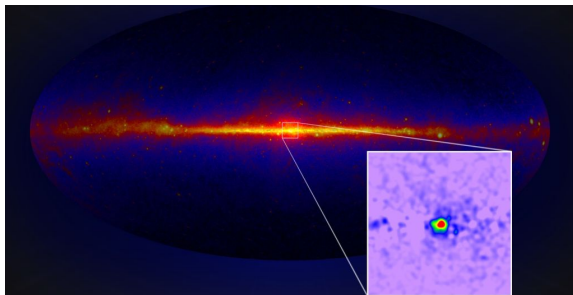


The gamma-ray mystery at the center of the Milky Way

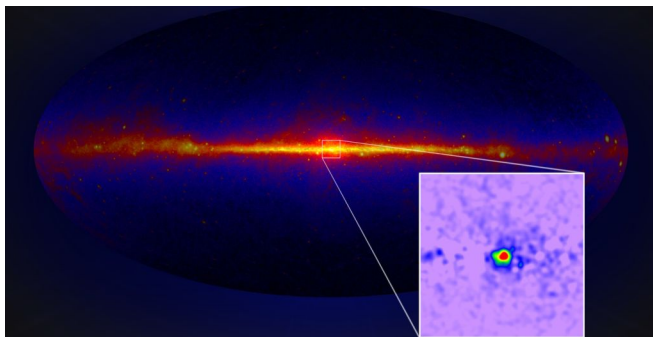


Ben Safdi
Massachusetts Institute of Technology

B.S., S. Lee, T. Linden, M. Lisanti, L. Necib, N. Rodd, S. Sharma, T. Slatyer, W. Xue

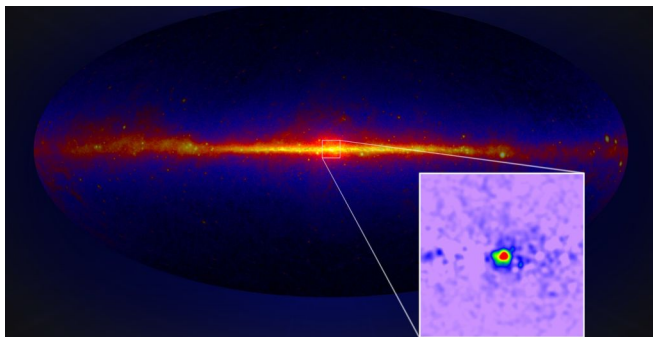
[JCAP 1505 (2015) , PRL 116 (2016), 1604.01026, 1606.04101]

Hints of dark matter annihilation in *Fermi* data?



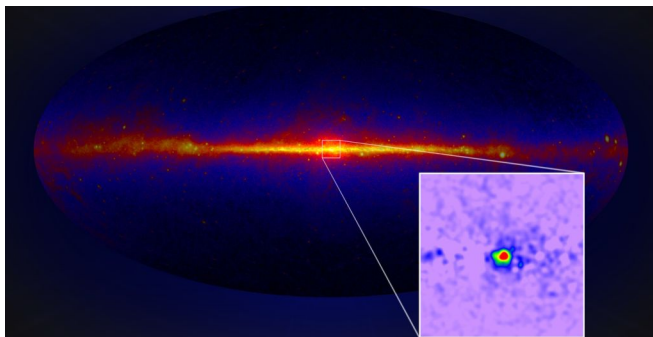
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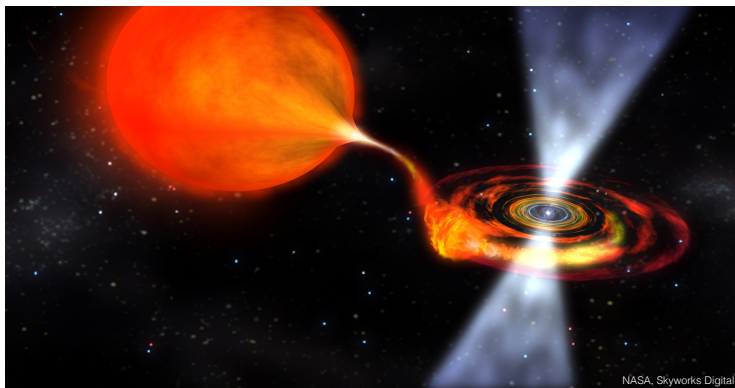
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(but see E. Carlson et. al. 2016)

