

Lorentz invariance violation search with the Cherenkov Telescope Array Observatory Large-Sized Telescope

8th Heidelberg symposium on high-energy gamma-ray astronomy

Cyann Plard & Sami Caroff

on behalf on the LST collaboration

The logo for the Cherenkov Telescope Array Observatory (CTAO) features the letters 'CTAO' in a bold, white, sans-serif font. A small, stylized blue and white starburst is positioned between the 'A' and 'O'. The background of the logo is a dark, circular shape with a white outline, resembling a telescope dish or a celestial body.

LST
COLLABORATION

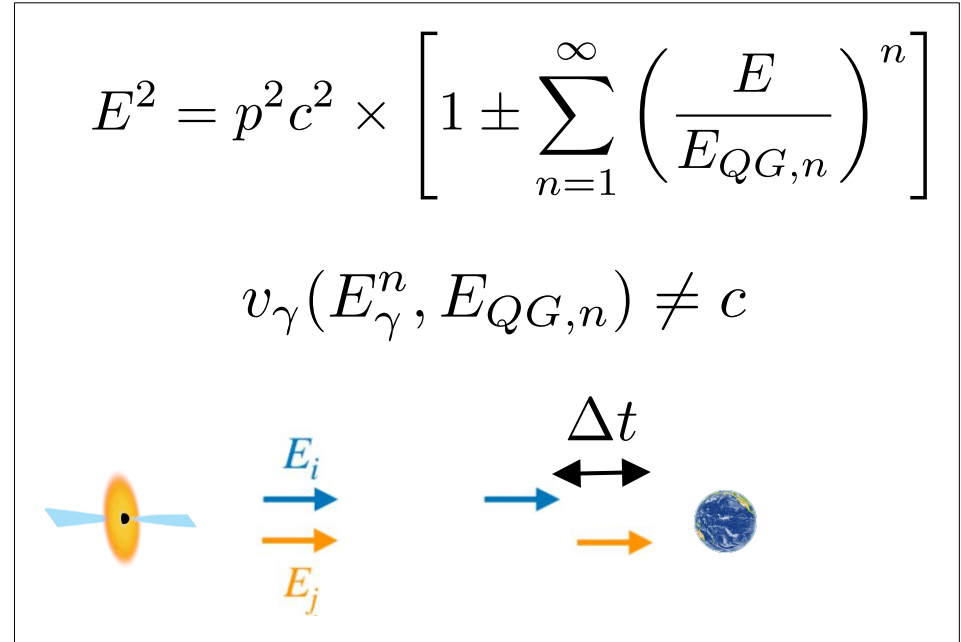
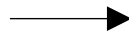
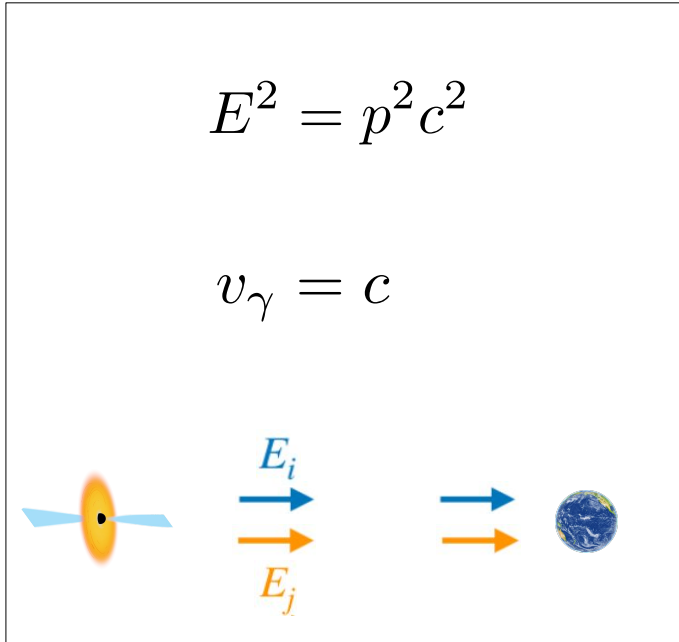
The logo for the Centre National de la Recherche Scientifique (CNRS) consists of the letters 'CNRS' in a white, sans-serif font, set against a dark blue circular background.The logo for the Laboratoire d'Annecy de Physique des Particules (LAPP) features the letters 'LAPP' in a white, sans-serif font, with a stylized blue and white particle track graphic to the left. Below the letters, the full name 'Laboratoire d'Annecy de Physique des Particules' is written in a smaller, white font.The logo for Enigmass features the word 'Enigmass' in a white, sans-serif font, with a stylized orange and white particle track graphic to the right.

Credit : Me

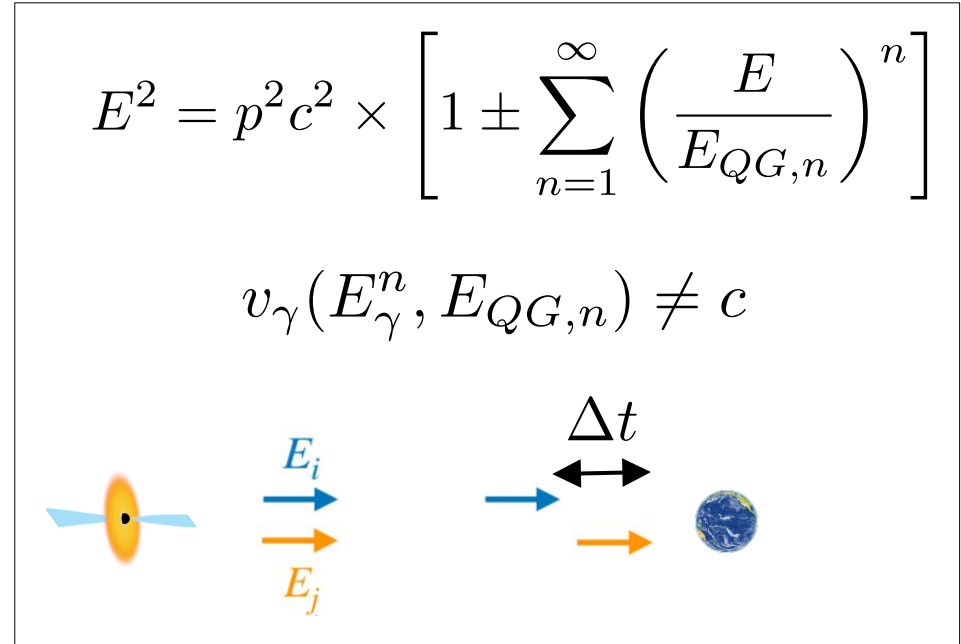
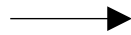
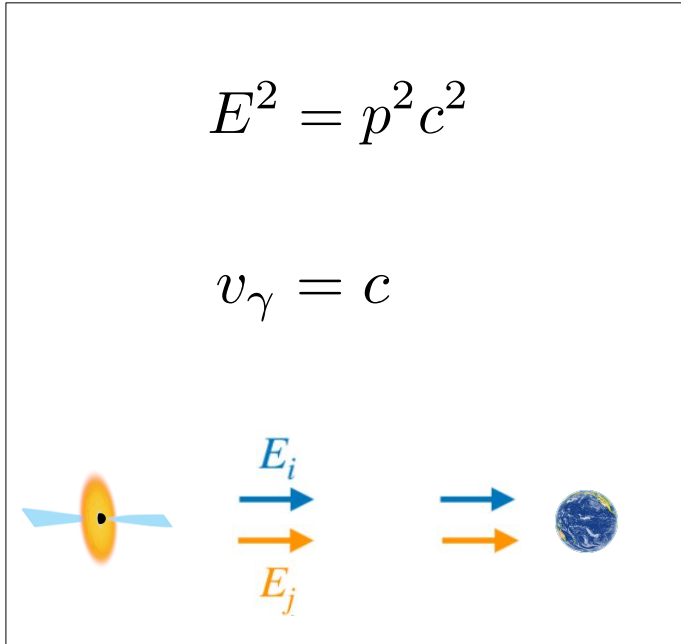
- Unification of general relativity and quantum field theory:
quantum gravity
 - ▶ effects of QG expected to appear at Planck scale $E_P \sim 10^{19} \text{GeV}$

- Some quantum gravity models allow a violation of Lorentz invariance
 - ▶ may be observable using high energetic gamma-rays

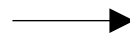
Lorentz invariance: speed of light C in vacuum is energy-independent



Lorentz invariance: speed of light C in vacuum is energy-independent



$$\lambda_n = \frac{\Delta t}{\Delta E^n \kappa_n(z)} = \pm \frac{n+1}{2H_0 E_{QG,n}^n}$$

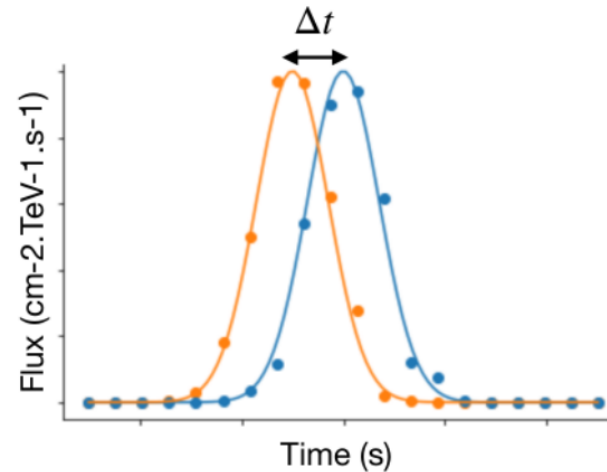
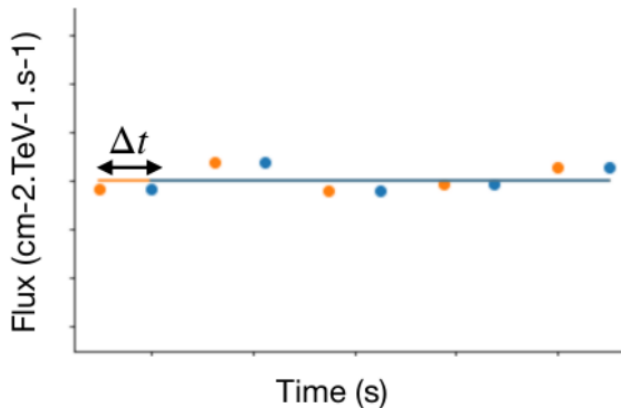
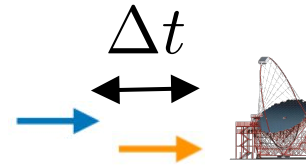
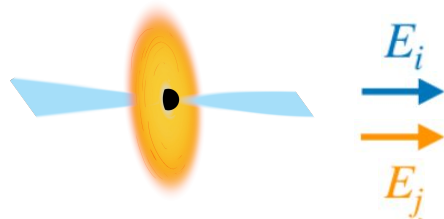


Search for a limit on $E_{QG,n}$ at the first order $n = 1$

- Large range of energy
- Cosmological distance
- Highly variable and active source

$$\Delta t_{LIV} = \pm \frac{n+1}{2} \frac{\Delta E^n}{E_{QG,n}^n} \times \kappa_n(z)$$

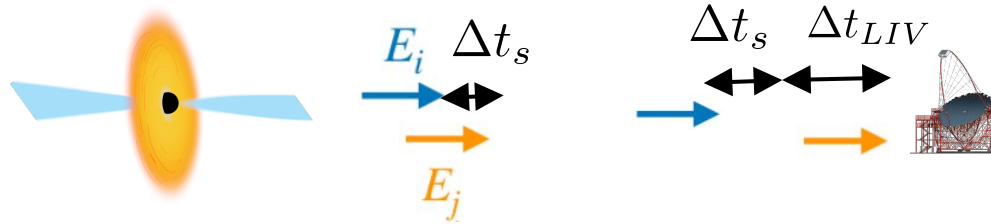
→ blazars, gamma-ray bursts, pulsars



No guarantee that photons are emitted at the same time

→ Intrinsic source delay:

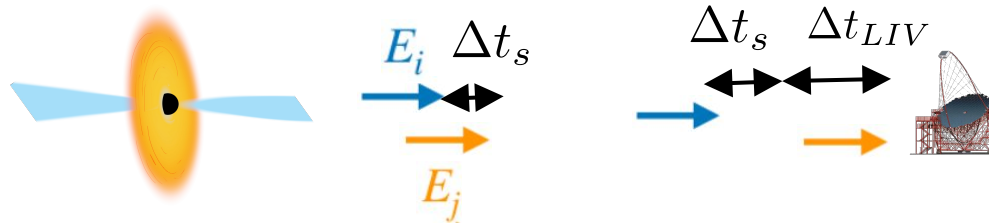
$$\Delta t = \Delta t_{LIV} + (1 + z) \Delta t_{source}$$



No guarantee that photons are emitted at the same time

→ Intrinsic source delay:

$$\Delta t = \Delta t_{LIV} + (1 + z)\Delta t_{source}$$



Δt_{source}

Redshift-independent

Sources and flares -dependent

Δt_{LIV}

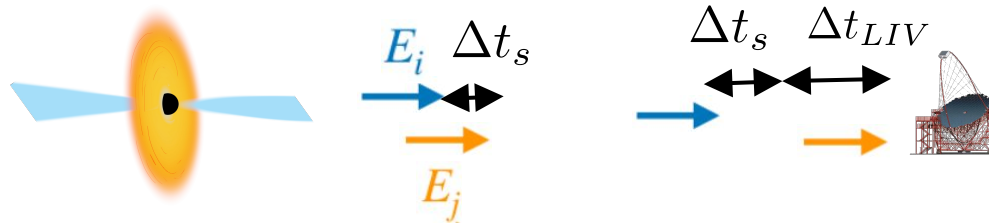
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→ Intrinsic source delay:

$$\Delta t = \Delta t_{LIV} + (1 + z)\Delta t_{source}$$


 Δt_{source}

Redshift-independent

Sources and flares -dependent

 Δt_{LIV}

Redshift-dependent

Sources and flares -independent

→ Combination of different flares and different (types of) sources

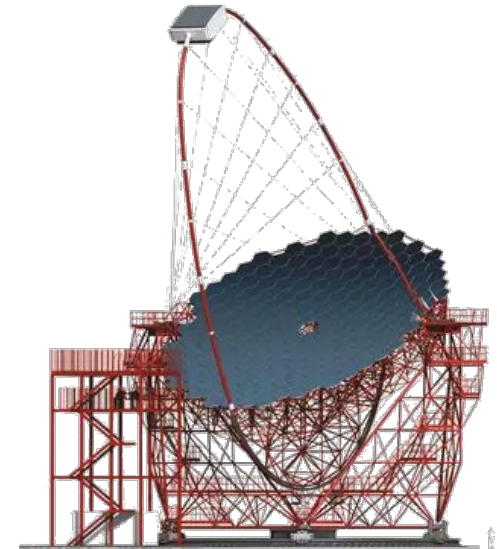
- Analysis already performed with various individual flares and/or sources observed with Cherenkov telescopes but **data were never combined**
- *Gamma-ray LIV Working Group (γ LIV WG)* between H.E.S.S., MAGIC, VERITAS and LST to **combine** different types of sources and flares from various experiments
- Combination of **simulations** of data from experiments involved in the γ LIV WG (without LST) already performed: *Bolmont et. al., 2022, ApJ 930 75*
- This work: prototype of a global and consistent analysis on all **blazar** data of **LST-1** then combine it with the γ LIV WG data



