From detection to high-level reconstruction

A statistical view on gamma-ray sources

2024.10.15, Č Astronomy Data School

Jonathan Biteau



Concepts: DL3-DL4-DL5, forward-folding, binomial / Poisson / Gauss, PMF / PDF / CDF / SF

II. Statistical inference

Concepts: Bayesian & frequentist, marginalizing & profiling, parameter estimation, (saturated) deviance

III. Model selection

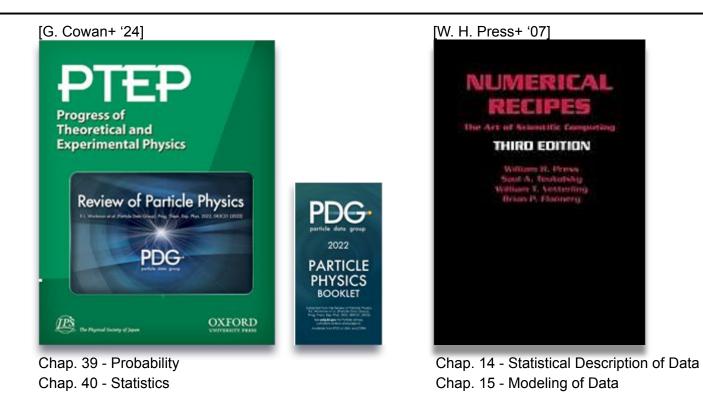
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IV. Dealing with uncertainties

Concepts: covariance, decorrelation energy, bowtie, systematic uncertainties, bracketing / marginalizing

V. Outro



Lesson adapted from my Master 2 notes, available at this link

Elements of statistics

Master 2 - Nuclei, Particles, Astroparticles and Cosmology NPAC 2024–2025, Jonathan Biteau

This document aims to introduce and illustrate the essential elements at Master's level for performing statistical inference. The process of inference is understood here as evaluating the relevance of a model with respect to a set of data. A concise bibliography can be found at the end of this document to start exploring the statistical approaches employed in high-energy (astro)physics and cosmology. The content of this document is largely inspired by the chapters "*Probability*" [1] and "*Statistics*" [2] of the *Particle Data Group* review, whose reading is strongly encouraged. The same applies to the reference book *Numerical Recipes* [3], in particular chapters 14 "Statistical *Description of Data*" and 15 "*Modeling of Data*". The whole book *Numerical Recipes* can be considered an essential prerequisite for a thesis in our fields. The book provides examples in C/C++, which serve as an excellent guide to understanding the analytic approaches. The present document favours the use of Python libraries, which can be tested in the exercises.

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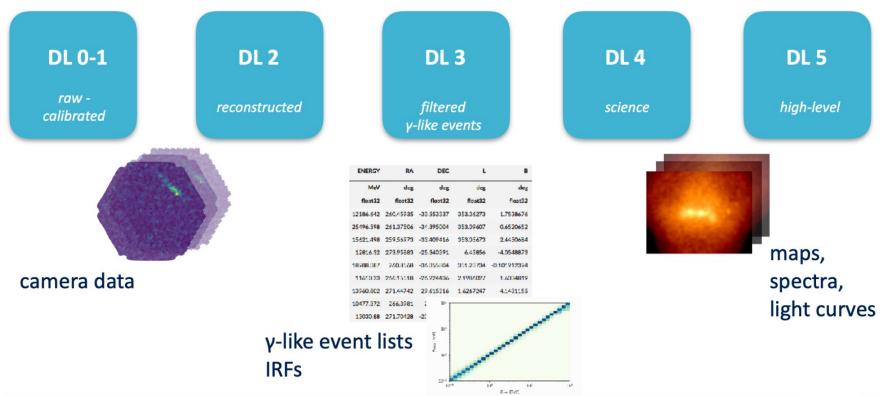
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[GammaPy team, CTAO School 2024]

