraction of events in each event type for a given selection in data and as predicted by the instrument simulation or the IRFs
- comparing the radial profiles of a point source in data and as predicted by the IRFs

http://fermi.gsfc.nasa.gov/ssc/data/analysis/LAT caveats.html







Maximum likelihood technique

Given a set of observed data:

- → Produce a model that accurately describes the data, including parameters that we wish to estimate
- → derive the probability (density) for the data given the model (PDF)
- → treat this as a function of the model parameters (likelihood function)
- → maximize the likelihood with respect to the parameters ML estimation.

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Maximum likelihood ingredients

Data: $X = \{x_i\} = \{x_1, x_2, ..., x_N\}$

Model parameters: $\Theta = \{\theta_j\} = \{\theta_1, \theta_2, ..., \theta_M\}$

Likelihood: $\mathcal{L}(\Theta|X) = P(X|\Theta)$

Conditional probability rule

for independent events: P(A,B) = P(A)P(B|A) = P(A)P(B)

For independent data:

$$P(X|\Theta) = P(\{x_i\}|\Theta) = P(x_1|\Theta)P(x_2,..,x_N|\Theta) = \cdots$$
$$= P(x_1|\Theta)P(x_2|\Theta)\cdots P(x_N|\Theta) = \prod_i P(x_i|\Theta)$$

 $\mathcal{L}(\Theta|X) = \prod_{i} P(x_{i}|\Theta)$ \mathcal{L} is the product of the

 $\ensuremath{\mathcal{L}}$ is the product of the probability of observing the detected counts in each bin.



Maximum likelihood estimation

Parameters can be estimated by maximizing likelihood.

→ Easier to work with log-likelihood:

$$\ln \mathcal{L}(\Theta) = \ln \mathcal{L}(\Theta|X) = \sum_{i} \ln P(x_{i}|\Theta)$$

Estimates of $\{\hat{ heta}_k\}$ from solving simultaneous

equations:

$$\left. rac{\partial \ln \mathcal{L}}{\partial heta_j}
ight|_{\{\hat{ heta}_k\}} = 0$$

For one parameter, if we have: $\mathcal{L}(\theta) \sim e^{-\frac{Q}{2\sigma_{\theta}^2}}$

then: $\left. \frac{\partial^2 \ln \mathcal{L}}{\partial \theta^2} \right|_{\hat{\theta}} = -\frac{1}{\sigma_{\theta}^2}$

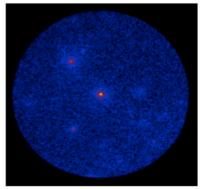
so 2nd derivative is related to "errors"

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The Challenge

- Gamma-ray data is a list of counts (photons) reconstructed in the detector.
- Qualitative exploration of the data suggests the presence of sources (spatial clustering).
- Quantitative analysis requires evaluating the significance of a 'model' of our region.



Count Map



The Procedure

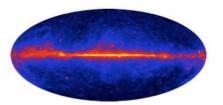
- Basically the initial 'model' is a guess of the various parameters of the sources in our region:
 - location
 - spectral shape
 - flux
 - etc.
- The guess can be easier if you have a starting point like the 2FGL (3FGL).
- We quantify (using the Likelihood Method) the statistical significance of the model and vary the parameters to determine the most likely parameter values.

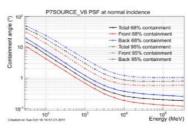
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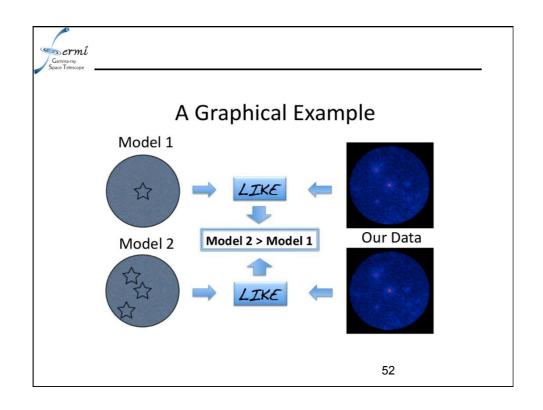


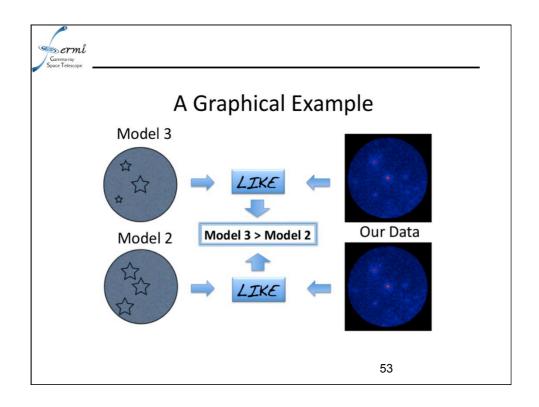
Why Model Fitting?