Several AGNs data of LST-1 from January 2021 to June 2023

	BL Lac	M87	1ES 1959+650	PG 1553+113	Mrk421	Mrk501	TON0599
Redshift z	0.069	0.0043	0.047	0.433 <i>(Jones et al. 2021)</i>	0.03002	0.034	0.72
Observation time (hours)	44	9	13	24	72	66	9

237 hours of observations

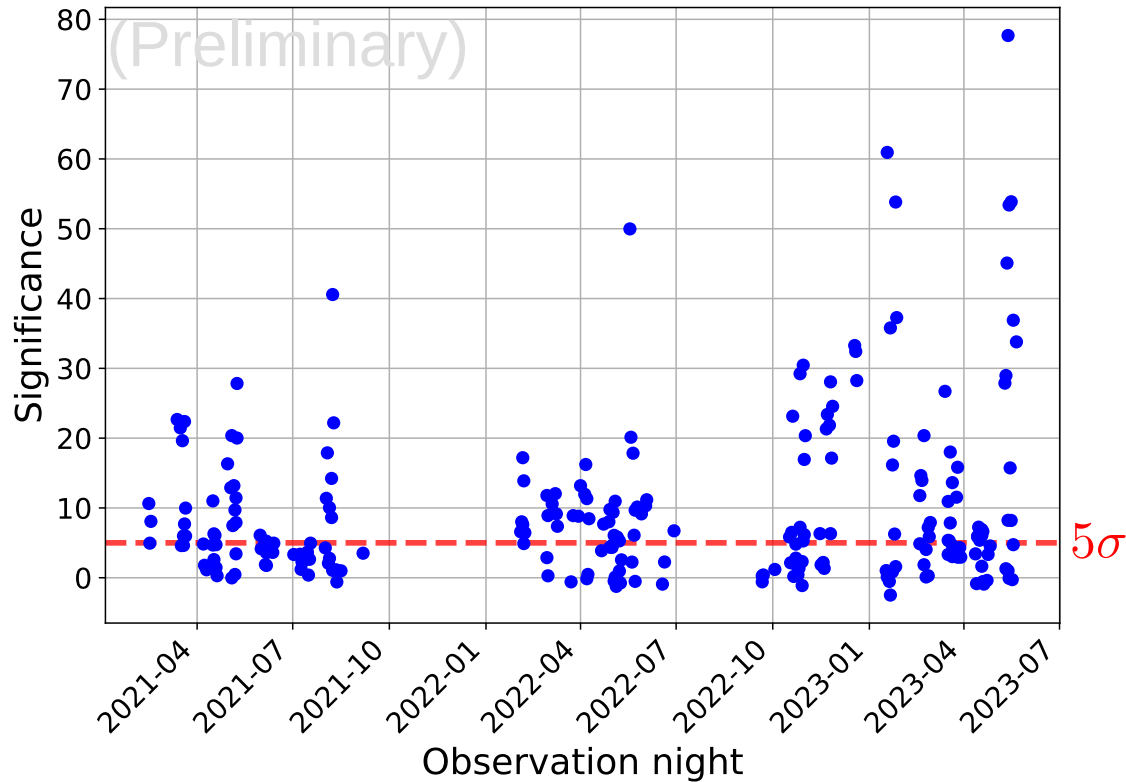


Several AGNs data of LST-1 from January 2021 to June 2023

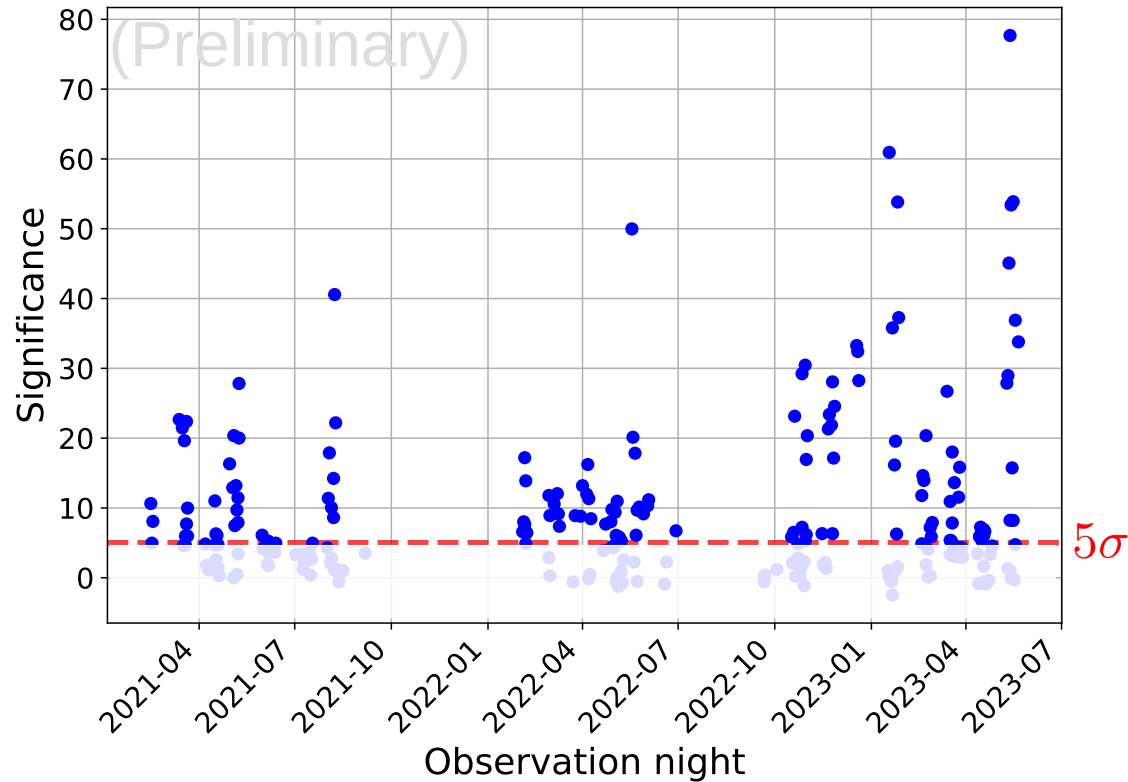


Variability test on each significant observation night

Several AGNs data of LST-1 from January 2021 to June 2023



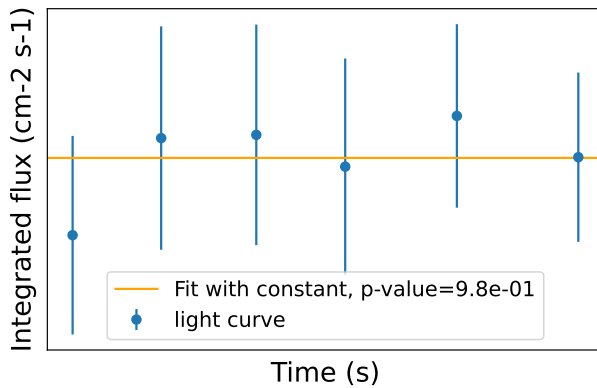
Several AGNs data of LST-1 from January 2021 to June 2023



Several AGNs data of LST-1 from January 2021 to June 2023

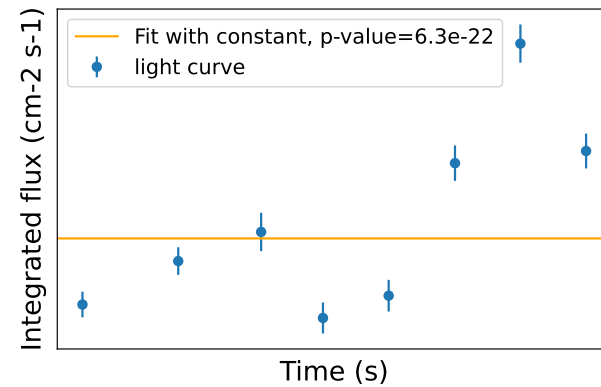
Variability test on each significant observation night

Non-variable sample

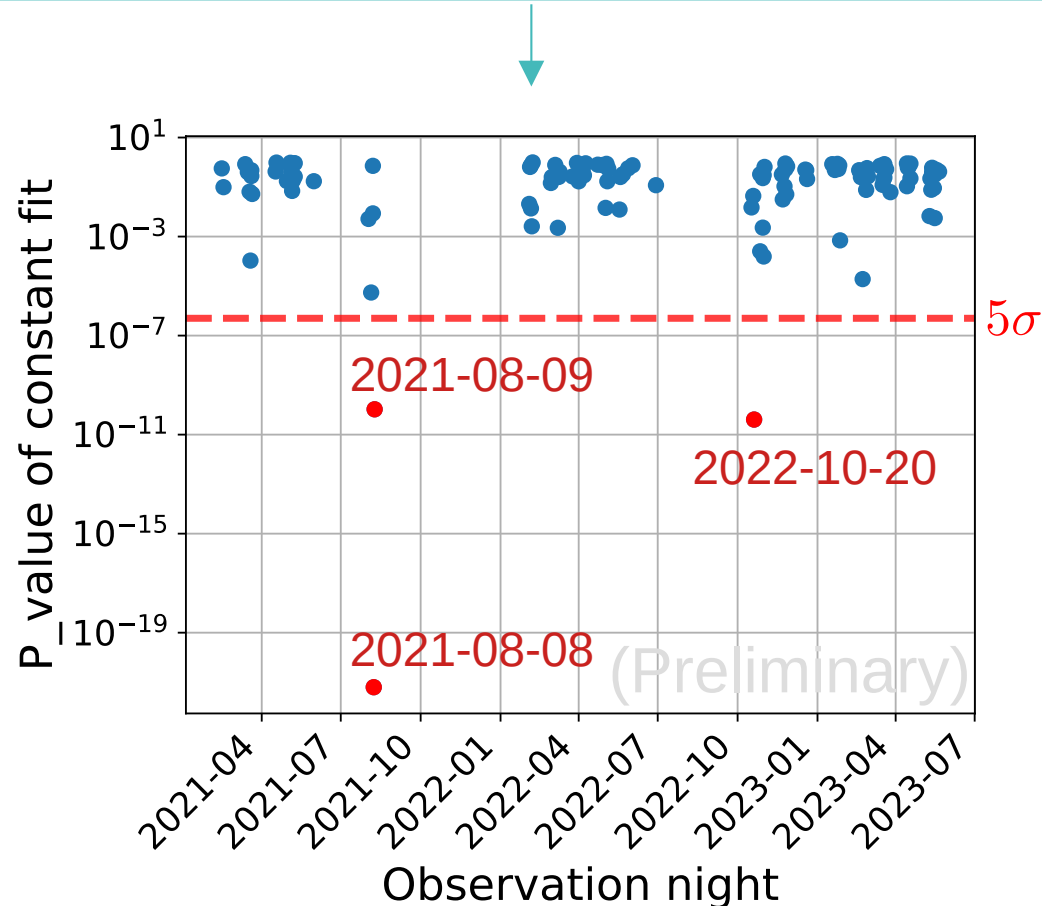


Constant fit
pvalue < 3×10^{-7}
(5 σ)

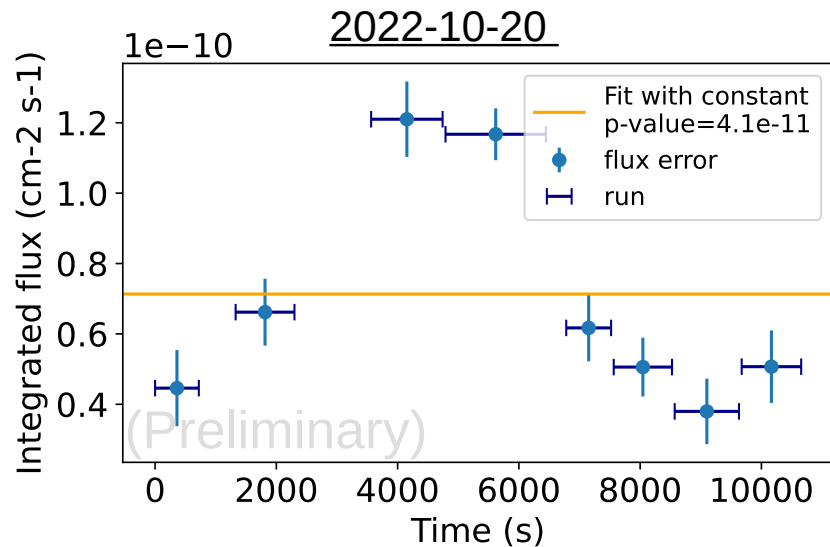
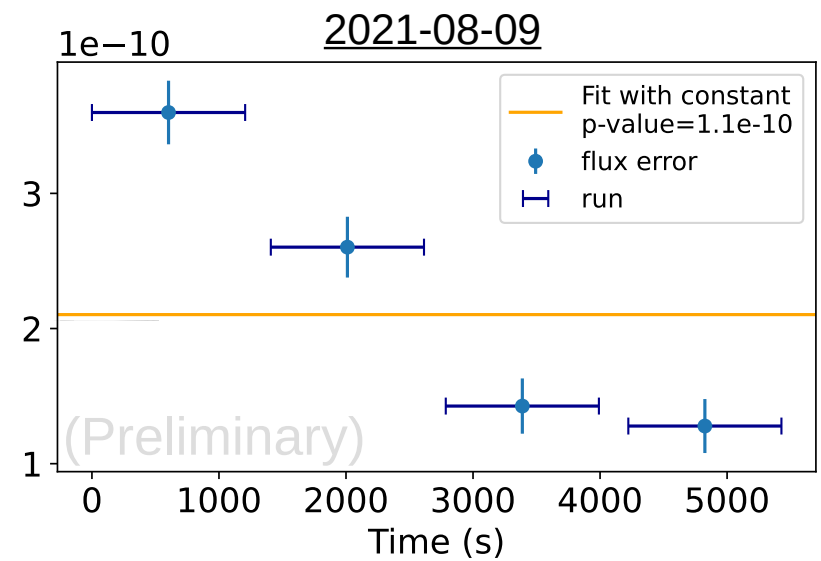
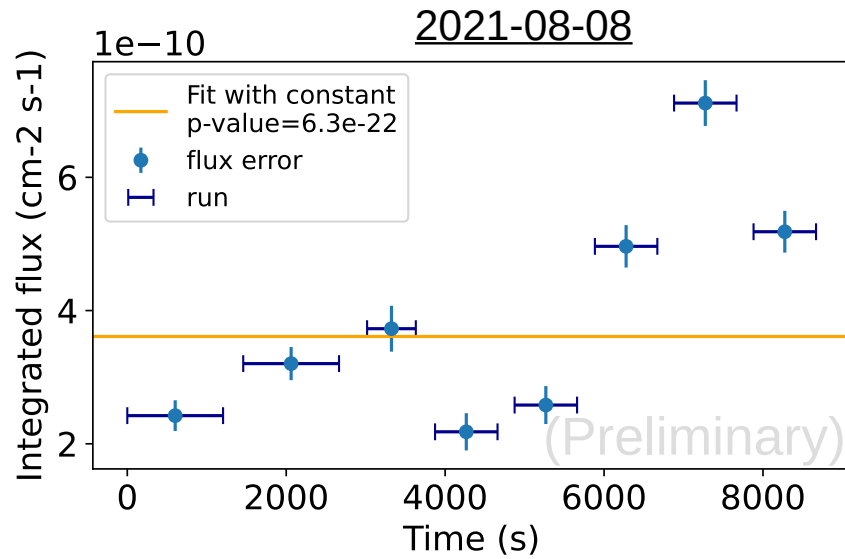
Variable sample