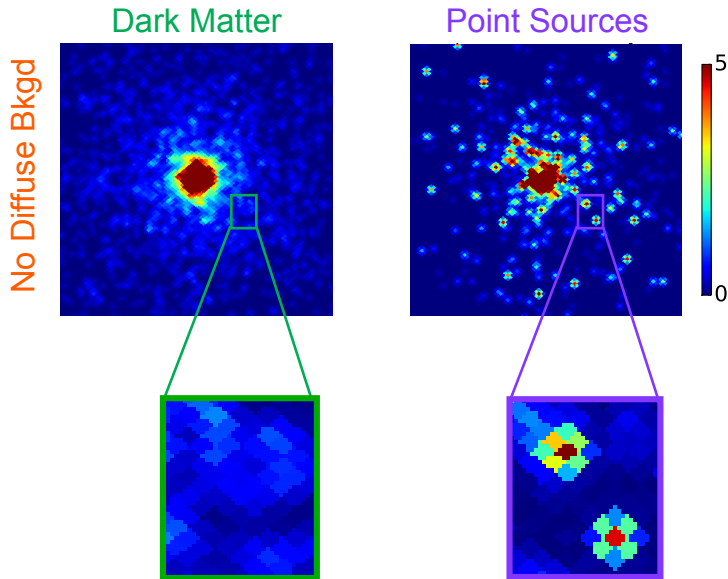
Dark Matter or dim Point Sources?



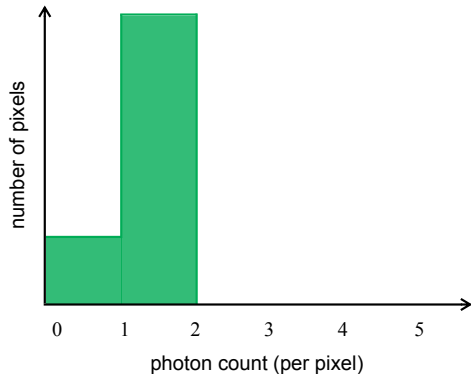
- New method: Non Poissonian Template Fit (NPTF)

- ▶ *JCAP* 2015: S. Lee, M. Lisanti, **B. S.**
- ▶ *Phys. Rev. Lett.* 2016: S. Lee, M. Lisanti, **B. S.**, T. Slatyer, W. Xue
- ▶ 1604.01026: T. Linden, N. Rodd, **B.S.**, T. Slatyer
- ▶ many works in progress: **B.S.**, ...

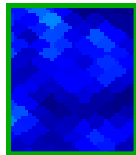
Photon Statistics: DM vs. Point Sources



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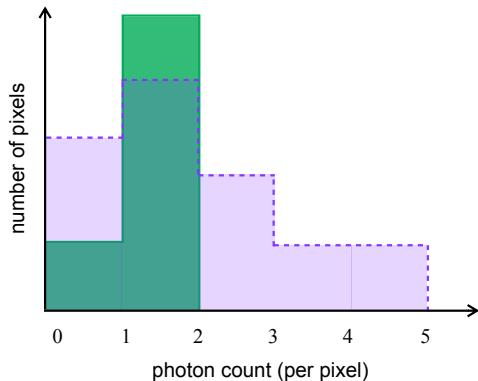


Dark Matter

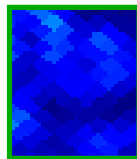


P(D) distribution in X-ray astronomy; Malyshev and Hogg, 2011; Lee, Lisanti, **BS** 2014

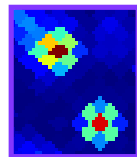
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Non-Poissonian template fit (NPTF)

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- ▶ The likelihood function:

$$p(d|\theta, \mathcal{M}) = \prod_{\text{pixels } p} p_{n_p}^{(p)}(\theta)$$

Thank you *Fermi* !

