Mapping the Cosmic Evolution of Fermi Blazars

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Counting Experiment

What are we counting?

- **Blazars -** Jet points along our line of sight.
- Two types- FSRQs and BL Lacs.

How does counting give us information about evolution?

● Counting the number of objects per unit comoving volume and luminosity interval gives us the Luminosity Function (LF).

Dataset

2497 Sources from Fermi LAT 8-year Source List (Marcotulli 2020).

Larger than previous samples (Ajello+14/15 ~400). Would lead to remarkable improvement in LF determination.

100 MeV-1 TeV range. Extragalactic sources.

Luminosity Function at z=0

Best-fit LF has double PL shape.

Spans 4 orders of magnitude lower than previous studies.

Space Density of Blazars

Change in curvature of density plot can hint towards the link between FSRQs and

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[Cavalie](https://ui.adsabs.harvard.edu/search/q=author:%22Cavaliere%2C+A.%22&sort=date%20desc,%20bibcode%20desc)re [& D'E](https://ui.adsabs.harvard.edu/search/q=author:%22D)lia 2002 [Böttch](https://ui.adsabs.harvard.edu/search/q=author:%22B%C3%B6ttcher%2C+M.%22&sort=date%20desc,%20bibcode%20desc)er [& Derm](https://ui.adsabs.harvard.edu/search/q=author:%22Dermer%2C+C.+D.%22&sort=date%20desc,%20bibcode%20desc)er 2002

Hint towards different peaks of FSRQs and BL Lacs.

Thank You :)

Luminosity Histogram

PDE Parameters

L_star = $1.630855e-03 * 1e48 = 1.63e45 erg/s$ $k = 15.98$ $\chi i = -0.117$ gamma1 = 1.23 gamma2 =2.05 $\tau = 3.325$

$$
\begin{split} \Phi(L_{\gamma},\varepsilon=0,\Gamma)=\frac{A}{\ln(10)L_{\gamma}}\Bigg[\Bigg(\frac{L_{\gamma}}{L_{\alpha}}\Bigg)^n\\ +\Bigg(\frac{L_{\gamma}}{L_{\alpha}}\Bigg)^{1/2}\Bigg]^{-1}\cdot e^{-\frac{L_{\gamma}(\gamma)}{2\sigma^2}}. \end{split}
$$

$$
e(z) = (1+z)^{k_d} e^{z/\xi},
$$

\n
$$
k_d = k^* + \tau \times (\log_{10}(L_\gamma) - 46).
$$

$$
\Phi(L_{\gamma}, z, \Gamma) = \Phi(L_{\gamma}, z = 0, \Gamma) \times e(z),
$$

Luminosity Function (LF)

LF (ϕ) informs us on the number of objects per unit comoving volume and luminosity interval.

We can get other quantities of interest from LF:

Number density Luminosity density $n = \int \Phi(L) dL$ $\int \Phi(L) L dL$

https://jila.colorado.edu

Redshift of BLL

- ❏ Photometric dropout technique (Lower or Higher limit).
- ❏ Metal line absorption systems (Lower limit).
- ❏ Host Galaxy Spectral Fitting (Lower limit) assuming M_R of elliptical galaxy to be same for all cases, host non-detection places lower limit on z.

Create PDF of *z* for each source. Compute LF using *z* from PDF. Use LF to predict observe *dN/dz*. Different? → replace with computed *dN/dz* and iterate to converge.

Redshift Distribution

Method

Maximum-Likelihood (ML) Method used to derive LF (Φ) .

Number density of BLL as function of L , z and Γ :

$$
\frac{\partial^3 N}{\partial L_{\gamma} \partial z \partial \Gamma} = \frac{\partial^3 N}{\partial L_{\gamma} \partial V \partial \Gamma} \times \frac{dV}{dz} = \Phi(L_{\gamma}, V(z), \Gamma) \times \frac{dV}{dz}
$$

The best-fit LF is found by comparing, through an ML estimator, the number of expected objects (for a given model LF) to the observed number while accounting for selection effects.