Several AGNs data of LST-1 from January 2021 to June 2023



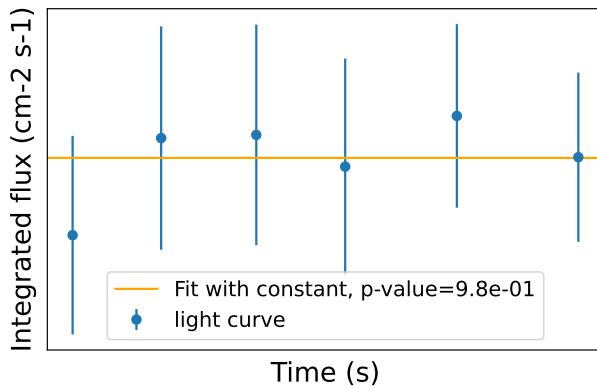
Found 1 source showing intra-night variability: **BL Lacertae** with 3 nights (~7h)



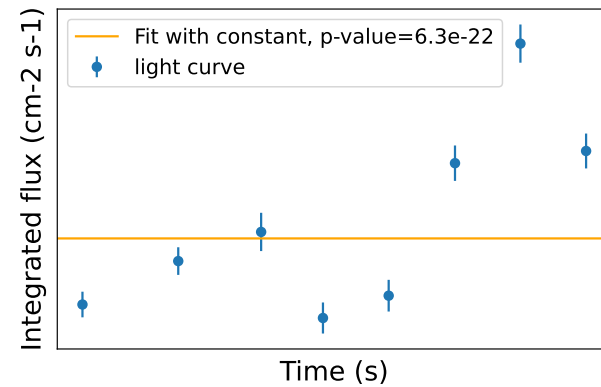
Several AGNs data of LST-1 from January 2021 to June 2023

Variability test on each significant observation night

Non-variable sample



Variable sample

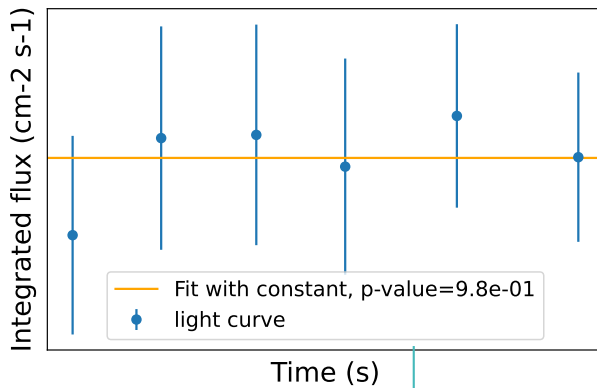


Constant fit
pvalue < 5σ

Several AGNs data of LST-1 from January 2021 to June 2023

Variability test on each significant observation night

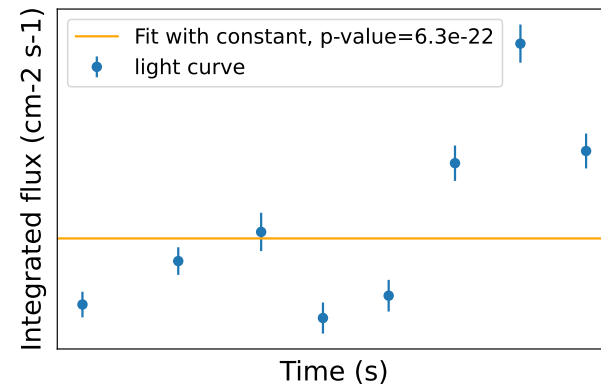
Non-variable sample



Optimization of analysis cut

Constant fit
pvalue < 5σ

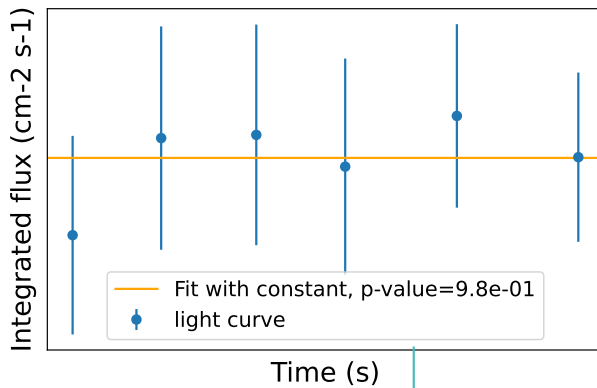
Variable sample



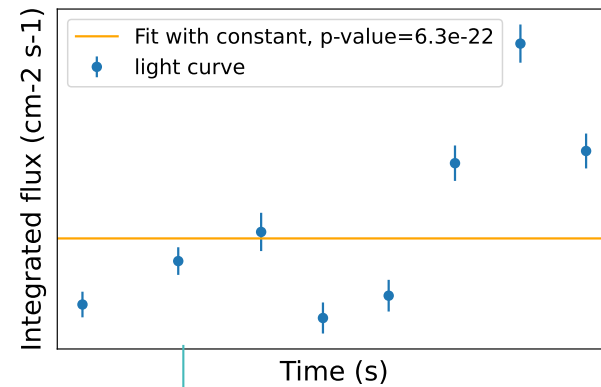
Several AGNs data of LST-1 from January 2021 to June 2023

Variability test on each significant observation night

Non-variable sample



Variable sample



Constant fit
pvalue < 5σ

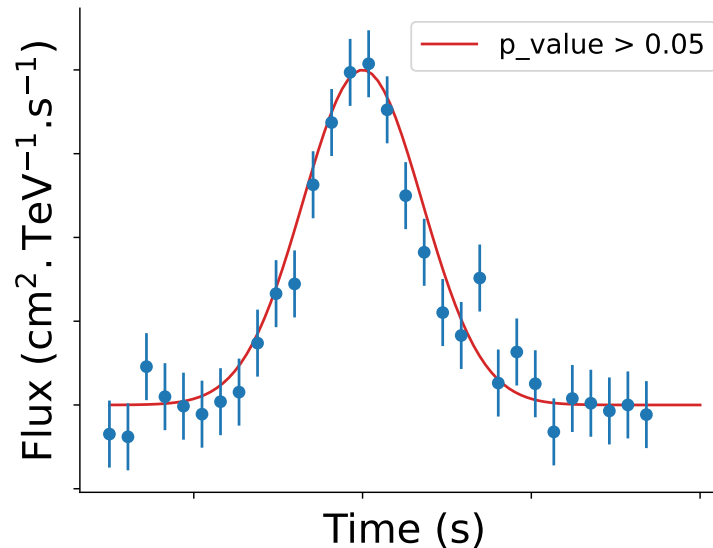
Optimization of
analysis cut

Search for a variability pattern and extract sample
properties (energetic and temporal distributions)

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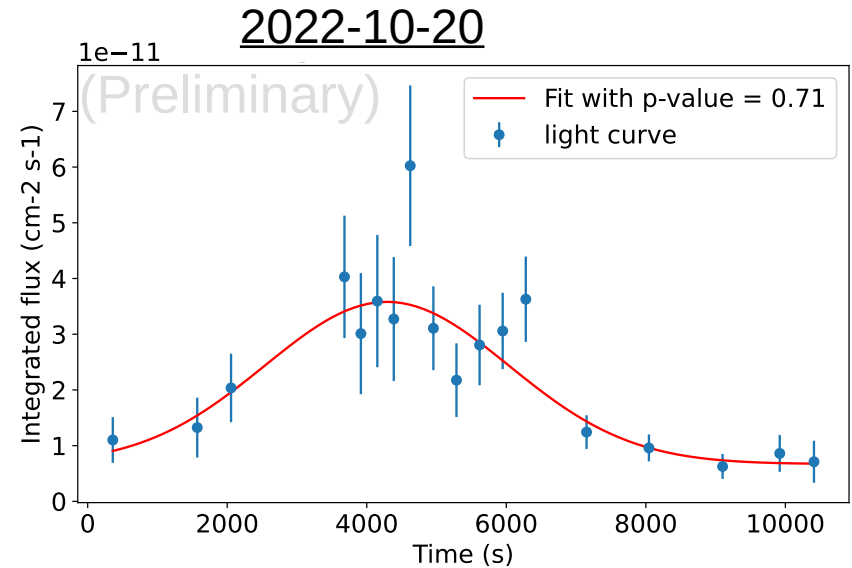
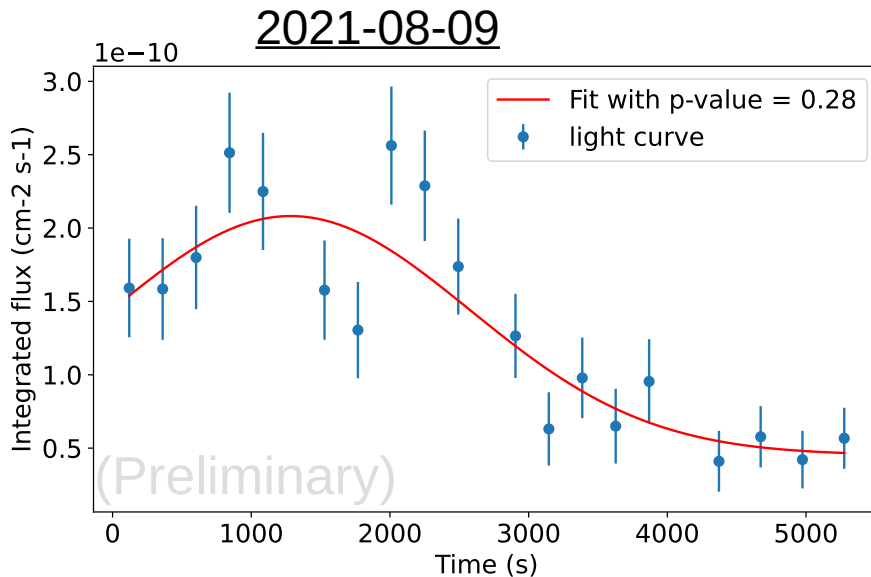
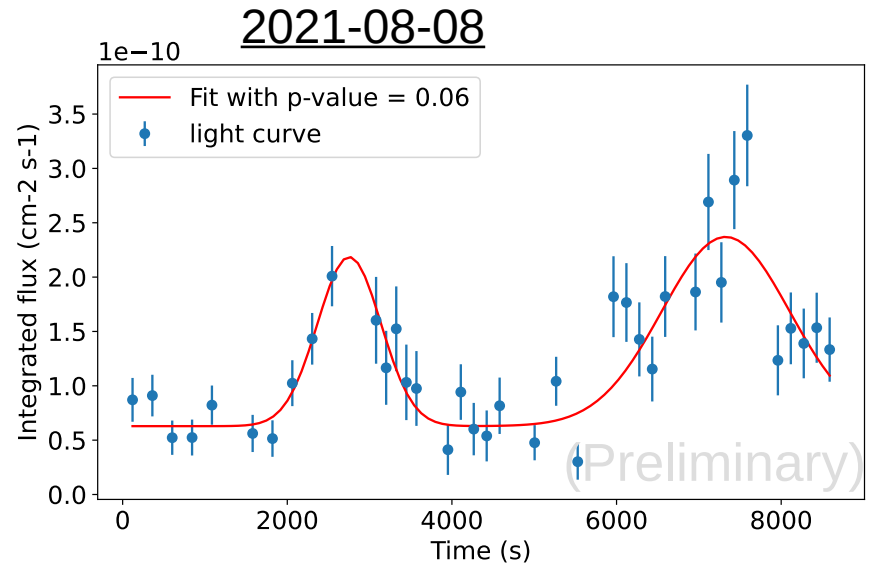


- Find a model to describe the lightcurve variability: non-rejected if $p\text{-value} > 0.05$ (2σ)



$$\blacksquare A_1 e^{-\frac{(t-\mu_1)^2}{2\sigma_1^2}} + A_2 e^{-\frac{(t-\mu_2)^2}{2\sigma_2^2}} + C_0$$

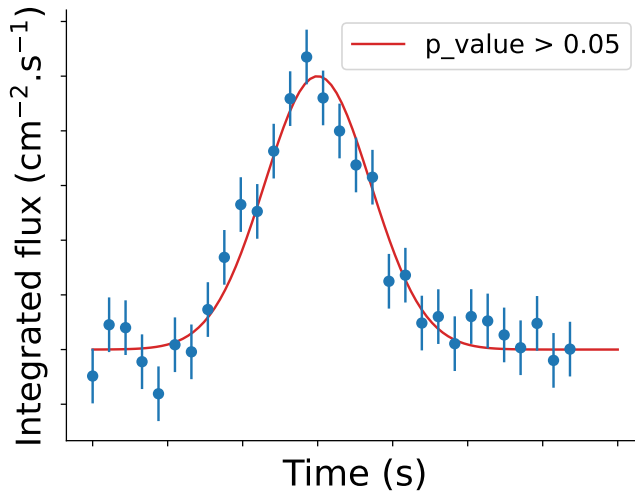
$$\blacksquare A e^{-\frac{(t-\mu)^2}{2\sigma^2}} + C$$



Search for a variability pattern and extract sample properties (energetic and temporal distributions)



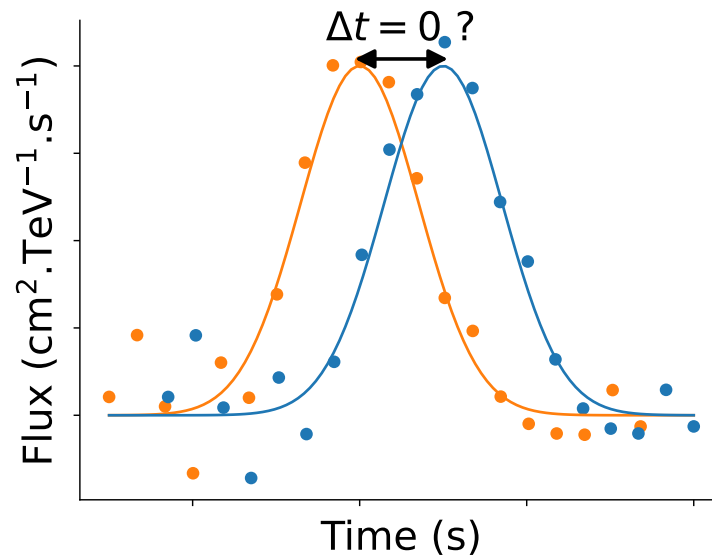
- Find a model to describe the lightcurve variability: non-rejected if $p\text{-value} > 0.05$ (2σ)



(Preliminary)	Lightcurve model (Gaussians)						
	C	A_1	μ_1	σ_1	A_2	μ_2	σ_2
	($\text{cm}^{-2} \cdot \text{s}^{-1}$)	($\text{cm}^{-2} \cdot \text{s}^{-1}$)	(s)	(s)	($\text{cm}^{-2} \cdot \text{s}^{-1}$)	(s)	(s)
2021-08-08	6.3e-11	1.6e-10	2752	386	1.8e-10	7326	778
2021-08-09	4.5e-11	1.7e-10	1284	1293			
2022-10-20	6.7e-12	2.9e-11	4288	1750			