- We use the Likelihood method because the LAT data are limited by statistics, a bright diffuse background and a broad PSF.
- The model construction defines the questions we want to ask which means:
 - It will not answer a question you are not asking (ie. unknown parameters).
- The Likelihood will not tell you
 if a fit is 'good'. If the model
 does not represent your data
 well, the results will also not
 represent reality well.











Likelihood Analysis

- The likelihood ∠ is the probability of obtaining your data given an input model.
- In our case, the input model is the distribution of gammaray sources on the sky and includes their intensity and spectra.
- One will maximize ∠ to get the best match of the model to the data. Given a set of data, one can bin them in multidimensional (energy, sky pixels, ...) bins.
- \mathcal{L} is the product of the probabilities of observing the detected counts in each bin.

$$\mathcal{L} = \prod p_k$$

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Write L as a function of the source model

- The source model is folded with the IRFs in order to obtain the predicted number of counts
- The IRFs can be decomposed into three funtions: Effective Area (proj area of the detector * efficency), Energy Dispesion, Point Spread Function
- Small number of counts in each bin --> Poisson distribution
- Bin size infinitesimally small
- Assume only steady source for standard analysis

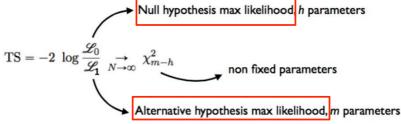
The function to maximize is:

$$\log \mathcal{L} = \sum_{j} \log M(E'_{j}, \hat{p}'_{j}, t_{j}) - N_{\text{pred}}$$

where the sum is performed over photons in the ROI. The predicted number of counts is $N_{\rm pred} = \int_{\rm ROI} dE' d\hat{p}' \, dt M(E',\hat{p}',t)$



Test Statistic



- In the limit of a large number of counts, Wilk's Theorem states that the TS for the null hypothesis is asymptotically distributed as $\mathcal{X}_n^{\mathcal{Z}}$ where n is the number of parameters characterizing the additional source.
- As a basic rule of thumb, the square root of the TS is approximately equal to the detection significance for a given source.

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Summarizing

- · Observed a photon from a location, at a time, with an energy.
- · Assume a model:
- $S(E,\hat{p},t) = \sum_{i} s_{i}(E,t)\delta(\hat{p}-\hat{p}_{i}) + S_{G}(E,\hat{p}) + S_{\text{eg}}(E,\hat{p}) + \sum_{l} S_{l}(E,\hat{p},t),$ $\uparrow \qquad \qquad \uparrow$ Prior Course
 Gallant & 1G diffuse Source
- Calculate the probability of that photon being detected assuming our model: $M(E',\hat{p}',t) = \int_{\text{cp}} dE d\hat{p} R(E',\hat{p}',t;E,\hat{p}) S(E,\hat{p},t)$
- Calculate the total number of predicted counts assuming our model.

$$N_{\mathrm{pred}} = \int_{\mathrm{ROI}} dE' d\hat{p}' \, dt M(E', \hat{p}', t)$$

· Adjust the model until this is maximized:

$$\log \mathcal{L} = \sum_{j} \log M(E'_{j}, \hat{p}'_{j}, t_{j}) - N_{\text{pred}}$$

· Calculate the TS:

$$\mathrm{TS} = -2 \; \log \frac{\mathscr{L}_0}{\mathscr{L}_1} \; \mathop{\to}_{N \to \infty} \; \chi^2_{m-h}$$



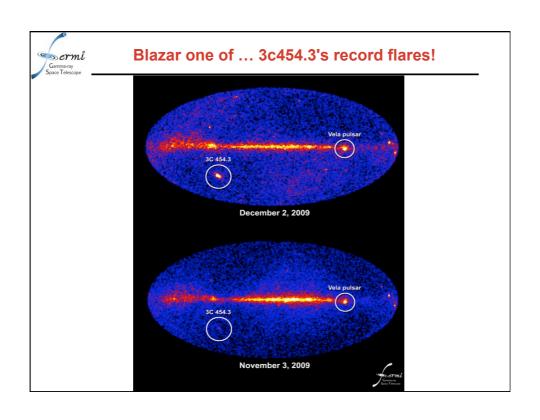
Keep in mind that ..