L-z plane parsed into tiny intervals *dLdz*. Expected number of blazars for each interval :

 $\lambda(L_{\gamma}, z, \Gamma) dL_{\gamma} dz d\Gamma = \Phi(L_{\gamma}, V(z), \Gamma) \cdot \Omega(L_{\gamma}, z, \Gamma)$ $\times \frac{dV}{dt} dL_{\gamma} dz d\Gamma,$

Intervals will contain either 0 or 1 BLL.

Maximum-Likelihood (ML) Method

We know that Poisson probability is: $f(x) = \frac{\lambda}{x} e^{-\lambda}$, $x = \text{no. of successes.}$

$$
f(x) = \frac{\lambda^x}{x!} e^{-\lambda}
$$

For our case, $x = 0$ or 1.

The Likelihood function is given by the product of probabilities, such that for each *L-z* bin there is exactly 1 BLL and 0 otherwise:

$$
L = \prod_i \lambda(L_{y,i}, z_i, \Gamma_i) dL_y d\zeta d\Gamma e^{-\lambda(L_{y,i} z_i, \Gamma_i)d\zeta_i d\tau d}
$$

$$
\times \prod_i e^{-i\pi L_{y,i} z_i, \Gamma_i)d\zeta_i d\tau d\zeta_i
$$

Function to minimize:

$$
\label{eq:action} \begin{split} S = & -2\sum_{i}\ln\frac{\partial^2 N}{\partial L_y\partial\partial\dot{X}} + 2\int_{\Gamma_{\rm{min}}}^{\Gamma_{\rm{min}}}\int_{L_{\rm{max}}}^{L_{\rm{max}}}\\ &\times\int_{L_{\rm{min}}}^{\Gamma_{\rm{max}}}\lambda(L_y,\Gamma,z) dL_ydz d\Gamma. \end{split}
$$

Test of LF Model

Comparison of Data and Predictions:

 $\frac{dN}{dz} = \int_{\Gamma_{\rm min}}^{\Gamma_{\rm max}} \int_{L_{\gamma,{\rm min}}}^{L_{\gamma,{\rm max}}} \lambda(L_\gamma,\Gamma,z) dL_\gamma d\Gamma,$

$$
N(>F)=\int_{\Gamma_{\min}}^{\Gamma_{\max}}\int_{z_{\min}}^{z_{\max}}\int_{L_{\gamma}(z,F)}^{L_{\gamma,\max}}\Phi(L_{\gamma},V(z),\Gamma)\frac{dV}{dz}d\Gamma dzdL_{\gamma},
$$

 $\begin{split} \frac{dN}{d\Omega_T} &= \int_{\Gamma_{\rm min}}^{\Gamma_{\rm max}} \int_{\Gamma_{\rm min}}^{\Gamma_{\rm max}} \lambda(L_y, \Gamma, z) dz d\Gamma, \\ \frac{dN}{d\Gamma} &= \int_{L_{\rm max}}^{L_{\rm max}} \int_{\Gamma_{\rm min}}^{\Gamma_{\rm max}} \lambda(L_y, \Gamma, z) dL_y dz. \end{split}$

Nobs/Nmodel Method:

$$
\Phi(L_{\gamma,i}, V(z_i), \Gamma_i) = \Phi^{\text{mdl}}(L_{\gamma,i}, V(z_i), \Gamma_i) \frac{N_i^{\text{obs}}}{N_i^{\text{mdl}}},
$$

LF Model

Photon Index \rightarrow Gaussian distribution.

Local LF (DPL):

$$
\begin{aligned} \Phi(L_\gamma,\varepsilon=0,\Gamma)=\frac{A}{\ln(10)L_\gamma}\bigg[\bigg(\frac{L_\gamma}{L_\gamma}\bigg)^n\\ +\bigg(\frac{L_\gamma}{L_\gamma}\bigg)^{12}\bigg]^{-1}\cdot e^{-\frac{L_\gamma(\varepsilon)}{2\gamma^2}} \end{aligned}
$$

Low luminosity - flatter, high lum - steeper slope.

Low lum objects are more common than high lum. Turnover luminosity is useful when comparing LFs for different *z* or *L*.