Search for a variability pattern and extract sample properties (energetic and temporal distributions)

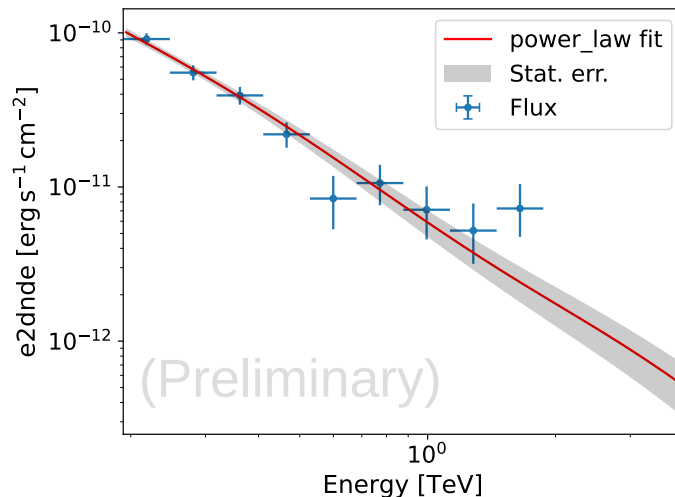
- Find a model to describe the lightcurve variability: non-rejected if p-value > 0.05 (2σ)
- Check: no significant disagreement between **low** and **high** energies (median of counts)



Search for a variability pattern and extract sample properties (energetic and temporal distributions)



- Find a model to describe the lightcurve variability: non-rejected if p-value > 0.05 (2σ)
- Check: no significant disagreement between low and high energies (median of counts)
- Check: no significant time-variation of energetic distribution (spectra parameters)



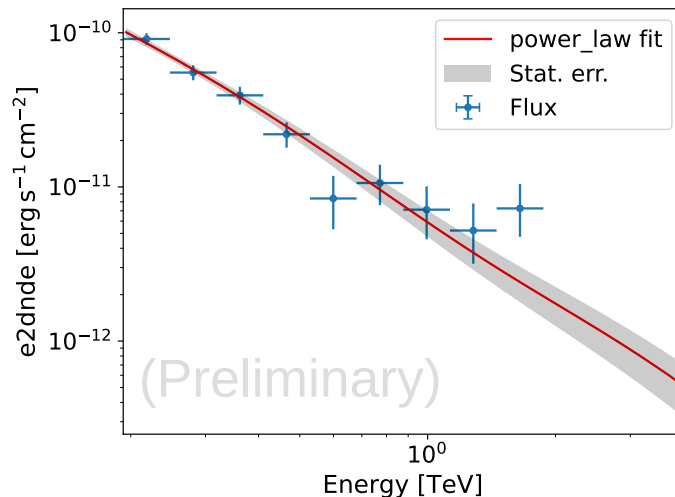
$$A_0 \left(\frac{E}{E_0} \right)^{-\alpha}$$

(Preliminary)	Spectra model (power law)
	Index α
2021-08-08	3.44±0.2
2021-08-09	3.45±0.1
2022-10-20	3.48±0.2

Search for a variability pattern and extract sample properties (energetic and temporal distributions)



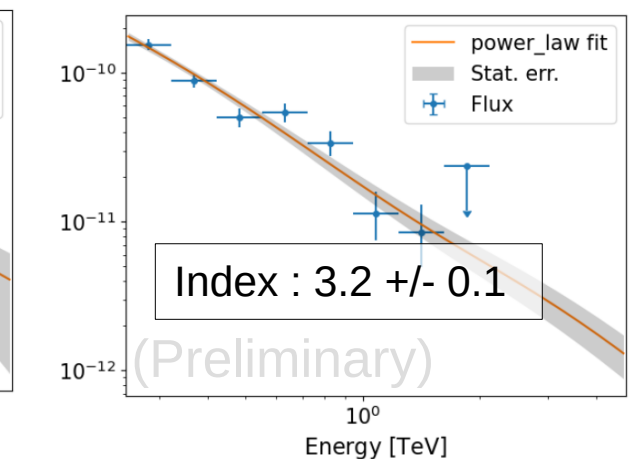
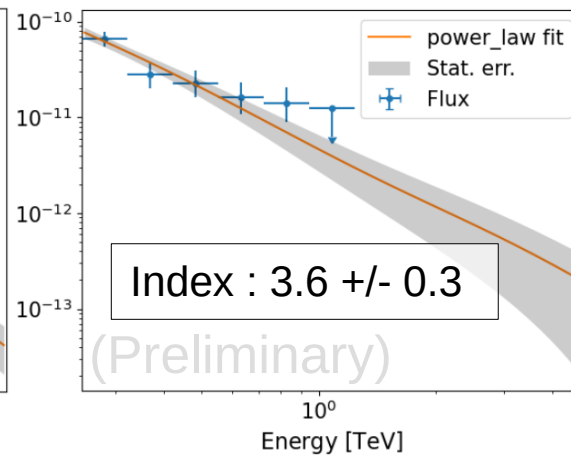
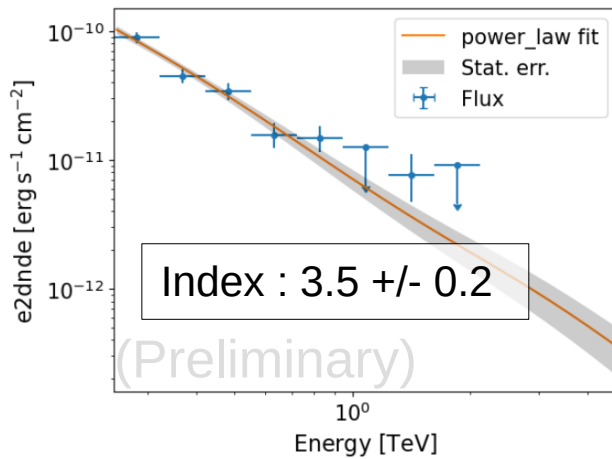
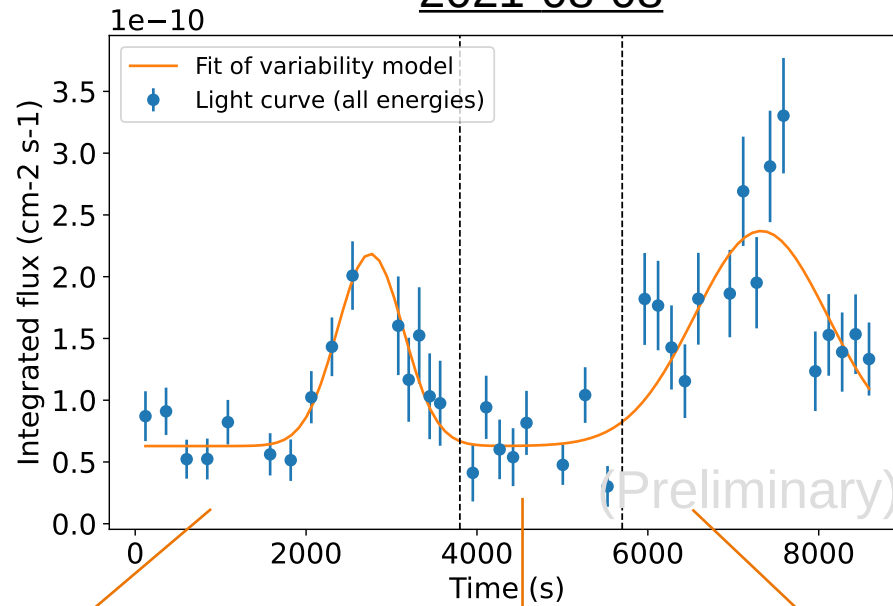
- Find a model to describe the lightcurve variability: non-rejected if p-value > 0.05 (2σ)
- Check: no significant disagreement between low and high energies (median of counts)
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$$A_0 \left(\frac{E}{E_0} \right)^{-\alpha}$$

time-independent index ?

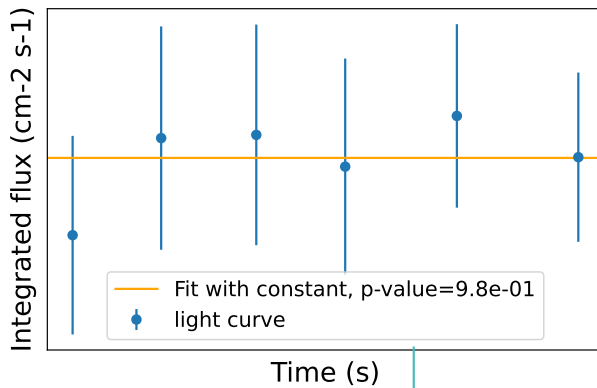
2021-08-08



Several AGNs data of LST-1 from January 2021 to June 2023

Variability test on each significant observation night

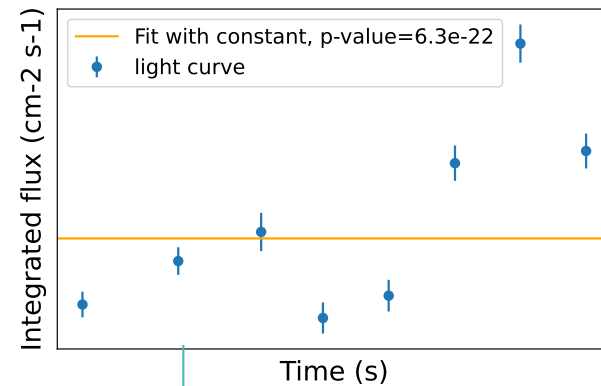
Non-variable sample



Optimization of analysis cut

Constant fit
pvalue < 5σ

Variable sample

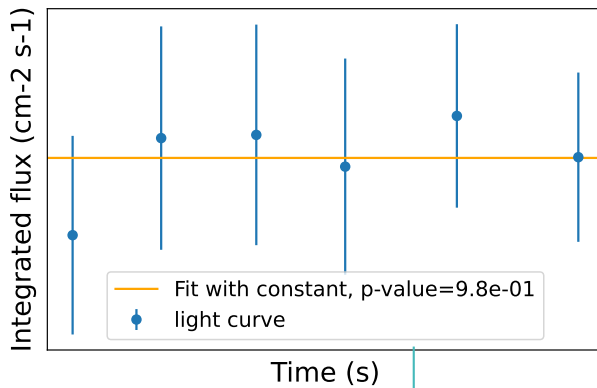


Search for a variability pattern and extract sample properties (energetic and temporal distributions)

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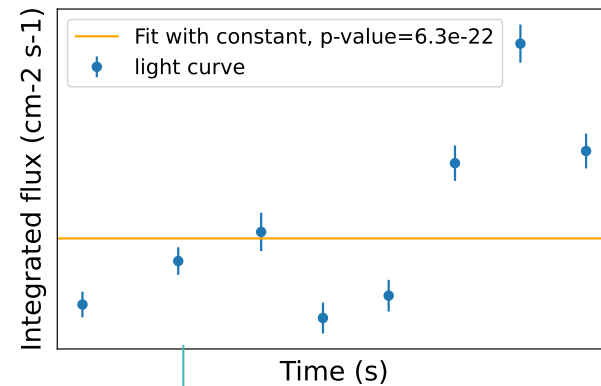
Variability test on each significant observation night

Non-variable sample



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Variable sample



Optimization of
analysis cut

Search for a variability pattern and extract sample
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LIV analysis

LIV analysis



LIVelihood software

- Created in the context of the γ LIV working group
- Uses the **likelihood method**:
 λ_n is a free parameter that minimizes the likelihood function

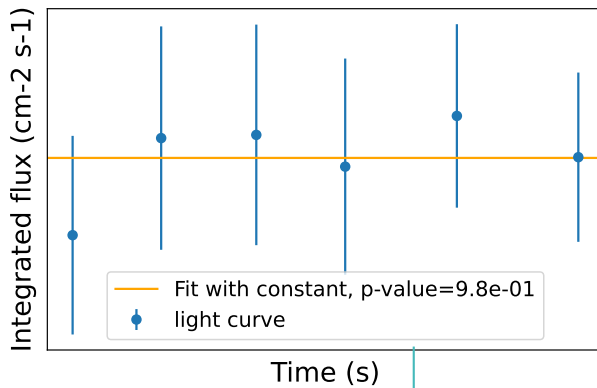
for one source (or night):
$$\mathcal{L}_S(\lambda_n) = - \sum_{\text{event } i} \log \left(\frac{dP(E_{R,i}, t_i; \lambda_n)}{dE_R dt} \right)$$

for combination:
$$\mathcal{L}_{\text{comb}}(\lambda_n) = \sum_{\text{source } S} \mathcal{L}_S(\lambda_n)$$

Several AGNs data of LST-1 from January 2021 to June 2023

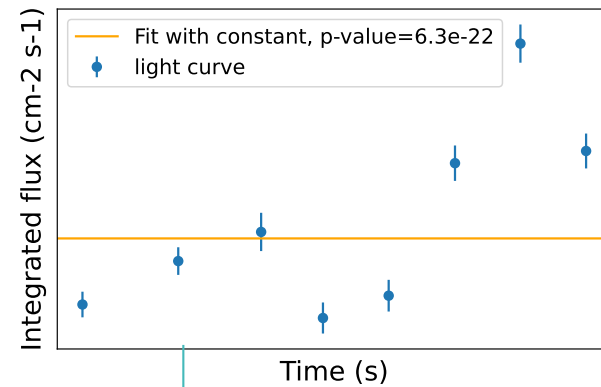
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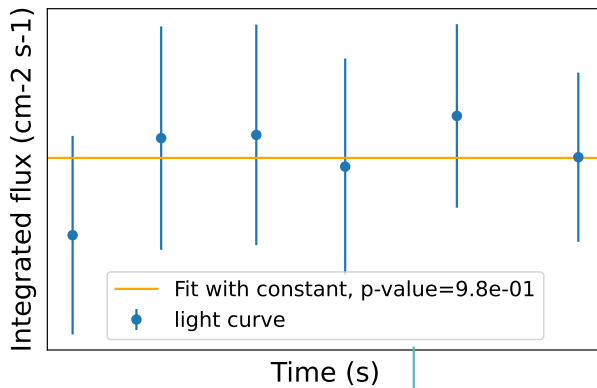
Search for a variability pattern and extract sample
properties (energetic and temporal distributions)

LIV analysis

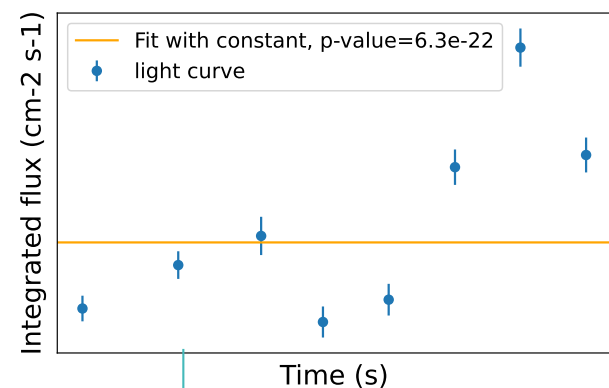
Several AGNs data of LST-1 from January 2021 to June 2023

Variability test on each significant observation night

Non-variable sample



Variable sample



Constant fit
pvalue < 5σ

Optimization of
analysis cut

Search for a variability pattern and extract sample
properties (energetic and temporal distributions)