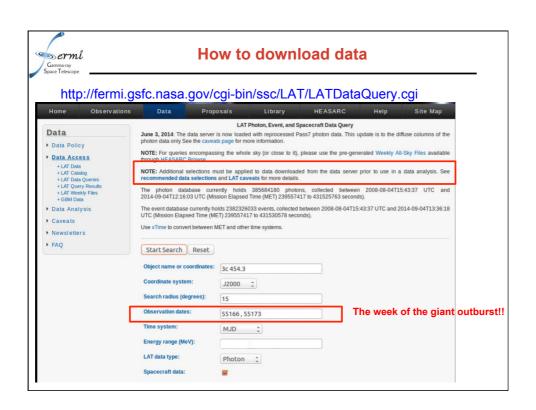
- Many variables may be calculated BEFORE selecting the models
- · Very important to have a reliable model
- Absolute value of likelihood meaningless!
 - Likelihood function has no meaning itself, e.g., it is not a probability. Its usefulness comes from theorems such as the LRT.
- Comparison between model w/ and w/o source to reject null hypothesis = no source (TS large → reject null hypothesis)

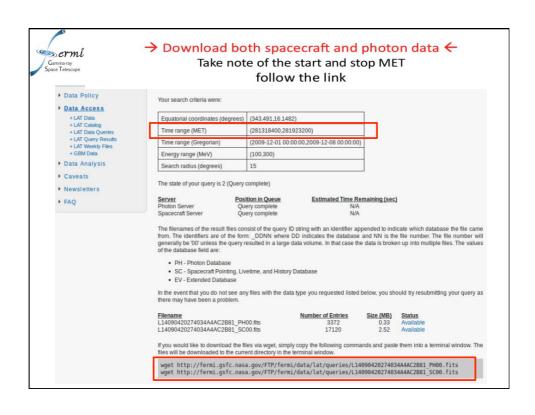
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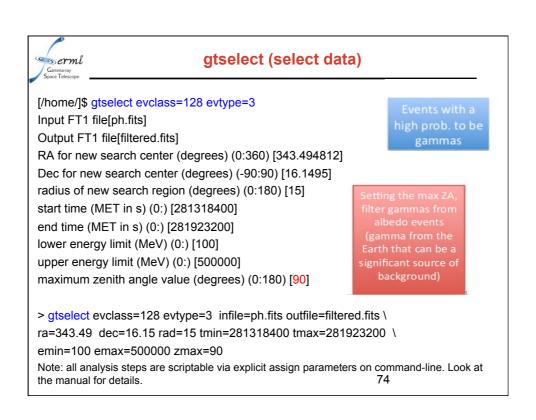


Analysis Tutorial











gtmktime (cut the bad time intervals)

Filter out events collected while passing in SAA and other low-quality events

[/home/]\$ gtmktime

Spacecraft data file[sc.fits]

Filter expression[(DATA_QUAL>0)&&(LAT_CONFIG==1)]

Apply ROI-based zenith angle cut[no]

Event data file[filtered.fits]

Output event file name[filtered_gti.fits]

Use ZA to filter only proper GTIs

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gtbin (Counts Map)

[/home]\$ gtbin

Type of output file (CCUBE|CMAP|LC|PHA1|PHA2|HEALPIX) [CMAP]

Event data file name[filtered_gti.fits]

Output file name[CMAP.fits]

Spacecraft data file name[sc.fits]

Size of the X axis in pixels[120]

Size of the Y axis in pixels[120]

Image scale (in degrees/pixel)[0.25]