PLE & PDE:

$$
e(z) = (1 + z)^{k_d} e^{z/\xi},
$$

\n
$$
k_d = k^* + \tau \times (\log_{10}(L_\gamma) - 46).
$$

For the PDE the evolution is defined as

$$
\Phi(L_{\gamma}, z, \Gamma) = \Phi(L_{\gamma}, z = 0, \Gamma) \times e(z), \tag{15}
$$

while for the PLE case it is

$$
\Phi(L_{\gamma}, z, \Gamma) = \Phi(L_{\gamma}/e(z), \Gamma). \tag{16}
$$

The PLE and PDE models have 10 free parameters (A, γ_1) , $L_*, \gamma_2, k^*, \tau, \xi, \mu^*, \beta,$ and σ).

LDDE

$$
\Phi(L_{\gamma}, z, \Gamma) = \Phi(L_{\gamma}, z = 0, \Gamma) \times e(z, L_{\gamma}), \qquad (17)
$$

where

$$
e(z, L_{\gamma}) = \left[\left(\frac{1+z}{1+z_c(L_{\gamma})} \right)^{p1(L_{\gamma})} + \left(\frac{1+z}{1+z_c(L_{\gamma})} \right)^{p2} \right]^{-1} \tag{18}
$$

$$
z_c(L_\gamma) = z_c^* \cdot (L_\gamma / 10^{48})^\alpha, \tag{19}
$$

$$
p1(L_{\gamma}) = p1^* + \tau \times (\text{Log}_{10}(L_{\gamma}) - 46). \tag{20}
$$

Methodology

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Fit various LF models on the data (PLE, PDE, LDDE).

Local Luminosity Function (z=0):

$$
\phi(L_{\gamma}, z=0) = \frac{dN}{dL_{\gamma}dV} = \frac{A}{ln(10)L_{\gamma}} \left[\left(\frac{dL_{\gamma}}{dL_{*}} \right)^{\gamma_{1}} + \left(\frac{dL_{\gamma}}{dL_{*}} \right)^{\gamma_{2}} \right]^{-1}
$$

Including Evolution (z dependence), e.g. PDE:

$$
\Phi(L_\gamma,z) \ \ = \ \ \Phi(L_\gamma,z=0) \times e(z,L_\gamma)
$$

Evolutionary Term:

$$
e(z,L_{\gamma})=(1+z)^{k_d}e^{z/\xi}
$$

$$
k_d = k^* + \tau \times [log(L_\gamma) - 46]
$$

The V/V_max Test

$$
\left\langle \frac{V}{V_{\text{max}}} \right\rangle = \frac{1}{n} \cdot \sum_{i=1}^{n} \frac{V_i}{V_{\text{max}}}
$$

Value within [0,1]

Value= $0.5 \rightarrow$ equally distributed. Value>0.5 \rightarrow positive evolution.

Would introduce bias if there is evolution within *z* bins.

$$
V_{\text{MAX}} = \int_{z_{\text{min}}}^{z_{\text{max}}} \Omega(L_i, z, \Gamma) \frac{e(z, L_i)}{e(z_{\text{min}}, L_i)} \frac{dV}{dz} dz,
$$

Algorithm

- ❏ Prior function for *dN/dz.*
- ❏ Simulate 1000 samples, with z from PDF.
- ❏ Use ML to derive best-fit LF for each sample. Final LF → average of 1000 LF.

This enables us to derive the uncertainty as well.

❏ Compare *dN/dz*, if different: use latest *dN/dz*. Repeat till both match.

BLL, FSRQ, BCU

Optical classification - different resources, in decreasing order of precedence: **optical spectra** from our intensive follow-up program (Shaw et [al. 20](https://iopscience.iop.org/article/10.1088/0004-637X/810/1/14)13), the BZCAT list (i.e., classification from this list, which is a compilation of sources ever classified as blazars, Massaro et [al. 20](https://iopscience.iop.org/article/10.1088/0004-637X/810/1/14)09), and **spectra available in the literature**, e.g., SDSS (Ahn et [al. 20](https://iopscience.iop.org/article/10.1088/0004-637X/810/1/14)12), 6dF (Jones et [al. 20](https://iopscience.iop.org/article/10.1088/0004-637X/810/1/14)09).