LIV analysis

Sample of simulations for calibration

Lag λ_n : free parameter, can be shared between sources with different redshifts

For one night:
$$\mathcal{L}(\lambda_n) = - \sum_{\text{event } i} \log \left(\frac{dP(E_{R,i}, t_i; \lambda_n)}{dE_R dt} \right)$$

with
$$\frac{dP}{dE_R dt} = W_s \frac{\int E_{\text{ff}} A(E_T, t) \text{MM}(E_T, E_R) \times F_s(E_T, t; \lambda_n) dE_T}{N'_s}$$

$$+ \sum_{\mathbf{k}=\{\mathbf{b}, \mathbf{h}\}} W_{\mathbf{k}} \frac{\int E_{\text{ff}} A(E_T, t) \text{MM}(E_T, E_R) \times F_{\mathbf{k}}(E_T) dE_T}{N'_{\mathbf{k}}}$$

Lag λ_n : free parameter, can be shared between sources with different redshifts

For one night:
$$\mathcal{L}(\lambda_n) = - \sum_{\text{event } i} \log \left(\frac{dP(E_{R,i}, t_i; \lambda_n)}{dE_R dt} \right)$$

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$$\frac{dP}{dE_R dt} = W_s \frac{\int E_{\text{ff}} A(E_T, t) \text{MM}(E_T, E_R) \times F_s(E_T, t; \lambda_n) dE_T}{N'_s}$$

Signal

$$+ \sum_{\mathbf{k}=\{\mathbf{b}, \mathbf{h}\}} W_{\mathbf{k}} \frac{\int E_{\text{ff}} A(E_T, t) \text{MM}(E_T, E_R) \times F_{\mathbf{k}}(E_T) dE_T}{N'_{\mathbf{k}}}$$

Backgrounds k: baseline and hadrons

Lag λ_n : free parameter, can be shared between sources with different redshifts

For one night:
$$\mathcal{L}(\lambda_n) = - \sum_{\text{event } i} \log \left(\frac{dP(E_{R,i}, t_i; \lambda_n)}{dE_R dt} \right)$$

with
$$\frac{dP}{dE_R dt} = W_s \frac{\int E_{\text{ff}} A(E_T, t) \text{MM}(E_T, E_R) \times F_s(E_T, t; \lambda_n) dE_T}{N'_s}$$

↓
Lightcurve * spectra
↑

$$+ \sum_{\mathbf{k}=\{\mathbf{b}, \mathbf{h}\}} W_{\mathbf{k}} \frac{\int E_{\text{ff}} A(E_T, t) \text{MM}(E_T, E_R) \times F_{\mathbf{k}}(E_T) dE_T}{N'_{\mathbf{k}}}$$

	Spectral model (power law)	Lightcurve model (Gaussians)						
	index α	C (cm ⁻² .s ⁻¹) (baseline background)	A ₁ (cm ⁻² .s ⁻¹)	μ_1 (s)	σ_1 (s)	A ₂ (cm ⁻² .s ⁻¹)	μ_2 (s)	σ_2 (s)
2021-08-08	3.44	6.3e-11	1.6e-10	2752	386	1.8e-10	7326	778
2021-08-09	3.45	4.5e-11	1.7e-10	1284	1293			
2022-10-20	3.48	6.7e-12	2.9e-11	4288	1750			
hadronic background	2.7 (Aguilar et al, 2015)	C						

(Preliminary)

with
$$\frac{dP}{dE_R dt} = W_s \frac{\int E_{ff} A(E_T, t) MM(E_T, E_R) \times F_s(E_T, t; \lambda_n) dE_T}{N'_s}$$

↓
Lightcurve * spectra

↑

$$+ \sum_{\mathbf{k}=\{\mathbf{b}, \mathbf{h}\}} W_k \frac{\int E_{ff} A(E_T, t) MM(E_T, E_R) \times F_k(E_T) dE_T}{N'_k}$$

Lag λ_n : free parameter, can be shared between sources with different redshifts

For one night:
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with
$$\frac{dP}{dE_R dt} = W_s \frac{\int E_{\text{ff}} A(E_T, t) \text{MM}(E_T, E_R) \times F_s(E_T, t; \lambda_n) dE_T}{N'_s}$$

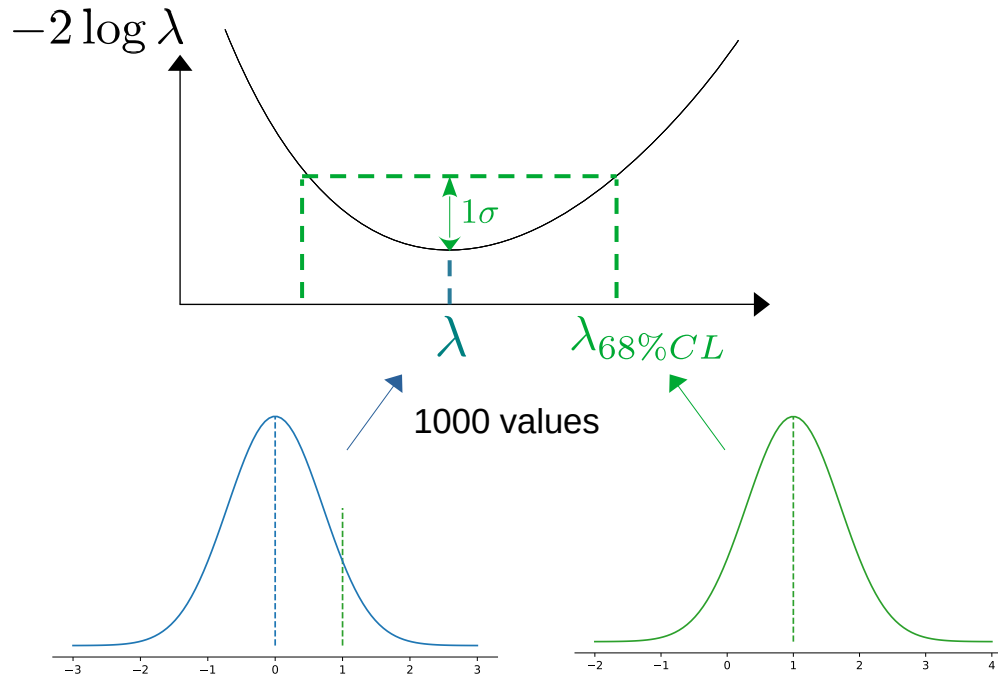
↓
Instrumental response functions
↑

$$+ \sum_{\mathbf{k}=\{\mathbf{b}, \mathbf{h}\}} W_{\mathbf{k}} \frac{\int E_{\text{ff}} A(E_T, t) \text{MM}(E_T, E_R) \times F_{\mathbf{k}}(E_T) dE_T}{N'_{\mathbf{k}}}$$

Simulations sample for calibration

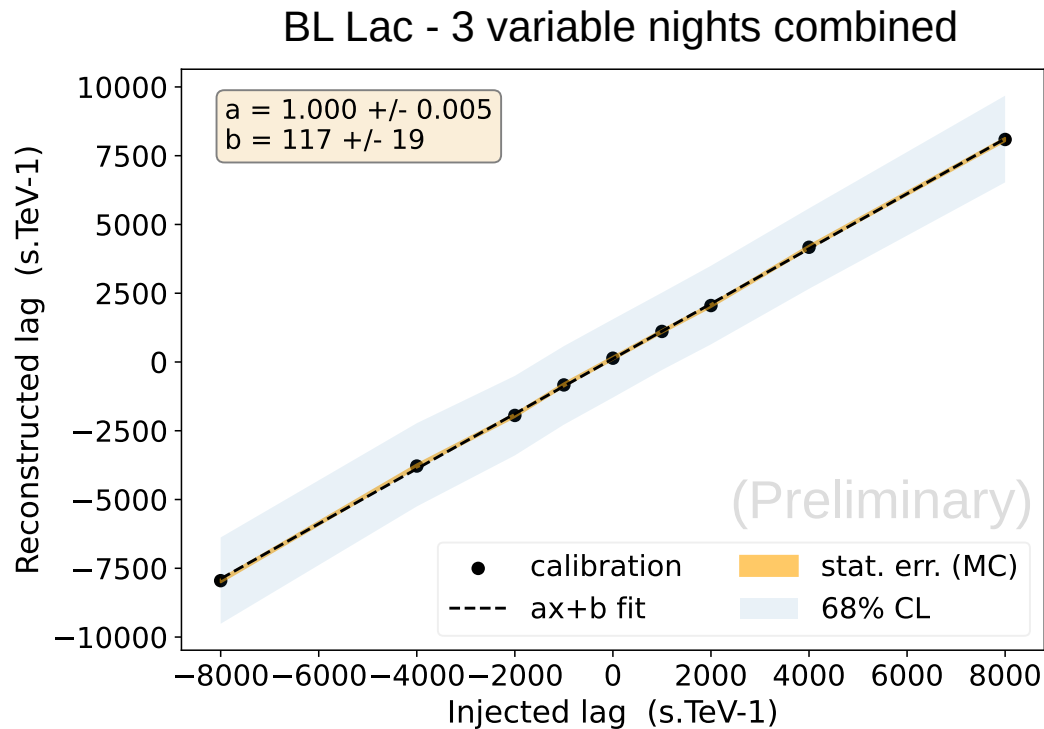


- Perform 1000 dataset simulations



Simulations sample for calibration

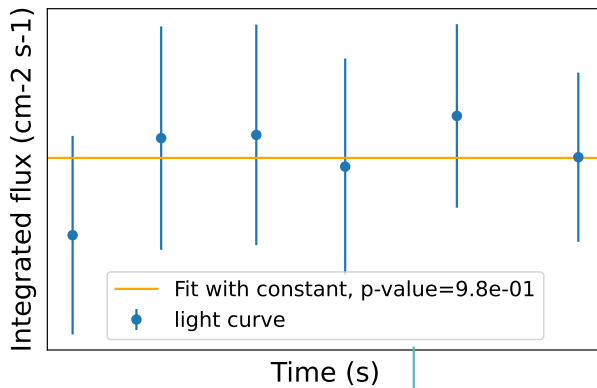
- Perform 1000 dataset simulations
- Calibration: inject lag to verify that LIVelihood reconstructs it well



Several AGNs data of LST-1 from January 2021 to June 2023

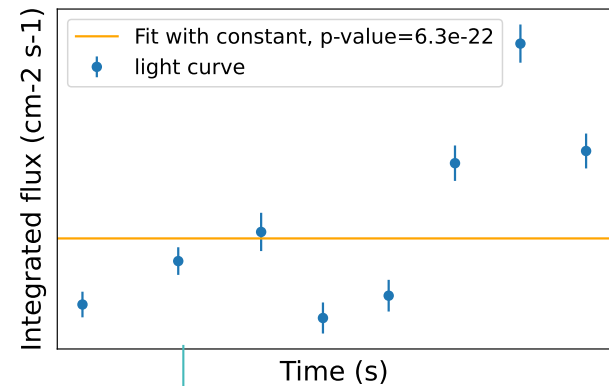
Variability test on each significant observation night

Non-variable sample



Constant fit
pvalue < 5σ

Variable sample



Optimization of
analysis cut

Search for a variability pattern and extract sample
properties (energetic and temporal distributions)

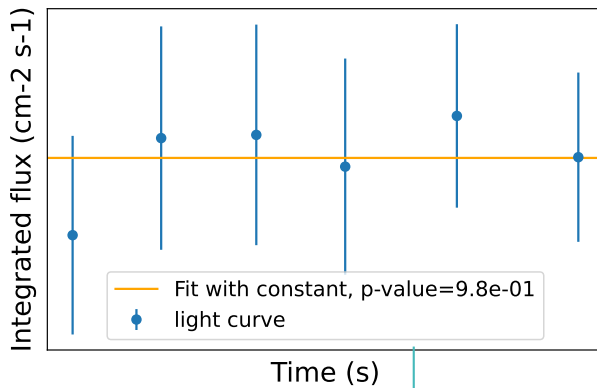
LIV analysis

Sample of simulations for calibration

Several AGNs data of LST-1 from January 2021 to June 2023

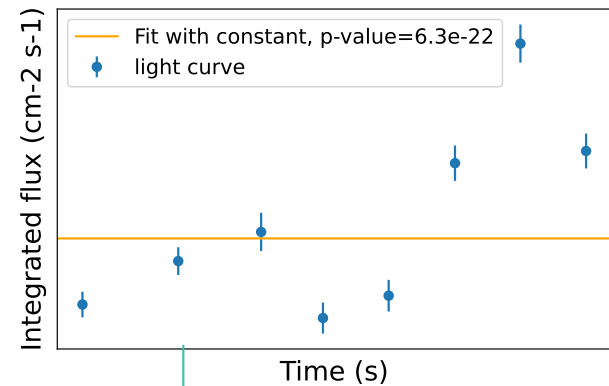
Variability test on each significant observation night

Non-variable sample



Constant fit
pvalue < 5σ

Variable sample



Optimization of
analysis cut

Search for a variability pattern and extract sample
properties (energetic and temporal distributions)

LIV analysis

Sample of simulations for calibration

Application on real data

Lag λ_n : free parameter, can be shared between sources with different redshifts

For one night:
$$\mathcal{L}(\lambda_n) = - \sum_{\text{event } i} \log \left(\frac{dP(E_{R,i}, t_i; \lambda_n)}{dE_R dt} \right)$$

with
$$\frac{dP}{dE_R dt} = W_s \frac{\int E_{\text{ff}} A(E_T, t) \text{MM}(E_T, E_R) \times F_s(E_T, t; \lambda_n) dE_T}{N'_s}$$

 signal

+
$$W_b \frac{\int E_{\text{ff}} A(E_T, t) \text{MM}(E_T, E_R) \times F_b(E_T) dE_T}{N'_b}$$