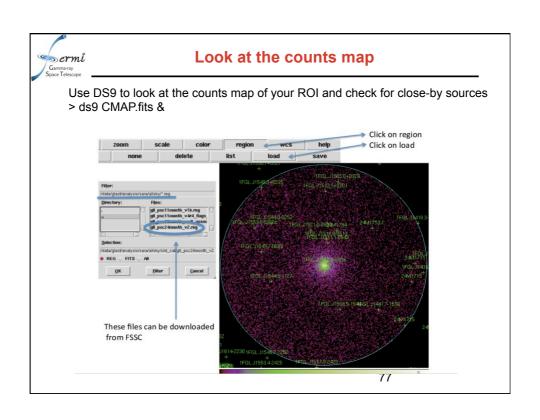
Coordinate system (CEL - celestial, GAL -galactic) (CEL|GAL) [CEL]

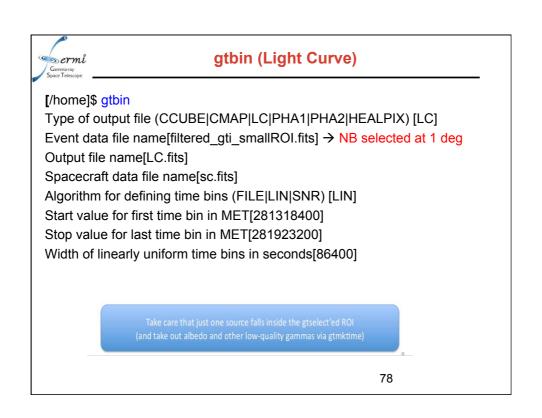
First coordinate of image center in degrees (RA or galactic I) [343.494812]

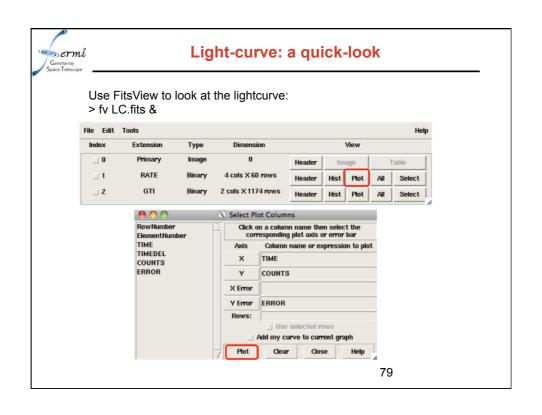
Second coordinate of image center in degrees (DEC or galactic b) [16.1495]

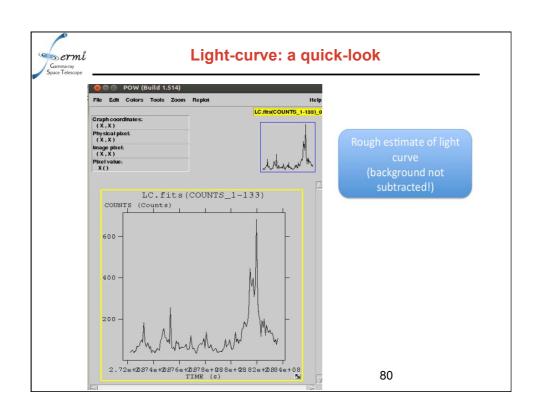
Rotation angle of image axis, in degrees[0]

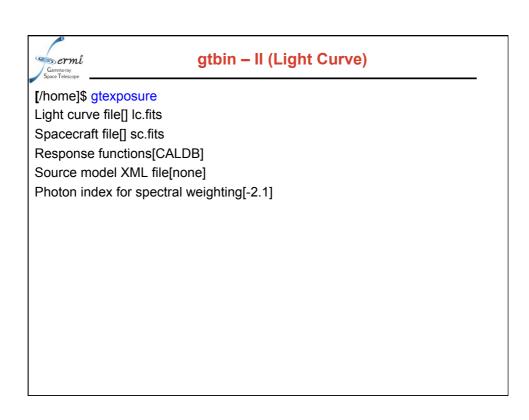
Projection method e.g. AIT|ARC|CAR|GLS|MER|NCP|SIN|STG|TAN: [AIT] 76