The *BCU***—blazar candidates of uncertain type** , from BZCAT/ radio data/ double-humped SED. BCUs are divided into three sub-types:

BCU I: the counterpart has a published optical spectrum but is not sensitive enough for a classification as an FSRQ or a BL Lac;

BCU II: the counterpart is lacking an optical spectrum but a reliable evaluation of the SED synchrotron-peak position is possible;

BCU III: the counterpart is lacking both an optical spectrum and an estimated synchrotron-peak position but shows blazar-like broadband emission and a flat radio spectrum;

3LAC: 10.1088/0004-637X/810/1/14

Fermi LAT Biases

1. Spectral bias (or photon-index bias). It is the selection effect which allows Fermi-LAT to detect spectrally hard sources at fluxes generally fainter than those of soft sources.

2. Malmquist bias: in [a brightnes](https://en.wikipedia.org/wiki/Brightness)s-limite[d surve](https://en.wikipedia.org/wiki/Astronomical_survey)y, where stars below a certain apparent brightness cannot be included. Since observ[ed star](https://en.wikipedia.org/wiki/Stars)s a[nd galaxie](https://en.wikipedia.org/wiki/Galaxies)s appear dimmer when farther away, the brightness that is measured will fall off with distance until their brightness falls below the observational threshold.

3. Eddington bias: The flux (F) from astrophysical sources has a fluctuation of ΔF . If a source falls closely to the detection threshold of the instrument, it would be more easily detected if $F = F + \Delta F$. Therefore, such objects (~3% of the sample) are found with a higher flux than their intrinsic one.

Detection Efficiency $(\boldsymbol{\omega})$

Marcotulli et al., 2020.

Efficiency Correction Method

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Algorithm:

- Point sources from LAT data Real catalog.
- 2) Monte Carlo simulations to generate isotropic blazar distribution.

Inputs \rightarrow spectral and flux parameters from real blazars.

- 1) Detect point sources from simulated sky.
- 2) Derive detection efficiency (*ω(S)*).
- 3) Using $\omega(S)$ to correct the real catalog \rightarrow account for biases.
- 4) Derive the intrinsic source count distribution (*logN-logS*).

Efficiency Correction Method

Intrinsic real sky not known. Running detection pipeline for observed real sky.

Simulating real sky and passing it through the detection pipeline to obtain detection efficiency.

Simulated

Sky

= Bright = Faint

Using the detection efficiency to obtain the intrinsic real sky.

Monte Carlo Simulations

To u[se randomne](https://en.wikipedia.org/wiki/Randomness)ss to solve problems that might [be determinis](https://en.wikipedia.org/wiki/Deterministic_system)tic in principle.

Monte Carlo methods vary, but tend to follow a particular pattern:

- 1. Define a domain of possible inputs
- 2. Generate inputs randomly fro[m a probability distributi](https://en.wikipedia.org/wiki/Probability_distribution)on over the domain
- 3. Perfor[m a determinis](https://en.wikipedia.org/wiki/Deterministic_algorithm)tic computation on the inputs
- 4. Aggregate the results

Markov Chains: At the core of MCMC is the concept of Markov chains. A Markov chain is a process where the probability of transitioning from one state to another depends only on the current state, not on the sequence of states that preceded it. In simpler terms, it's a random process where the future state depends only on the present state, not on the past. One of the critical aspects of MCMC methods is ensuring convergence, meaning that the sequence of sampled states eventually settles into the true distribution.

Diffused EGB

There are two possibilities for the origin of the diffused extragalactic EGB:

(1) it is truly diffuse; and

(2) it is the integrated emission of various distant unresolved gamma-ray sources.

It may also be a combination of diffuse and point sources and may have different origins in different portions of the gamma-ray band. From below 10 MeV to 100 MeV, particle-antiparticle annihilation, bremsstrahlung and inverse Compton interactions between cosmic ray particles and lower-energy photons are the most likely gamma-ray production mechanisms.