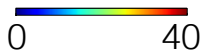
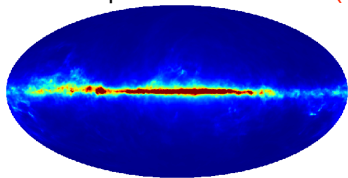


Fermi (NASA)

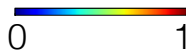
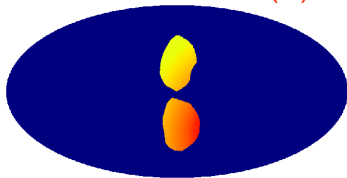
- ▶ **Pass 8 data:**
Ultracleanveto class, top quartile by PSF (August 4, 2008—June 3, 2015)
- ▶ **Energy range:** $\sim 2\text{--}12$ GeV

The models: Poissonian templates

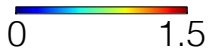
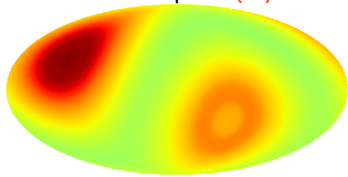
Fermi p6v11 diffuse (1)



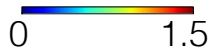
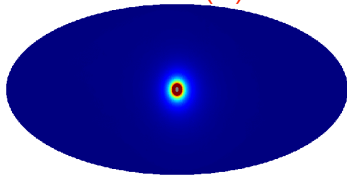
Fermi bubbles (1)



Isotropic (1)

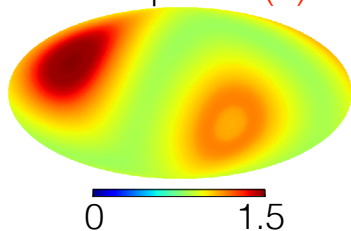


NFW (1)

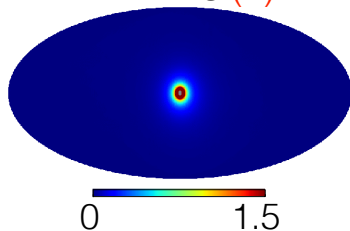


The models: Non-Poissonian templates

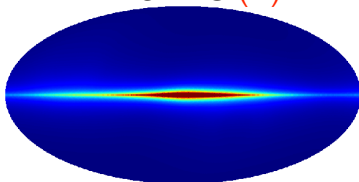
Isotropic PS (4)



NFW PS (4)



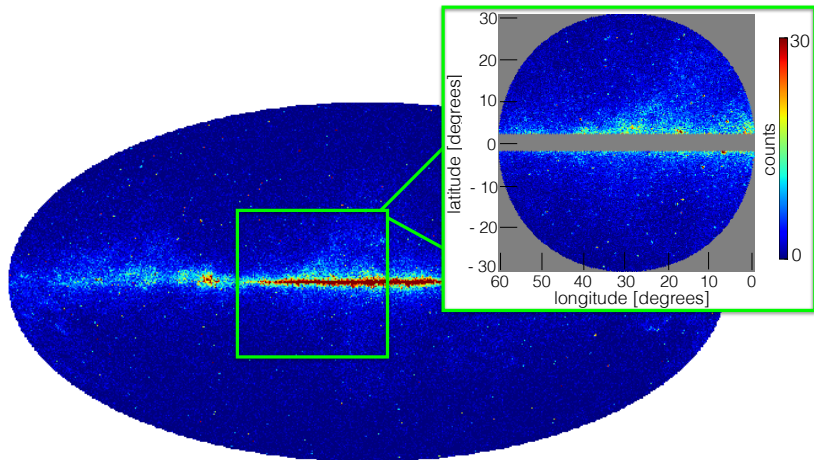
Disk PS (4)