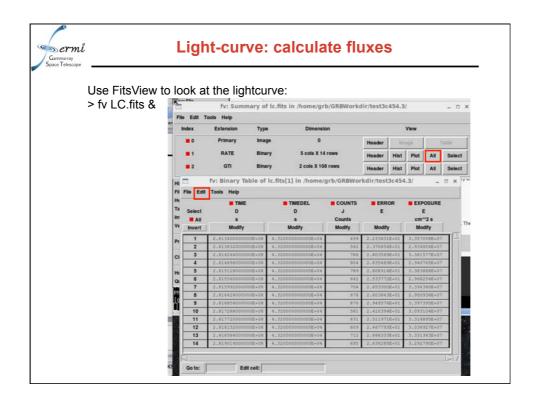


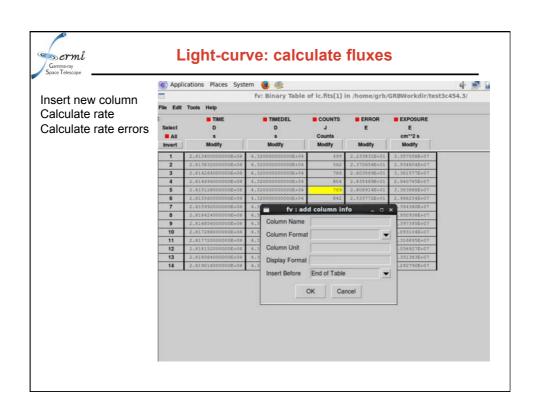


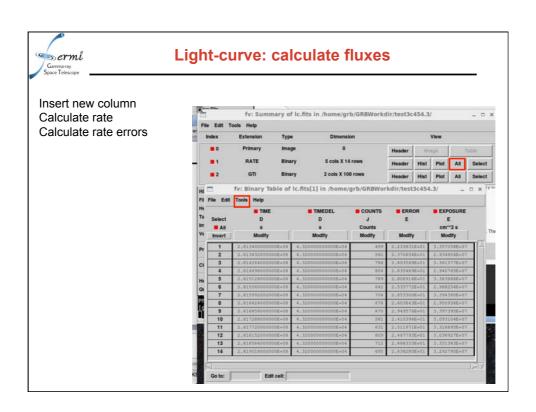


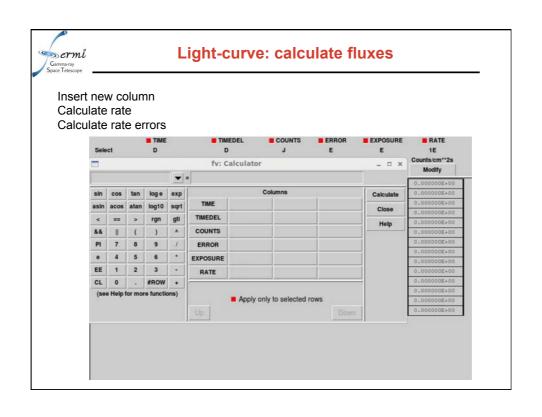


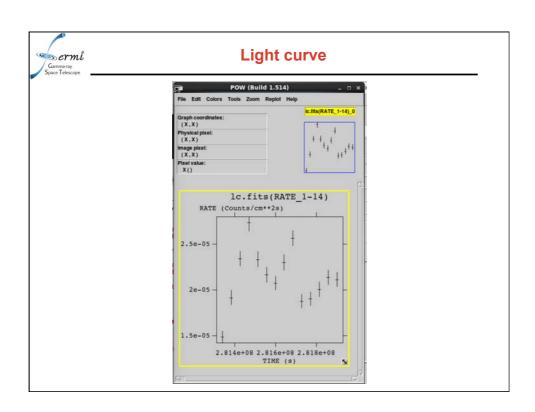














Perform the fit: the likelihood approach

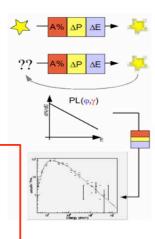
In high energy gamma rays it is never possible to really isolate a source because of limited statistics and strong and structured background.

Therefore statistical techniques have to be applied.

The most used method is the <u>likelihood</u> analysis based on the <u>Poisson statistics</u>.

The method requires to assume a model for the signal detected by the telescope.