Above 100 MeV, the dominant process is decay from nucleon interactions. E.g. A cosmic-ray proton strikes another proton. The protons survive the collision, but their interaction creates an unstable particle — a pion — with only 14 percent the mass of a proton. In 10 millionths of a billionth of a second, the pion decays into a pair of gamma-ray photons.

Diffused process also include Dark matter annihilation.

Blazars contribute to the unresolved isotropic diffused EGB.

Extragalactic Gamma-ray Background

Discussions and Conclusions

❏ Contribution to EGB.

$$
F_{000}(E_1) = \int_{\frac{E_1 - 10^4}{E_1 - 10^4}}^{\frac{E_1 - 10^4}{E_1 - 10^4}} \int_{\frac{10^4}{(1 - 10^4)^4}}^{\frac{10^4}{(1 - 10^4)^4}} dz
$$

×
$$
\int_{\frac{E_1 - 10^4}{E_1 - 10^4}}^{\frac{10^4}{(1 - 10^4)^4}} dL_1 \cdot \Phi(L_2, \xi, \Gamma) \cdot \frac{dK_1}{dE} \cdot \frac{dV}{dz}
$$

× (ph cm⁻³ s⁻¹ sr⁻¹ GeV⁻³),

Correcting the Real Sky

$$
\frac{dN}{dS} = \frac{1}{\Omega \Delta S_i} \frac{N_i}{\omega(S_i)},
$$

For efficiency < 1, in the denominator, there is an increase in the number. This accounts for low flux sources that were missed out in the uniform survey.

Total blazer contribution to EGB = 50 %. 27% of this 50% from unresolved sources.

Resolved sources Unresolved sources

$$
S_{\text{EGB}} = \sum_{i=1}^{N} \frac{S_{\text{PS},i}}{\Omega} + \int_{S_{\text{min}}}^{S_{\text{max}}} (1 - \omega(S')) S' \frac{dN}{dS'} dS'
$$

Importance of Redshift Constraints

Using only spectroscopic z introduces bias in the. *dN/dz*, and therefore the LF computed utilizing it.

More BLL at lower z.

Electromagnetic Spectrum

$1eV \rightarrow 10^4K$

Blazar Model

Park et al 2022

Synchrotron Emission + Beaming

$$
E_{rad} = \frac{2}{Re^{2}} \{ \pi x (n \times \vec{u}) \}
$$
\n
$$
|E_{rad}| = q \vec{u} \sin \theta
$$
\n
$$
S = \frac{c}{4\pi} E_{rad}^{2} = \frac{c}{4\pi} \frac{q^{2} \vec{u}^{2} \sin^{2} \theta}{4\pi R^{2} \epsilon^{4}}
$$
\n
$$
P = \frac{dW}{dt} = \frac{q^{2} \vec{v}^{2}}{4\pi c^{3}} \sin^{2} \theta d\theta
$$
\n
$$
= \frac{2q^{2} \vec{u}^{2}}{3c^{3}}
$$
\n
$$
a_{1} = \omega_{B}v_{+} = \frac{q}{2\pi} \vec{v}_{+} \times \vec{e}
$$
\n
$$
\Rightarrow \omega_{B} = \frac{q_{B}}{3mc}
$$

P =
$$
\frac{2a^{2}}{3e^{3}}
$$
 Y A $(a^{2}+b^{2}a^{2})$
\nL = P = $\frac{4}{3}5 \cdot c^{32} \cdot r^{2}v_{a}$
\nN CDJE = $ce^{-1}dE$
\nR = P x $w = x^{2}$
\nR = P²
\

Accretion Disk

Turbulent motions in the disk generate stresses that transport angular momentum outward.