$DL3_{\gamma-like events}$					
EVENT_ID	TIME	RA	DEC	ENERGY	
	5	deg	deg	TeV	
int64	float64	float32	float32	float32	
5407363825684	123890826.66805482	84.97964	23.89347	10.352011	
5407363825695	123890826.69749284	84.54751	21.004095	4.0246882	
5407363825831	123890827.23673964	85.39696	19.41868	2.2048872	
5407363825970	123890827.79615426	81.93147	20.79867	0.69548655	
5407363826067	123890828.26131463	85.98302	21.053099	0.86911184	
5407363826095	123890828.41393518	86.97305	21.837437	4.1240892	
5407363826128	123890828.52555823	83.40073	19.771587	1.6680022	
5407363826168	123890828.6829524	82.25036	19.22003	4.7649446	
5407363826383	123890829.53362775	83.18322	22.008213	0.7920148	

Observation and / or time selection

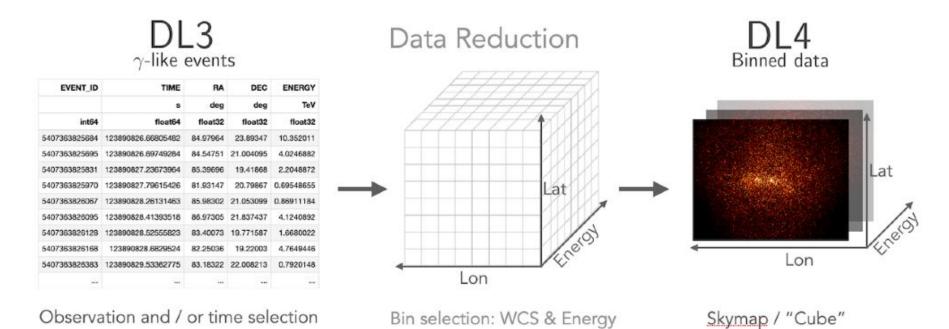
Bin selection: Region & Energy

Spectrum

Energy

DL4 Binned data

[GammaPy team, CTAO School 2024]



2.2 IRF factorisation

Equation 2.2 implies 7-dimensional instrument response functions that in general are computationally unmanageable. Simplifications can be achieved by making further assumptions, and in existing Imaging Air Cherenkov Telescope (IACT) experiments the IRF is generally factorised as follows:

$$R_i(\hat{\alpha}, \hat{\delta}, \hat{E} | \alpha, \delta, E, t) = A_i(\alpha, \delta, E, t) \times \text{PSF}_i(\hat{\alpha}, \hat{\delta} | \alpha, \delta, E, t) \times D_i(\hat{E} | \alpha, \delta, E, t)$$
(2.3)

where $A_i(\alpha, \delta, E, t)$ is the effective area in units of cm², $PSF_i(\hat{\alpha}, \hat{\delta} | \alpha, \delta, E, t)$ is the point spread function in units of sr⁻¹, with

$$d\hat{\Omega} \operatorname{PSF}_{i}(\hat{\alpha}, \hat{\delta} | \alpha, \delta, E, t) = 1$$
(2.4)

and $D_i(\hat{E}|\alpha, \delta, E, t)$ is the energy dispersion in units of TeV⁻¹, with

$$\int d\hat{E} \,\mathcal{D}_i(\hat{E}|\alpha,\delta,E,t) = 1 \tag{2.5}$$

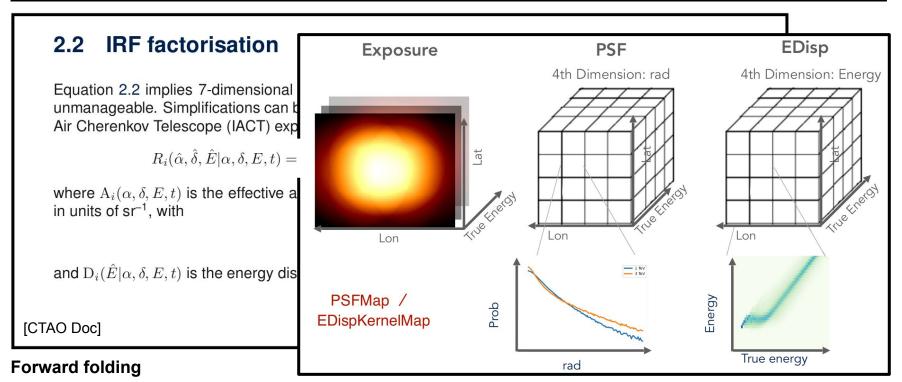
[CTAO Doc]