 baseline background

+
$$W_h \frac{dN_{\text{off}}}{dE_R} \times \frac{1}{T} \times \frac{1}{N'_h}$$

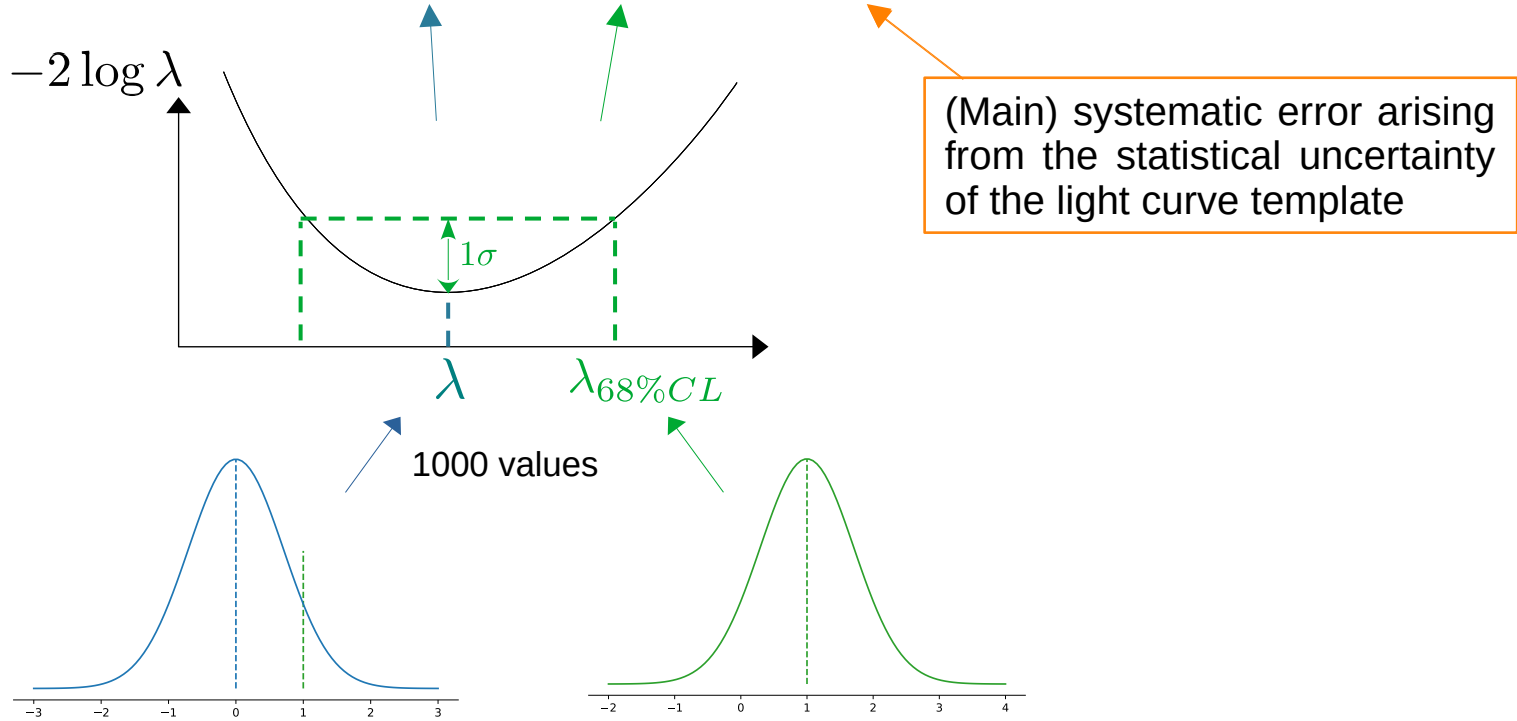
 hadronic background

from reflected region background technique applied on real data



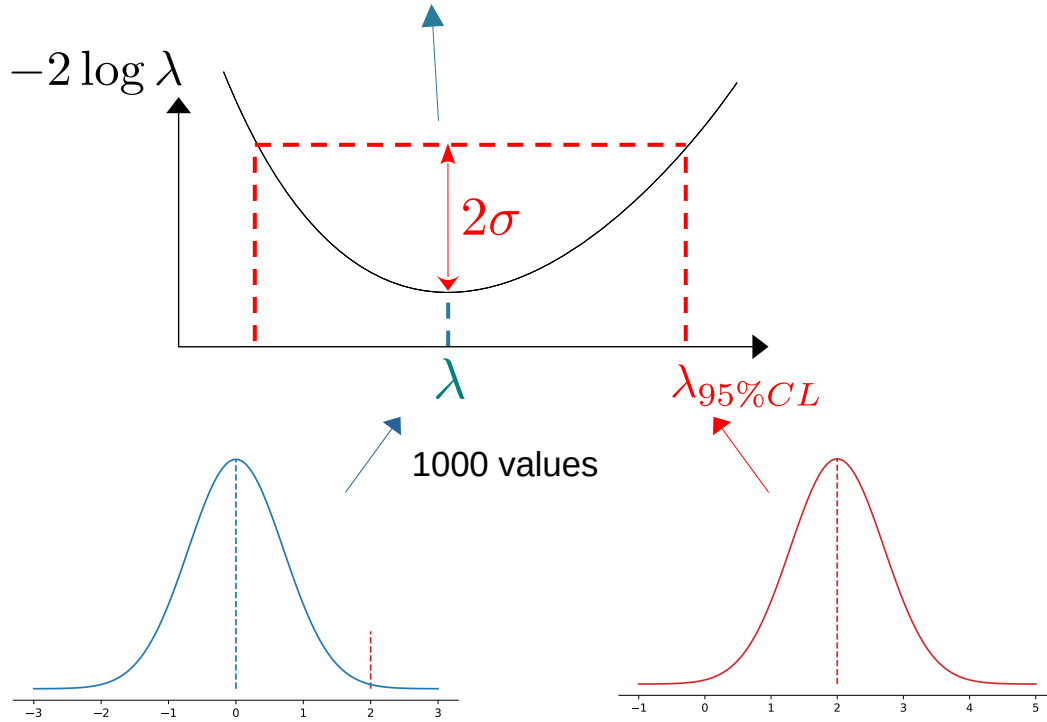
Application on real data

Time delay: $\lambda_1 = (3768 + 1475 + 3433 - 1466 - 3414) \text{ s.TeV}^{-1}$

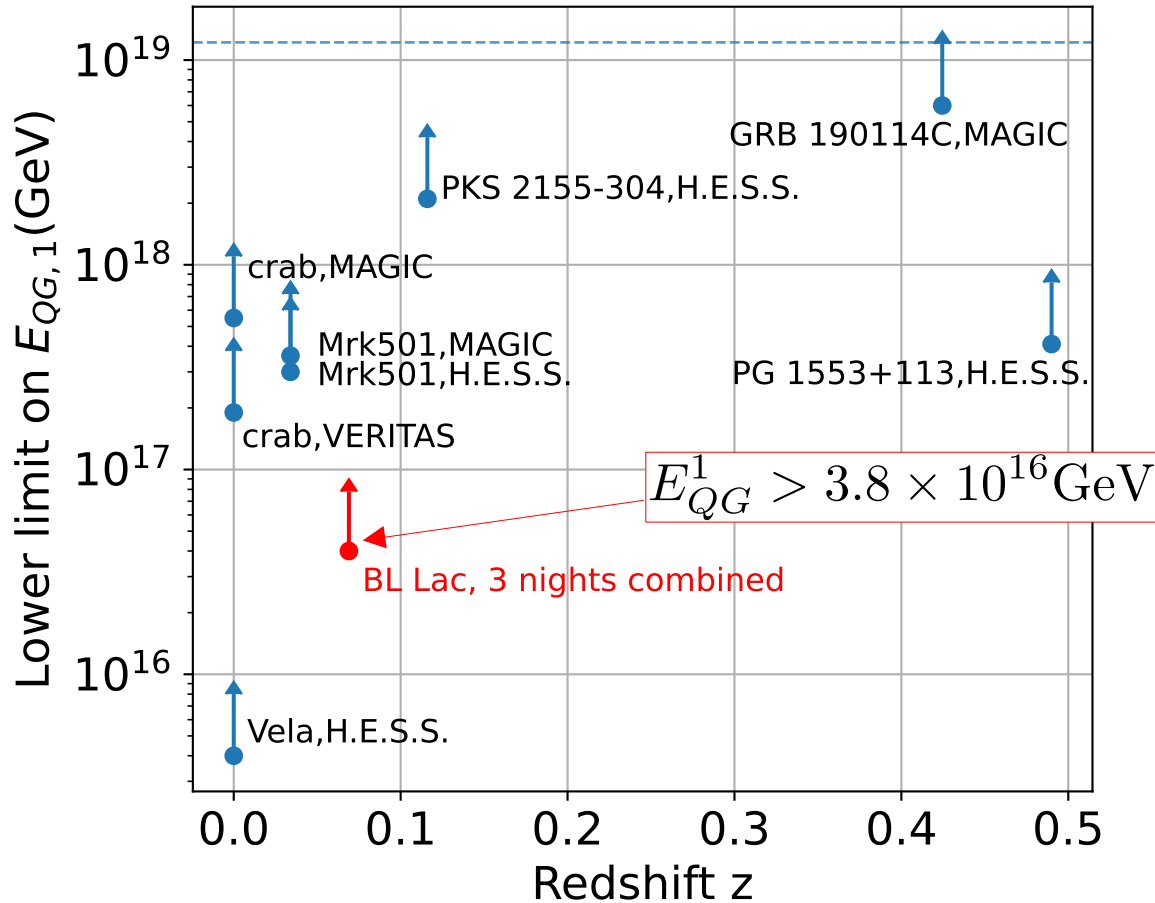


Application on real data

Time delay: $\lambda_1 = (3768 + 1475 + 3433 - 1466 - 3414) \text{ s.TeV}^{-1}$

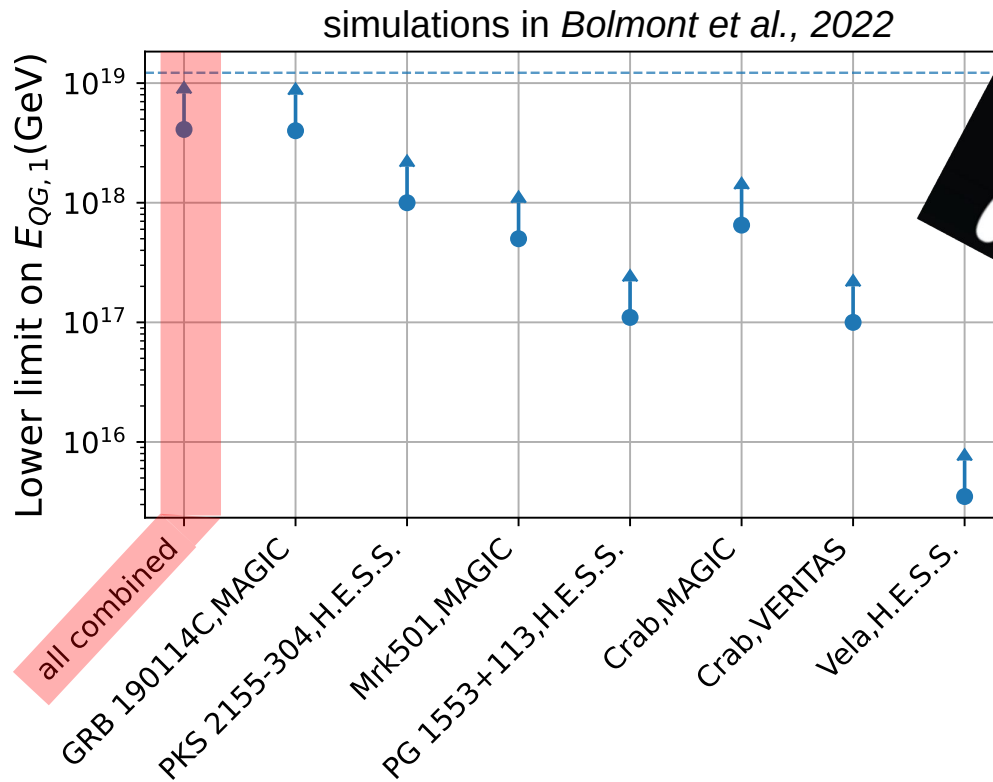


Use $\lambda_{1,95\%CL} = \pm \frac{n+1}{2H_0 E_{QG,lim}^1}$ to extract: $E_{QG}^1 > 3.8 \times 10^{16} \text{ GeV}$



Bolmont et al., 2022

$$\mathcal{L}_{\text{comb}}(\lambda_n) = \mathcal{L}_{H.E.S.S.}(\lambda_n) + \mathcal{L}_{MAGIC}(\lambda_n) + \mathcal{L}_{VERITAS}(\lambda_n)$$



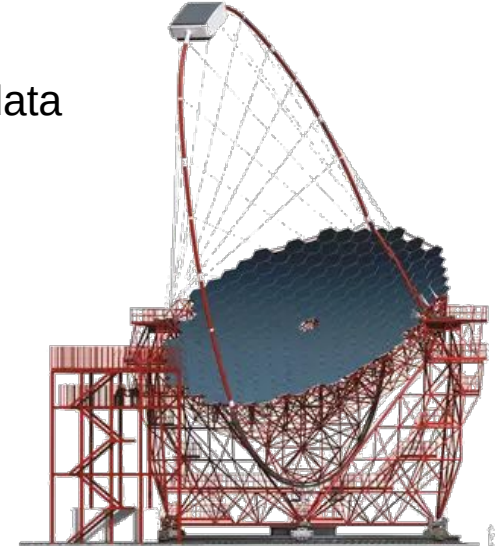
LIVelihood is ready for real data combination, including LST-1 data !

$$\mathcal{L}_{\text{comb}} = \mathcal{L}_{H.E.S.S.} + \mathcal{L}_{MAGIC} + \mathcal{L}_{VERITAS} + \mathcal{L}_{LST-1}$$

- Systematic analysis of several AGN from LST-1 data until June 2023, searching for variability in the lightcurve of a given night
- Found 3 nights showing intra-night variability (BL Lac)
- Combined these 3 nights to extract a limit on E_{QG}^1 using real data
- First analysis performing a combination of flares with Cherenkov telescope data

Ongoing work:

- Working on extended dataset and most recent data
- Implementation of all systematics (template, spectral index, background, energy scale, distance)
- $E_{QG,lim}$ for $n=2$ and various $\kappa(z)$ models (used here *Jacob & Piran, 2008*)
- **Combination of LST-1 analysis with the γ LIV WG data**



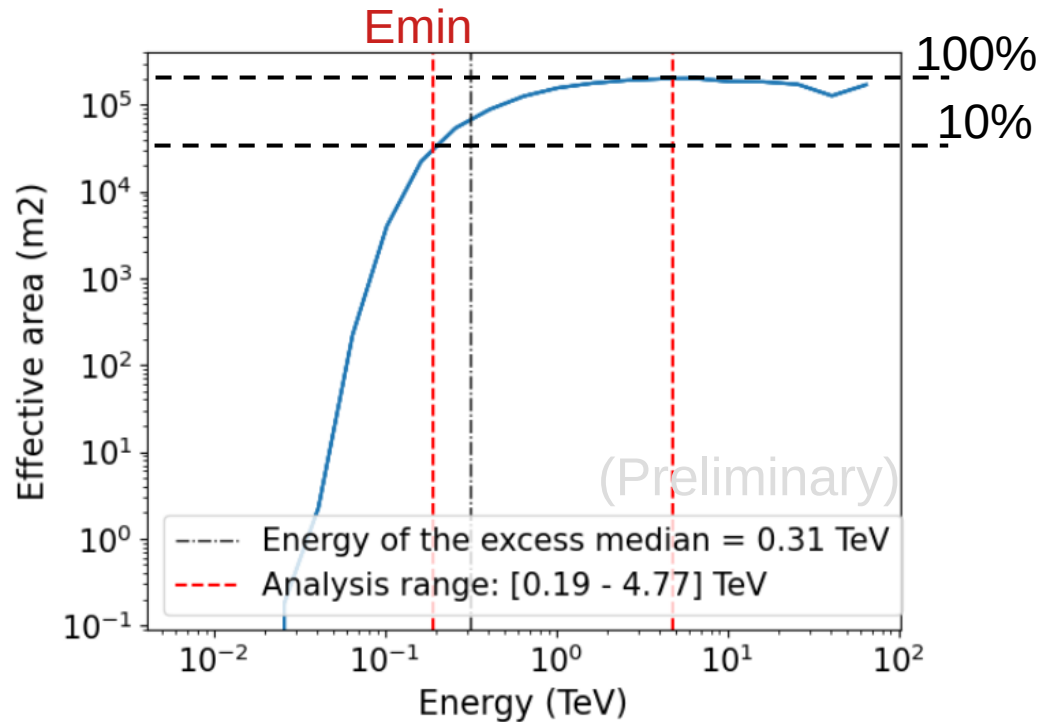


Backup

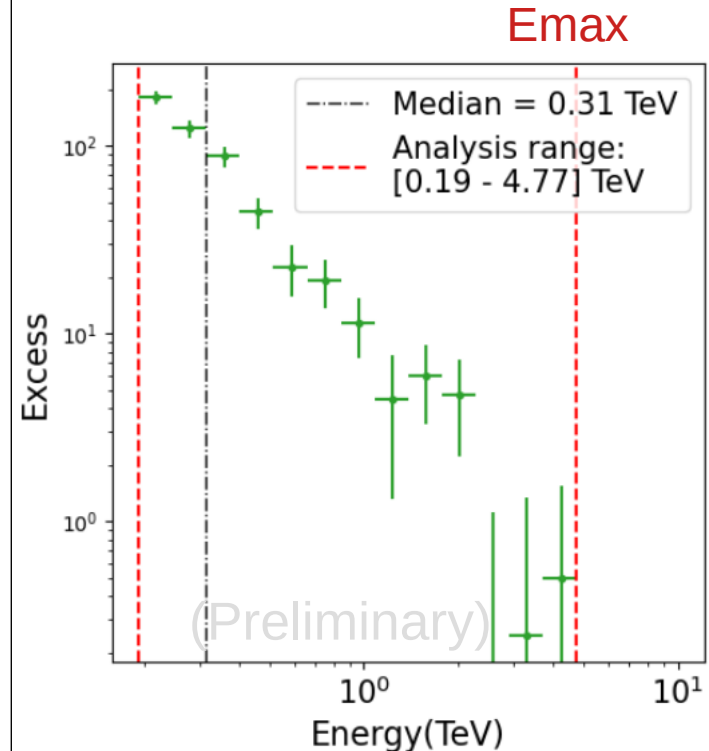
Night-wise energy ranges:

2021-08-08	2021-08-09	2022-10-20
[0.25 ; 4.8] TeV	[0.19 ; 4.8] TeV	[0.40 ; 7.7] TeV

1. Apply a safe mask threshold of 10% of effective area for each run and keep most conservative values for the whole night

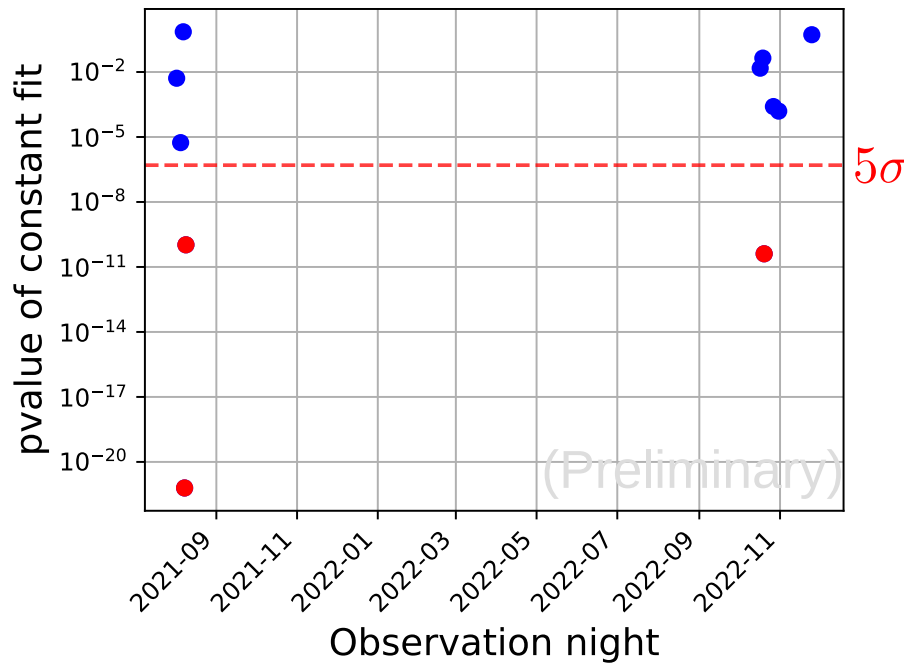


2. Keep the extrema non-zero energy bins of excess

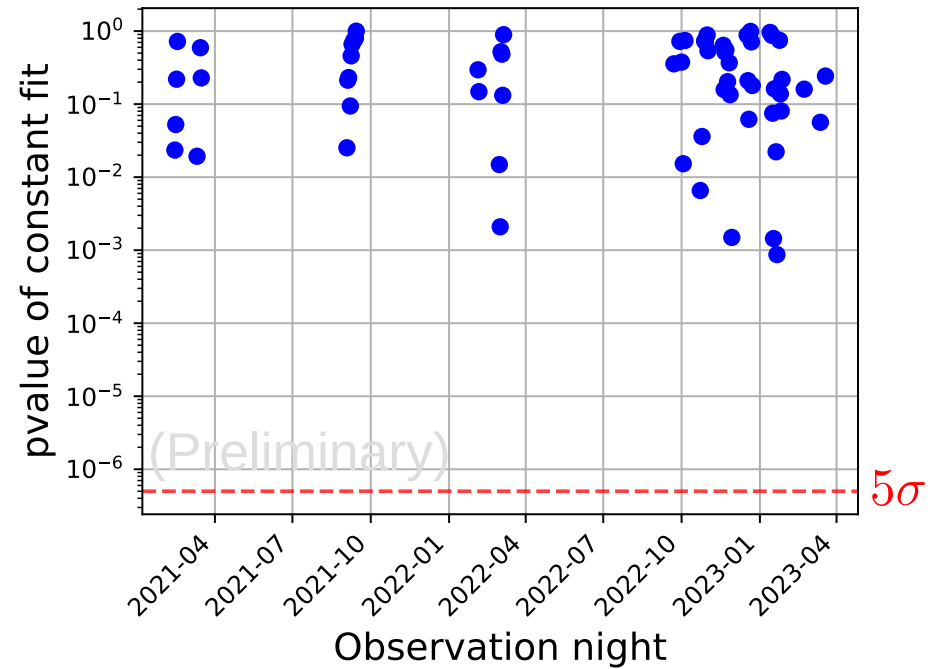


crab is a stable source \leftrightarrow expecting 0 night with intra-night variability

BL Lac

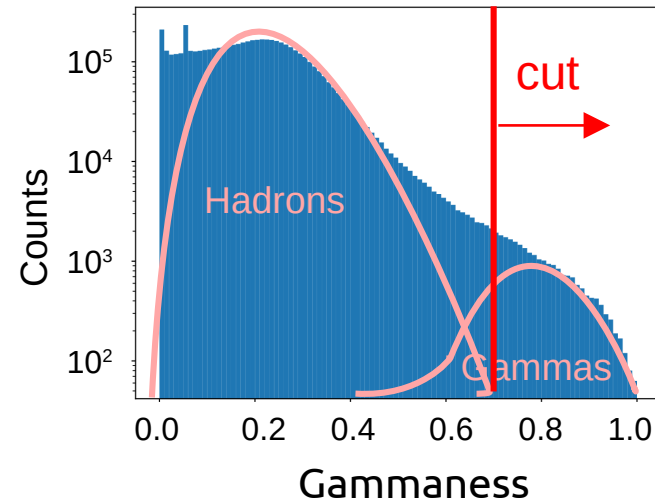
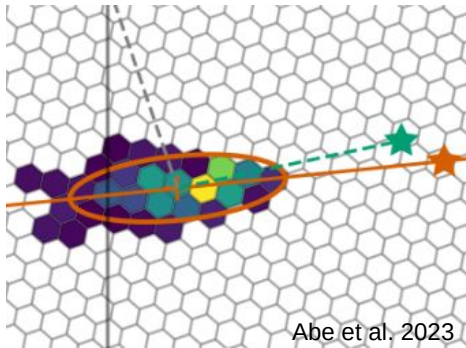


Crab



Non-variable sample

Optimization of gammaness cut



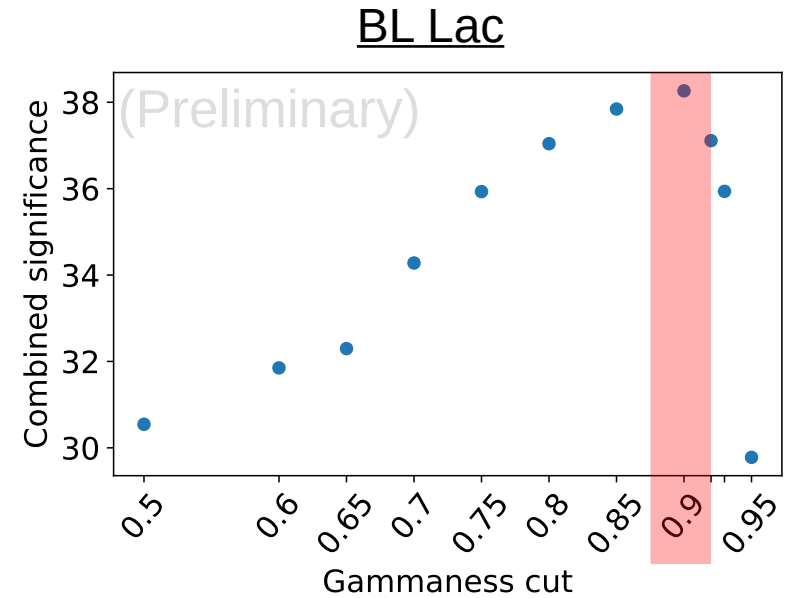
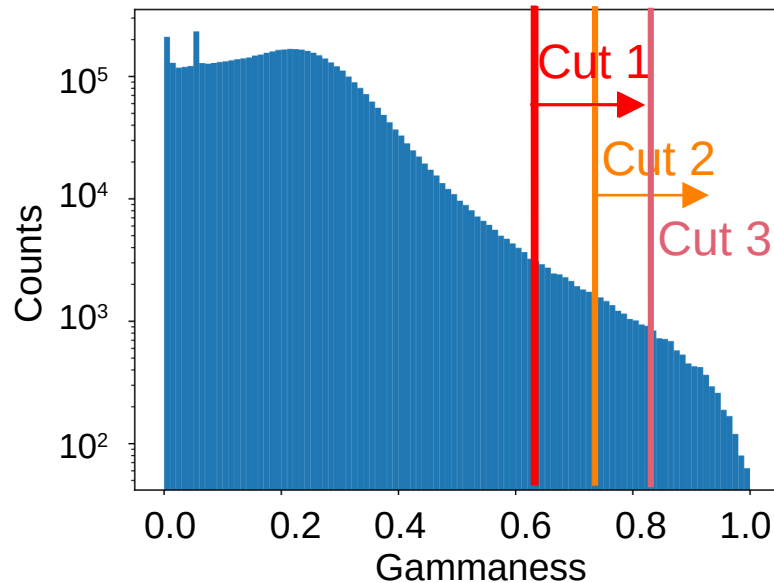
Reconstruction of event properties

- direction
- energy
- type of particle (gamma, hadron, ...)

score of how likely an event is expected to be a gamma rather than background

Non-variable sample

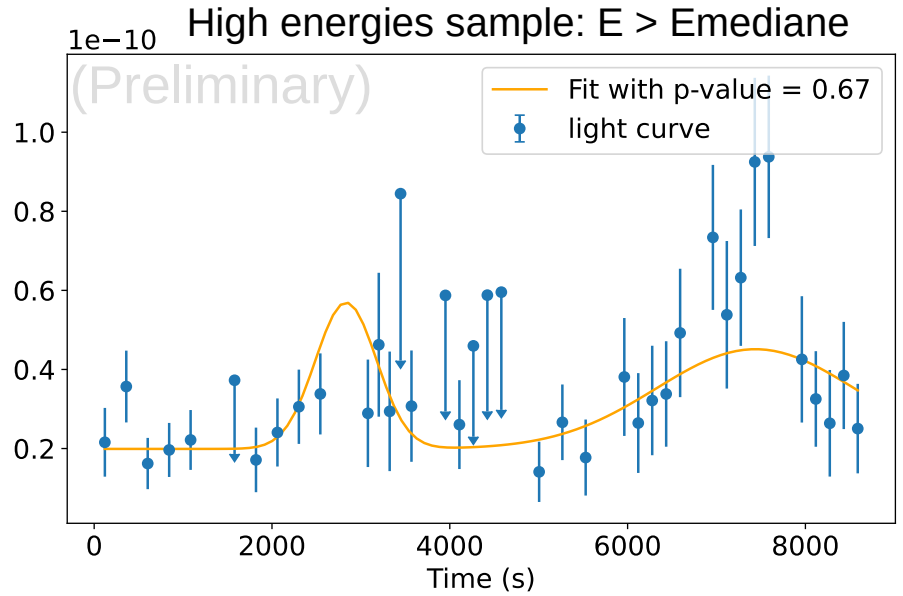
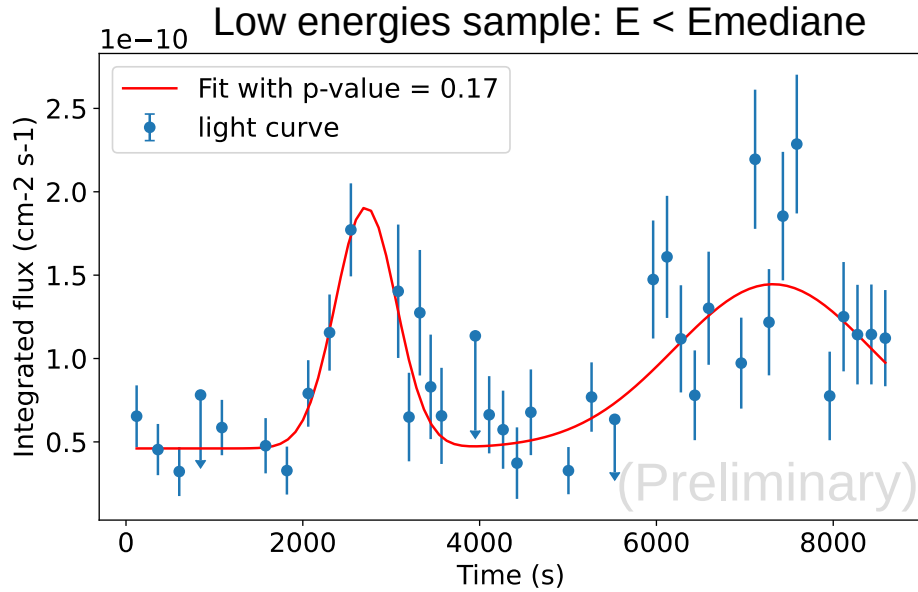
Select different gammaness cuts and search for the significance