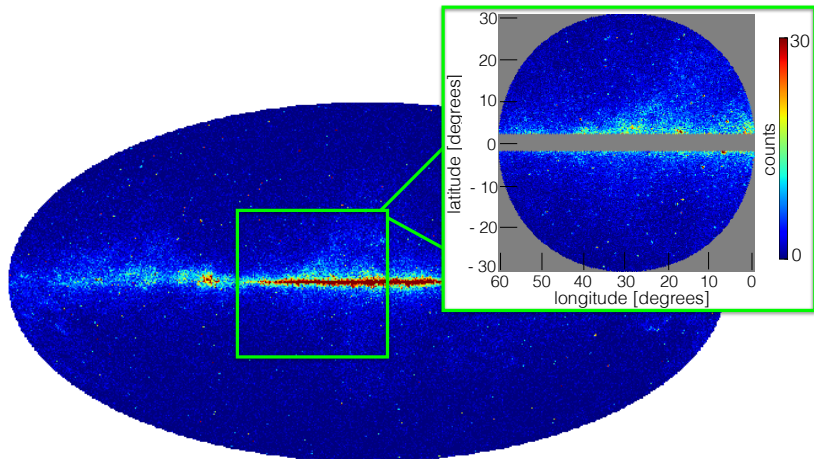
- Disk: $n \propto \exp(-R/5 \text{ kpc}) \exp(-|z|/0.3 \text{ kpc})$

Check 1: the $\ell = 30^\circ$ excess

Mask 4° around plane, out to 30° around $\ell = 30^\circ$



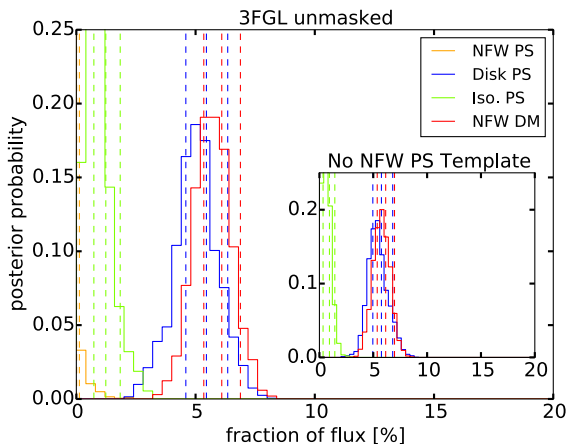
Mask 4° around plane, out to 30° around $l = 30^\circ$



- Plots normalized for region within 10° of ROI center ($b \geq 4^\circ$).

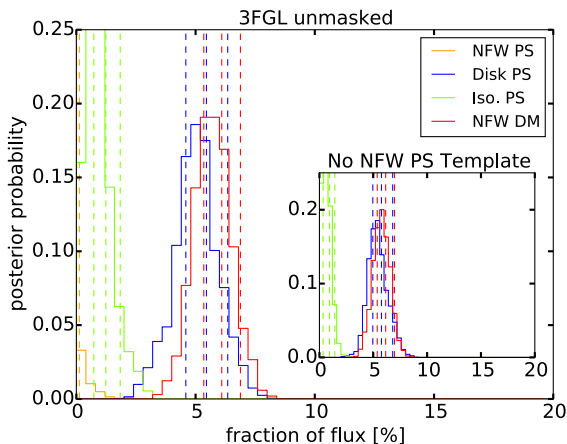
The $\ell = 30^\circ$ excess: no evidence for spherical PSs