Forward folding

Convolve true emission (i.e. model as a function of true energy, direction...)

with transfer matrix (Instrument Response Function linking true and observed quantities)

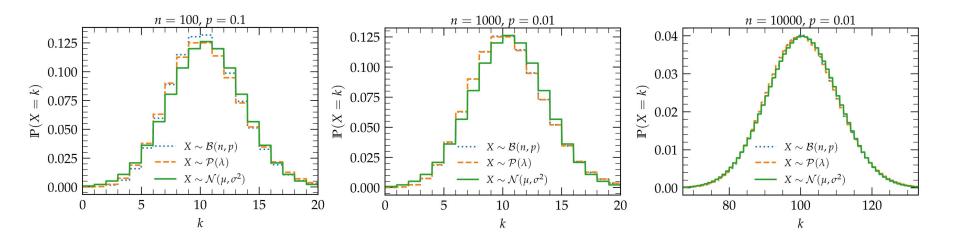
to enable a comparison to data in the observed space



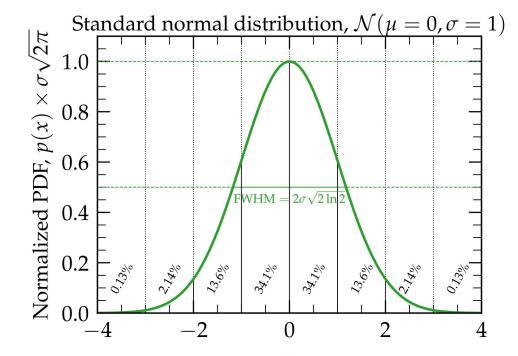
Convolve true emission (i.e. model as a function of true energy, direction...) with transfer matrix (Instrument Response Function linking true and observed quantities) to enable a comparison to data in the observed space

[GammaPy team, CTAO School 2024]

When Bernoulli meets Poisson and Gauss



From Gauss PMF to Gauss PDF



X

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V. Outro

Hessian and covariance matrix

OptimizeResult		θ_i	
backend method success message nfev total stat	: minuit : migrad : True : Optimization terminated successfully. : 244 : 86.12	$\hat{\theta}_i$	σ_i
CovarianceResult backend method	: minuit : hesse	,	$\begin{bmatrix} \sigma_i \\ \bullet \\ $
success message	: True : Hesse terminated successfully.		$ - \sigma_j - $
[GammaPy team]			[G. Cowan+ '24] $\hat{ heta}_j$ $ heta_j$

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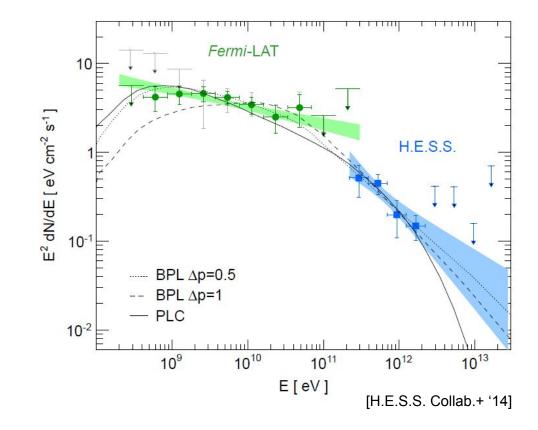
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V. Outro More... if we have time

Dealing with uncertainties

Bowtie / butterfly



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