- · Assume a model
- Model convolved with Instrument responce Function (IRF)
- Maximizing likelihood find the best set of parameters that reproduce the observed spectrum



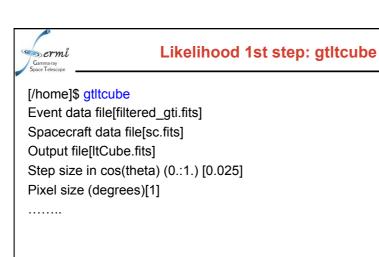
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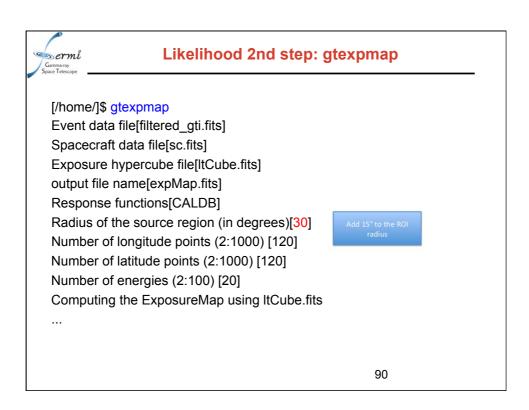
Perform the fit: the likelihood approach

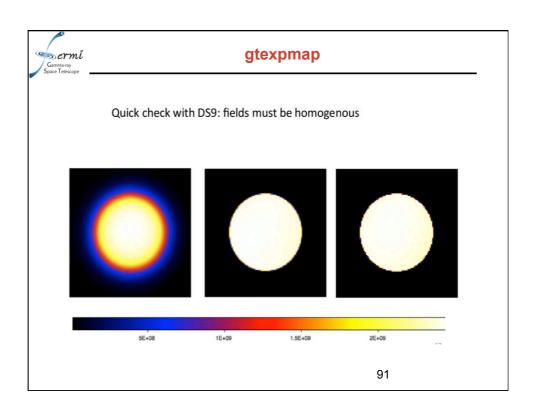
- Absolute value of likelihood meaningless!
- Comparison between model w/ and w/o source to reject H₀ = no source
- Many variables may be calculated BEFORE selecting the models

IRFs depend on inclination angle: Livetime Cube: seconds in $\Delta\Omega$ with a given z, the time that the LAT observed a given position on the sky at a given inclination angle Exposure Map: integration of the effective area over the FoV weighted by the livetime over a position-energy grid, $N_{\text{model}} = \int \Phi_{\text{model}}(\Omega, E(t)) \times A_{\text{LAT}}(\Omega, E)$



The "livetime cube" must be re-calculated when a new time interval or a new ZA is selected







Likelihood 3rd step: the XML model

5. Create a source model XML file

The gtilike tool reads the source model from an XML file. The model file contains your best guess at the locations and spectral forms for the sources in your data. A source model can be created using the model editor tool, by using the user contributed tool make3FGLxml.py (available at the user-contributed tools page), or by editing the file directly within a text editor.

Here we cannot use the same source model that was used to analyze six months of data in the Unbinned Likelihood tutorial, as the 2-year data set contains many more significant sources and will not converge. Instead, we will use the 3FGL catalog to define our source model by running make3FGLxml.py. To run the script, you will need to download the current LAT catalog file and place it in your working directory:

prompt> make3FGLxml.py gll_psc_v16.fit 3C279_binned_gti.fits -o 3C279_input_model.xml -G \$FERMI_DIR/refdata/fermi/galdiffuse/gll_iem_v06.fits -g gll_iem_v06 -I \$FERMI_DIR/refdata/fermi/galdiffuse/iso_P8R2_SOURCE_V6_v06.txt -i iso_P8R2_SOURCE_V6_v06 -s 120 -p TRUE -v TRUE This is make3FGLxml version 01r0. The default diffuse model files and names are for pass 8 and assume you have v10r00p05 of the Fermi Science Tools or higher. Creating file and adding sources from 3FGL Added 312 point sources, note that any extended sources in R0I were modeled as point sources because psForce option was set to True prompt>

Note that we are using a high level of significance so that we only fit the brightest sources and we have forced the extended sources to be modeled as point sources. This only affects the lobes of Centarus A which are just outside the FOV.

It is also necessary to specify the entire path to location of the diffuse model on your system. The resulting XML model contains 312 sources. Clearly, the simple 4-source model we used for the 6-month Unbinned Likelihood analysis would have been too simplistic.