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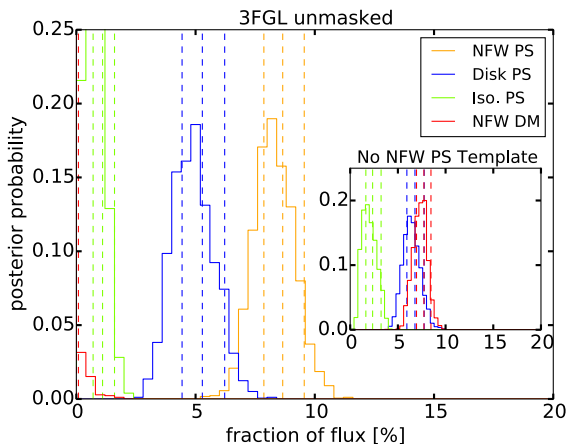


- Bayes factor ~ 0.1

ROI: the $\ell = 0^\circ$ excess

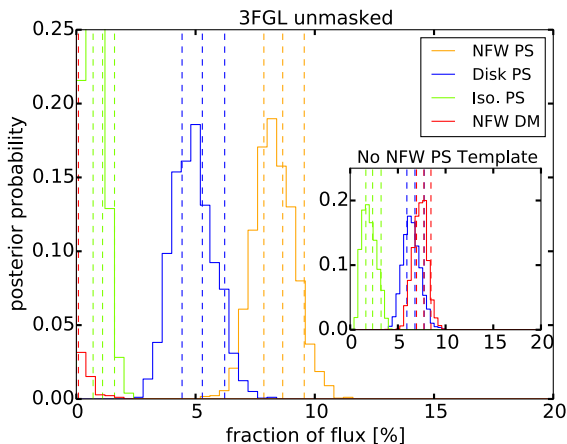
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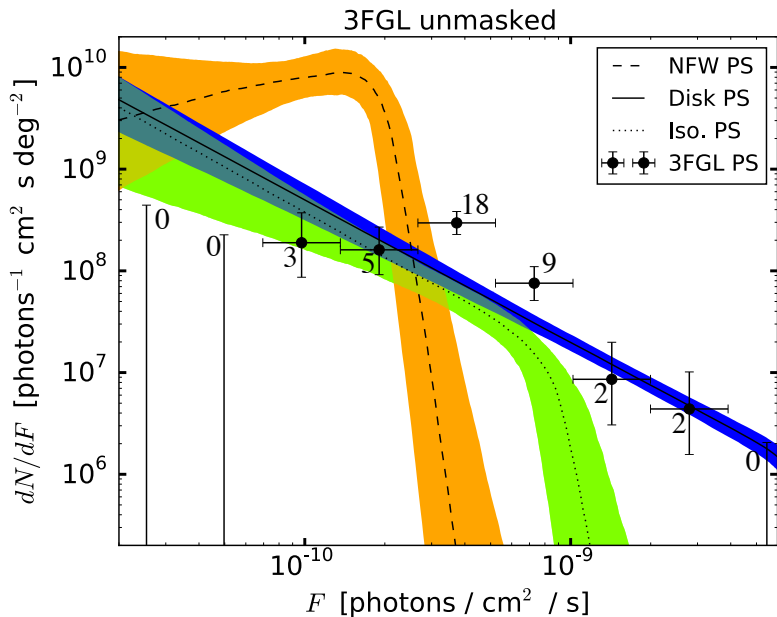
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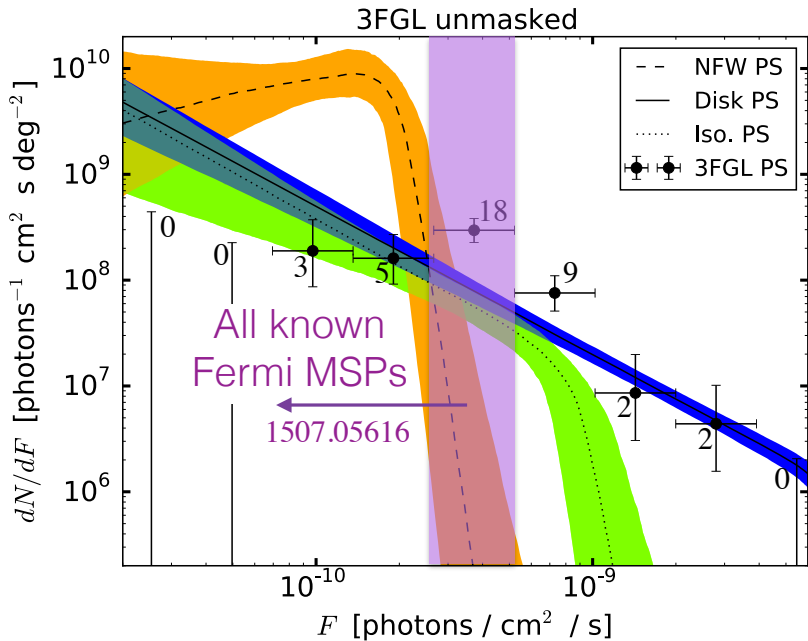