Given a source s and its sample of non-variable nights N_s :

$$S_{s, \text{cut}} = \sqrt{\sum_{n \in N_s} S_n^2}$$

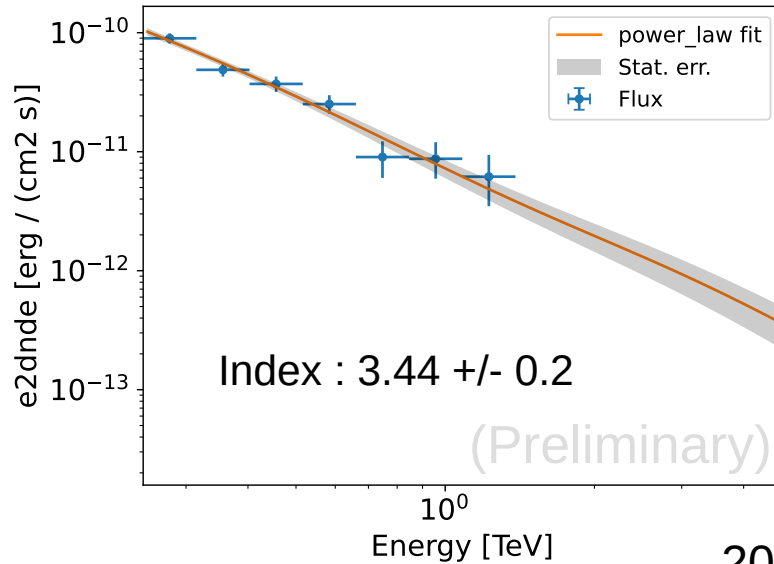
Example of 2021-08-08 night



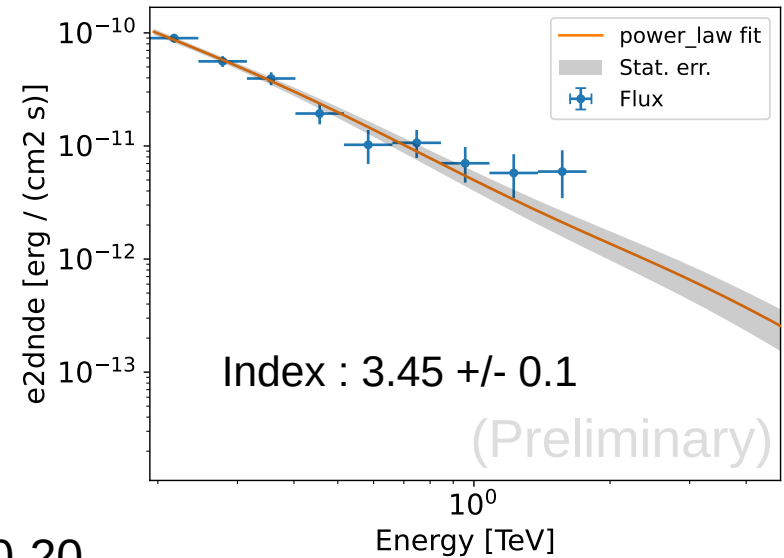
$$G_{LE}(t) = A_1 e^{-\frac{(t-\mu_1)^2}{2\sigma_1^2}} + A_2 e^{-\frac{(t-\mu_2)^2}{2\sigma_2^2}} + C_0 \quad G_{HE}(t) = (G_{LE}(t - \Delta t) - C_0) \times A + C$$

(Preliminary)	Energy range (TeV)	Median (TeV)	Delay Δt (s)	Delay significance (σ)
2021-08-08	[0.25 ; 4.8]	0.40	79±69	1.1
2021-08-09	[0.19 ; 4.8]	0.31	-136±397	0.3
2022-10-20	[0.40 ; 7.7]	0.69	-953±526	1.8

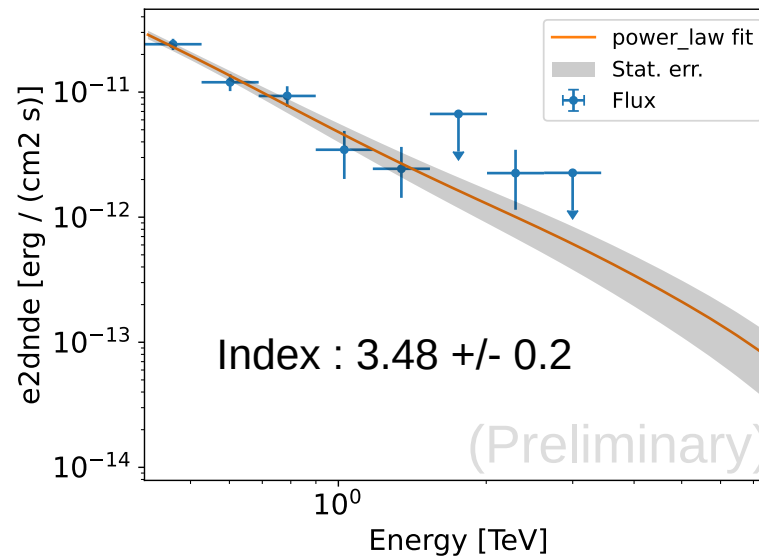
2021-08-08

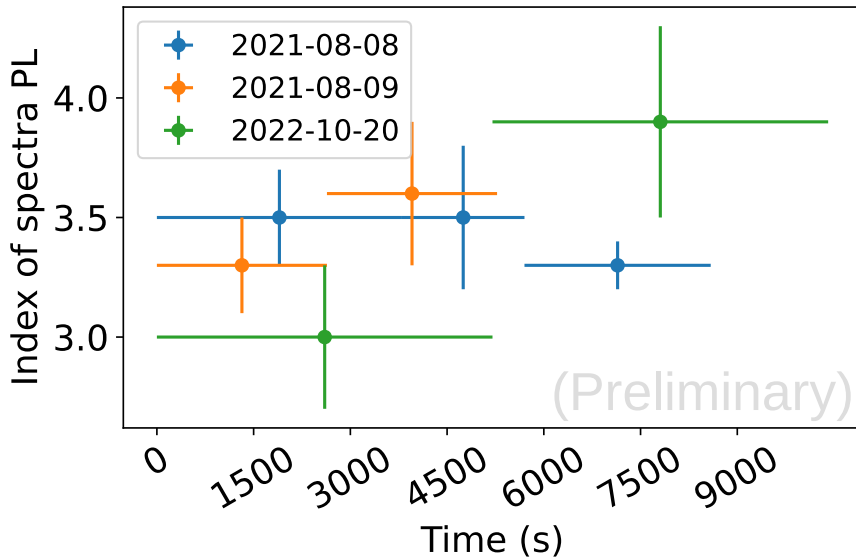
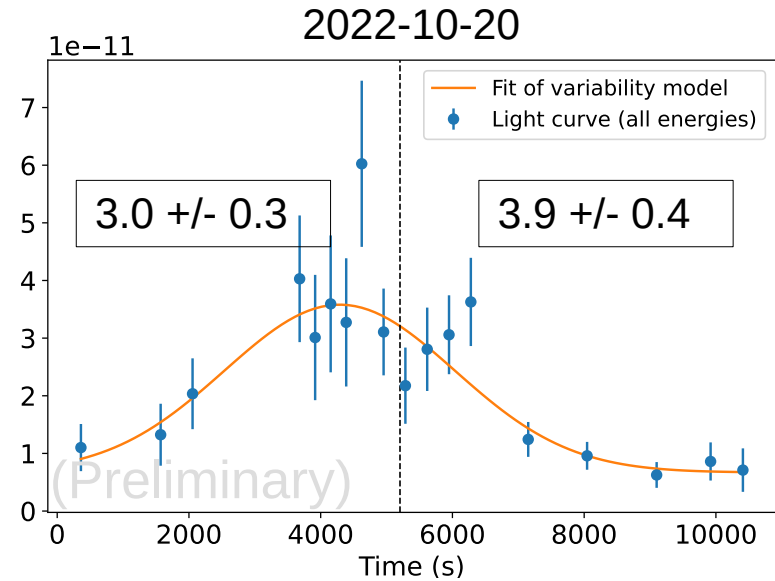
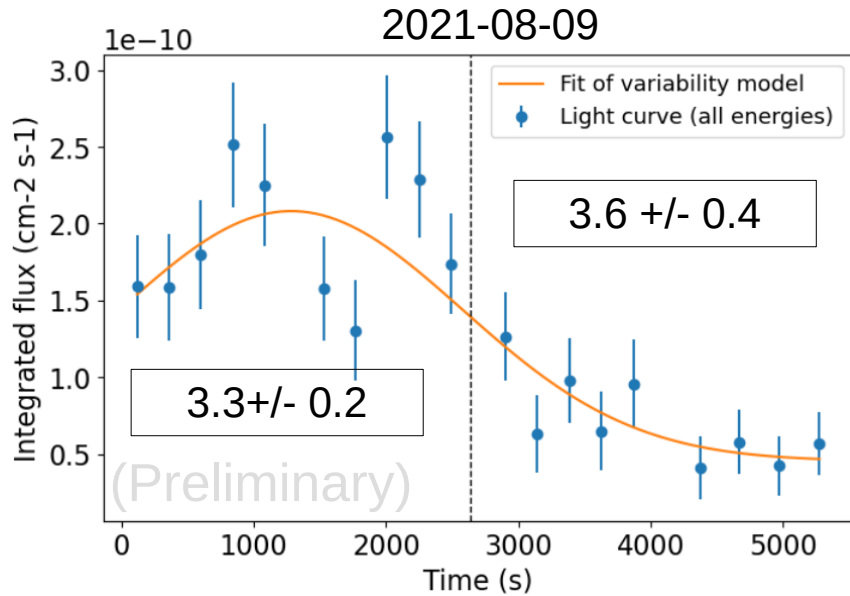


2021-08-09



2022-10-20





for 2 given time bins j and k:

$$S = \frac{|\text{index}_j - \text{index}_k|}{\sqrt{\sigma_j^2 + \sigma_k^2}}$$

	2021-08-08	2021-08-09	2022-10-20
1 st - 2 nd : 0.3σ		0.9σ	2.2σ
2 nd - 3 rd : 1.1σ			
1 st - 3 rd : 1.0σ			

(Preliminary)

$$\mathcal{L}(\lambda_n, \vec{\theta}) = \mathcal{L}_S(\lambda_n, \vec{\theta}) +$$

$$+ \mathcal{L}_{\text{template}}(\vec{\theta}_C) + \mathcal{L}_\gamma(\vec{\theta}_\gamma) + \mathcal{L}_B(\vec{\theta}_B) + \mathcal{L}_{ES}(\vec{\theta}_{ES}) + \mathcal{L}_z(\vec{\theta}_z)$$

parameters of
lightcurve analytic
parametrization

power-law index
of signal events
spectrum

$\frac{\text{signal}}{\text{total events}}$ & $\frac{\text{hadrons}}{\text{total background}}$

energy
scale

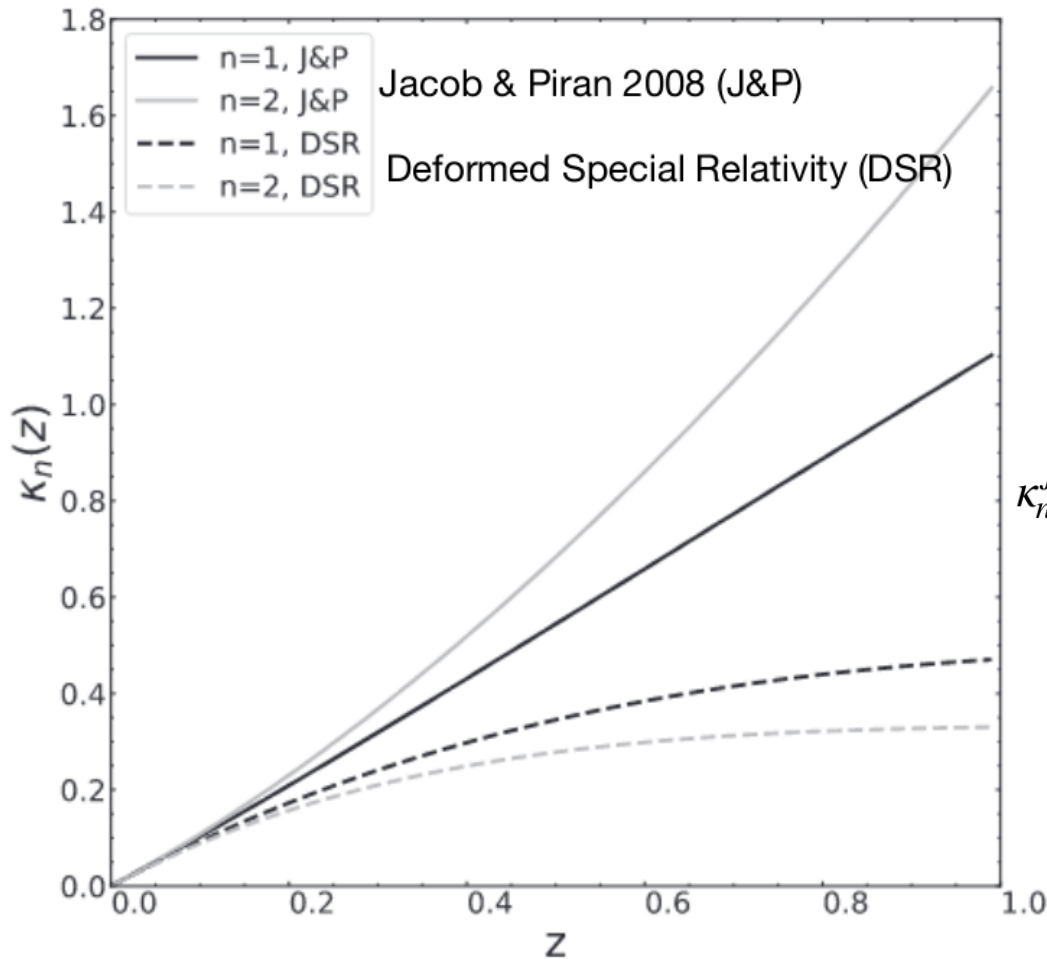
distance
(redshift)

dominant

(Bolmont et al. 2022)

**can increase
when n=2**

~ 0

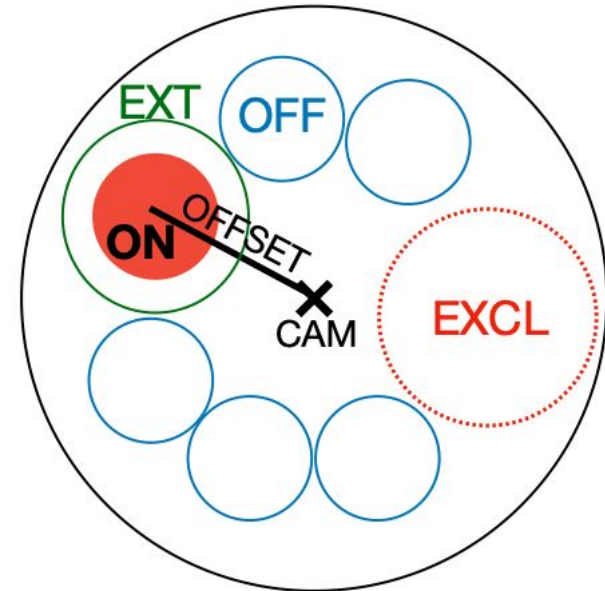


Bolmont et al 2022

$$\kappa_n^{J\&P}(z) = \frac{1}{z_0} \int_0^z \frac{(1+z')^n}{\sqrt{\Omega_m(1+z')^3 + \Omega_\Lambda}} dz'$$

Hypothesis: radial symmetry of background in the field-of-view

- X CAM: camera pointing direction
- OFFSET: regions dispersion radius
- ON: source (gammas) + background
- EXT: exclusion of potential remaining source events
- EXCL: exclusion of a potential other source
- OFF: background



$$N_{\gamma} = N_{excess} = N_{on} - \frac{1}{n} \sum_n N_{n,off}$$