This XML file uses the spectral model from the 3FGL catalog analysis for each source. (The catalog file is available at the LAT 4-yr Catalog page.) However, that analysis used a subset of the available spectral models. A dedicated analysis of the region may indicate a different spectral model is preferred. For more details on the options available for your XML models, see:

- . Examples of XML Model Definitions for Likelihood



Likelihood 3rd step

./make3FGLxml.py gll_psc_v16.fit filtered_gti.fits -o 3c454.3.xml -G / home/grb/software/GlastExt/diffuseModels/v2r0/gll_iem_v06.fits -g gll_iem_v06 -I /home/grb/software/GlastExt/diffuseModels/v2r0/ iso_P8R2_SOURCE_V6_v06.txt -i iso_P8R2_SOURCE_V6_v06 -s 120 -p TRUE -v TRUE



Likelihood 3rd step: the XML model

Backgrounds

- <!-- Diffuse Sources -->
- <source name="galactic_background" type="DiffuseSource">
- <spectrum type="PowerLaw">
- <parameter free="1" max="10" min="0" name="Prefactor" scale="1" value="1"/>
- <parameter free="0" max="1" min="-1" name="Index" scale="1.0" value="0"/>
- <parameter free="0" max="2e2" min="5e1" name="Scale" scale="1.0" value="1e2"/>
- </spectrum>
- <spatialModel file="gll_iem_v06.fits" type="MapCubeFunction">
- <parameter free="0" max="1e3" min="1e-3" name="Normalization" scale="1.0" value="1.0"/>
- </spatialModel>
- </source>
- <source name="extragalactic_background" type="DiffuseSource">
- <spectrum file="iso_P8R2_SOURCE_V6_v06.txt" type="FileFunction">
- <parameter free="1" max="10" min="1e-2" name="Normalization" scale="1" value="1"/>

- </spectrum>
- <spatialModel type="ConstantValue">
- <parameter free="0" max="10.0" min="0.0" name="Value" scale="1.0" value="1.0"/>
- </spatialModel>
- </source>



XML model

Typical source entry for an assumed powerlaw

```
{
source name="....." type="PointSource">...
</source> Your sources here
```

<source name="3c454.3" type="PointSource">

<spectrum type="PowerLaw2">

<!-- Source is in ROI center -->

<parameter error="0.00" free="1" max="1000" min="1e-06" name="Integral" scale="1e-04"
value="1.000"/>

<parameter free="0" max="3e6" min="20" name="UpperLimit" scale="1" value="300000."/>
</spectrum>

<spatialModel type="SkyDirFunction">

<parameter free="0" max="360.0" min="-360.0" name="RA" scale="1.0" value="343.494812"/>
<parameter free="0" max="90" min="-90" name="DEC" scale="1.0" value="16.149500"/>

</spatialModel>

</source>

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XML model

Test different models... power law * HE exponential cut-off

<source name="3c454.3" type="PointSource">

<spectrum type="PLSuperExpCutoff">

<parameter free="1" max="1000" min="1e-05" name="Prefactor" scale="1e-07"
value="1"/>

<parameter free="1" max="0" min="-5" name="Index1" scale="1" value="-1.7"/>

<parameter free="0" max="1000" min="50" name="Scale" scale="1" value="200"/>

<parameter free="1" max="30000" min="500" name="Cutoff" scale="1" value="3000"/>

<parameter free="0" max="5" min="0" name="Index2" scale="1" value="1"/>

</spectrum>

•Look here for source model definition and XML model definitions:

http://fermi.gsfc.nasa.gov/ssc/data/analysis/scitools/xml_model_defs.html http://fermi.gsfc.nasa.gov/ssc/data/analysis/scitools/source_models.html

•Useful python script to load 2FGL sources that belongs to your ROI in your XML file model (make3FGLxml.py)

http://fermi.gsfc.nasa.gov/ssc/data/analysis/user/



Diffuse response

[/home/]\$gtdiffrspEvent data file[]filtered_gti.fits Spacecraft data file[] sc.fits

Source model file[] 3c454.3.xml

Response functions to use[] CALDB

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Finally... gtlike performig the actual fit

[/home/]\$ gtlike plot=yes

Statistic to use (BINNED|UNBINNED) [UNBINNED]

Spacecraft file[sc.fits]

Event file[filtered_gti.fits]

Unbinned exposure map[expMap.fits]