- Bayes factor $\sim 10^9$ (3FGL unmasked), 10^4 (3FGL masked)

The $\ell = 0^\circ$ excess: source-count function

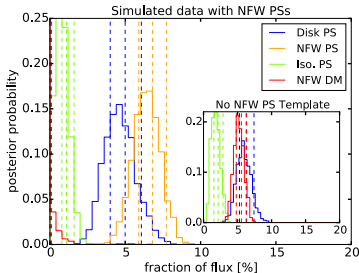
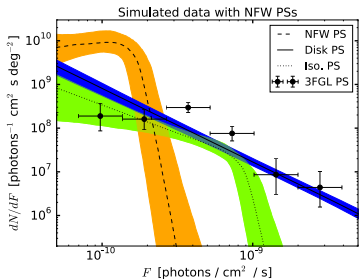


PSs consistent with MSP luminosity function?

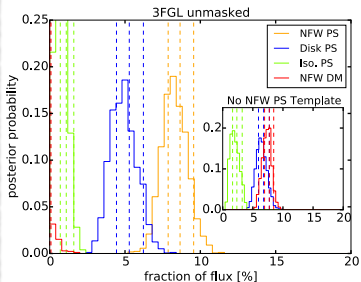
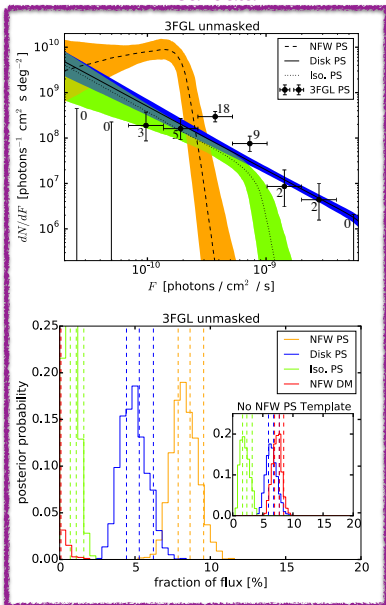