The alpha-disk model provides a convenient way to parameterize the effects of turbulence.

$$
T \propto r^{-3/4}
$$

\n
$$
T = \frac{G_{\text{min}}}{2r} = 2\pi r^2 \sigma T^4
$$

\n
$$
T = \frac{G_{\text{min}}}{2r} = 2\pi r^2 \sigma T^4
$$

\n
$$
V = \alpha C_s h
$$

\n
$$
V = \frac{3G_{\text{min}}}{8\pi \sigma r^3} \left(1 - \frac{R_{\text{min}}}{r}\right)^{1/4}
$$

\n
$$
B_Y = \frac{2h\nu^3}{c^2} = \frac{1}{e^{h_Y/l} - 1}
$$

\n
$$
k = \frac{n \pi}{L} \Rightarrow dr_x = \frac{k \times L_x}{2\pi}
$$

\n
$$
dN = 2 \times \frac{V d^3 k}{(2\pi)^3} \qquad (3D)
$$

\n
$$
C = \gamma \lambda \Rightarrow d\mu = 2\pi d\nu
$$

\n
$$
dN = \frac{2V \nu^2}{c^3} \qquad d^2\nu
$$

\n
$$
dN = \frac{h\nu}{c^3} \qquad \frac{2\nu^2}{c^3}
$$

 $uv = Lv = \frac{By}{c}$

Inverse Compton Scattering

$$
\Delta \lambda = \frac{h}{mc} (1-cos \theta)
$$

\n
$$
c_{f} = \frac{c_{i}}{1 + \frac{e_{i}}{mc^{2}} (1-cos \theta)}
$$

\n
$$
\frac{\partial h}{\partial r} \times \text{cm}e^{2}
$$

\n
$$
h\tilde{v} = \gamma^{2}h\nu
$$

\n
$$
\frac{\alpha}{3} \sigma_{T} c \gamma^{2} \beta^{2} U_{ph}
$$

Relativistic Doppler Effect

Absorption + Doppler Broadening

$$
\gamma_{\nu}(s) = \int_{S_{0}}^{S} \alpha_{\nu}(s') ds'
$$

\n
$$
T_{\nu}(r_{\nu}) = T_{\nu}(0) e^{-r_{\nu}}
$$

\n
$$
+ \int_{0}^{r_{\nu}} \frac{e^{-(r_{\nu} - r_{\nu})} S_{\nu}(r_{\nu}) d r_{\nu}}{r_{\nu}}
$$

$$
\frac{\gamma - \gamma_o}{\gamma_e} = \frac{v_e}{c}
$$
\n
$$
\phi(y) = \frac{1}{\Delta v_b \sqrt{\pi}} e^{-\frac{(\gamma - \gamma_o)^2}{\Delta v_b^2}}
$$
\n
$$
\Delta v_b = \frac{\gamma_o}{c} \sqrt{\frac{2kT}{\pi}}
$$

Superluminal Motion in Jets

Dataset

2497 Sources from LAT 8-year Source List (Marcotulli 2020).

100 MeV-1 TeV range. |b|>20°.

Larger than any samples used to date for the determination of the blazar LF, broader extent of redshift. Would lead to remarkable improvement in LF determination

Marcotulli, L., Di Mauro, M. and Ajello, M., 2020. ApJ *896*(1), p.6.

Understanding Evolution from Count Dataset

We can see the trend of enrollment for 4 different years.

On adding the total enrollment, we can see how the enrollment has changed over the years.

Evolution

Evolution: Shift of peak.

Major evolution at z~1.1

Luminosity Function of Blazars

- ❏ LF encodes evolution. Blazar LF is not well constrained.
- ❏ Blazars are good cosmological probes owing to their high luminosities and large redshifts.
- ❏ A major topic of debate is whether FSRQs and BL Lacs have the same evolution.
- ❏ Estimate the contribution of blazar emission to the extragalactic γ-ray background (EGB).

[Cavalie](https://ui.adsabs.harvard.edu/search/q=author:%22Cavaliere%2C+A.%22&sort=date%20desc,%20bibcode%20desc)re [& D'E](https://ui.adsabs.harvard.edu/search/q=author:%22D)lia 2002. Ajello et al. 2014. Ajello et al. 2015.

Conclusion

- ❏ LF follows a double power-law shape.
- ❏ LF extends to 4 orders of magnitude lower in luminosities than previous studies.
- ❏ LF seems to be best represented by LDDE.
- Hint of a link between FSRQs and BL Lacs can be seen.
- Separate analysis of FSRQs and BL Lacs is required.