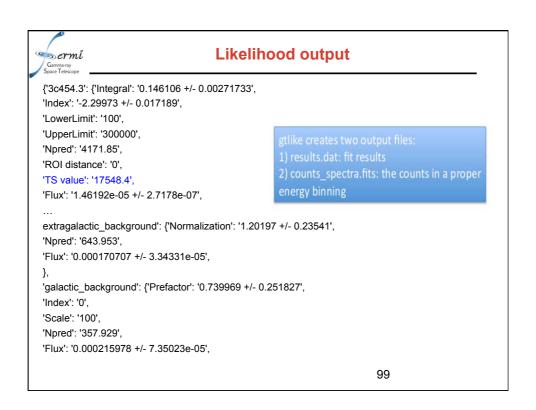
Exposure hypercube file[ItCube.fits]

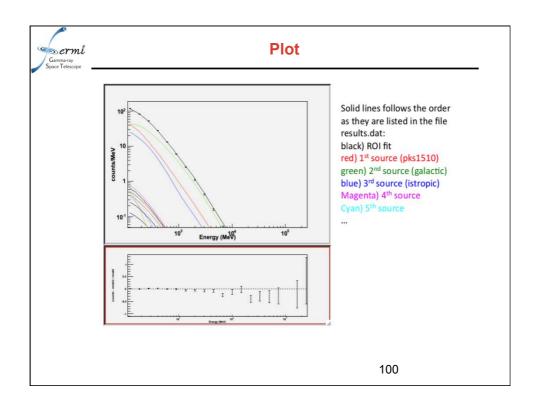
Source model file[../xml_models/_3c454.3_model_ROI15.xml]

Response functions to use[P7REP_SOURCE_V15]

Optimizer (DRMNFB|NEWMINUIT|MINUIT|DRMNGB|LBFGS) [NEWMINUIT]

Typically use DRMNGB/DRMNFB to find a rough estimate of the likelihood maxima and refine later on with MINUIT (or NEWMINUIT)







Comparison of different models

Powerlaw * HE exp cut-off

{'3c454.3': {'Prefactor': '0.39194 +/- 0.00793161',

'Index1': '-2.12802 +/- 0.03056',

'Cutoff': '5495.55 +/- 934.857 (MeV)

'Npred': '4157.04', 'ROI distance': '0', 'TS value': '17604.2',

'Flux': '1.41693e-05 +/- 2.72878e-07'

(ph cm-2 s-1)

----> Comparing TS values for different models!

For this source, in this time interval, the model with the HE exponential cutoff is favoured with respect to the

Simple Powerlaw

You can repeat analyses by yourself also following instructive and complete Tutorials on the FSSC website:

- Standard Likelihood: http://fermi.gsfc.nasa.gov/ssc/data/analysis/scitools/likelihood_tutorial.html
- PyLike: http://fermi.gsfc.nasa.gov/ssc/data/analysis/scitools/python_tutorial.html

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SED modeling

Generate Spectral Points

To generate spectral points to plot on top of the butterfly that we just produced, you need to go back to the data selection part and use gisselect (filter in python) to divide up your data set in energy bins and run the likelihood fit on each of these individual bins. Luckliy, there's a script that does this for us that we'll employ to get a final result. This script ogenerates the butterfly plot we just produced so you won't have to redo that again and it also does a lot of checks on the data to make sure that everything is going ok. If you've made it this far, you might be a little curious as to why we didn't jump right into using this tool but now you're in a position to play with the python tools and make them do what you want them to. The script is also much more careful in handling units and saving everything to files than we have been in this interactive session.

So download the likeSED.py user contributed tool (it's in the SED_scripts package) and load it up into python. You can find information on the usage of this tool on the same page where you downloaded it. It can be used to generate an SED plot for both binned and unbinned analyses but we're only going to work on a binned analysis here.

>>> from likeSED import *
This is likeSED version 12.1, modified to handle Pass 7 selections.

Now you need to create a likeInput object that takes our unbinnedAnalysis object, the source name and the number of bins we want as arguments. We're going to make 9 bins here like in the paper and we're also going to make custom bins and bin centers. You can have the module chose bins and bin centers for you (via the getECent function) but we're going to do it this way so we can have the double-wide bin at the end. We're also going to use the 'fit2' file (the result of our fit on the full dataset with the NewMinuit optimizer) for the fits of the individual bins but we need to edit it first to make everything fixed except for the integral value of PG 1553-113. Go ahead and do that now. We're basically assuming that the spectral index is the same for all bins and that the contributions of the other sources within the ROI are not changing with energy.

>>> inputs = likeInput(like,'3FGL J1555.7+1111',nbins=9)
>>> inputs.plotBins()