Check 2: Monte Carlo

The $\ell = 0^\circ$ excess: Monte Carlo

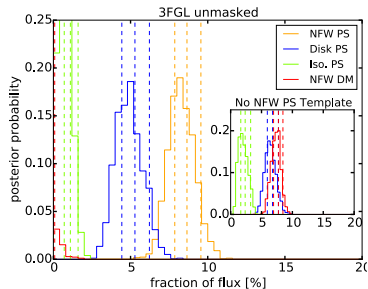
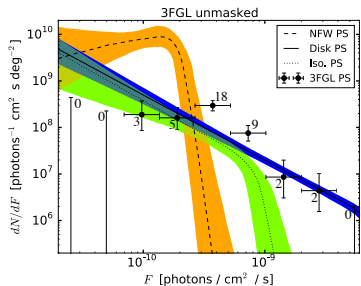
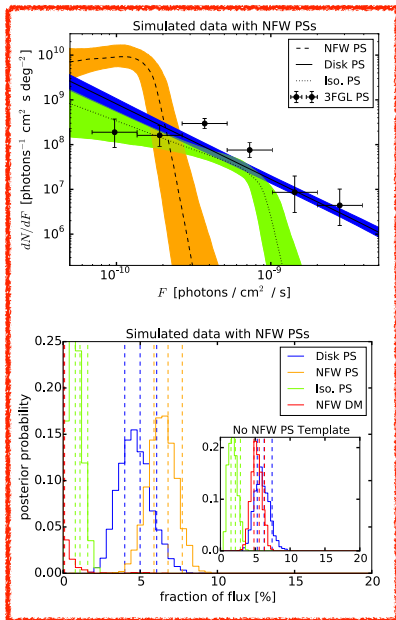


Real data



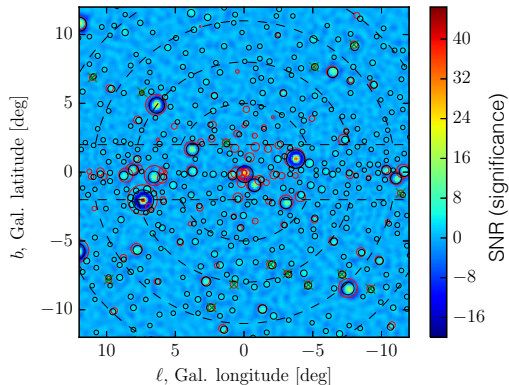
The $\ell = 0^\circ$ excess: Monte Carlo

Simulated data



Wavelet approach comes to same conclusion

- Bartels, Krishnamurthy, Weniger (PRL 2016)
- Signal-to-noise ratio (SNR; S) of wavelet transform
- Filters out structure of specific size (PSF)



- **red: 3FGL**, black: $S > 2$

Radio followup survey

- Follow-up survey in radio (**Green Bank, Parkes**) for **MSPs**
- **Submitted and submitting:** proposals for Green Bank observing time
- Simulation results: **~ 100 hours** of observation time, find **~ 5 MSPs** in the bulge

Parkes (Australia)



Green Bank (West Virginia)



with T. Linden, S. Ransom, N. Rodd, P. Ray, J. Thaler, C. Weniger, . . . , *Fermi* members (E. Charles, M. Di Mauro)

Tentative conclusion: GeV excess better fit by point-source emission than smooth (DM) emission

The NPTF Code Package (NPTFit)

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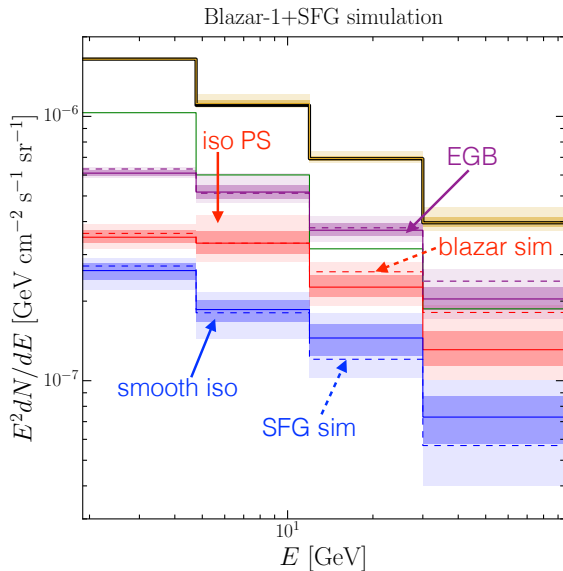
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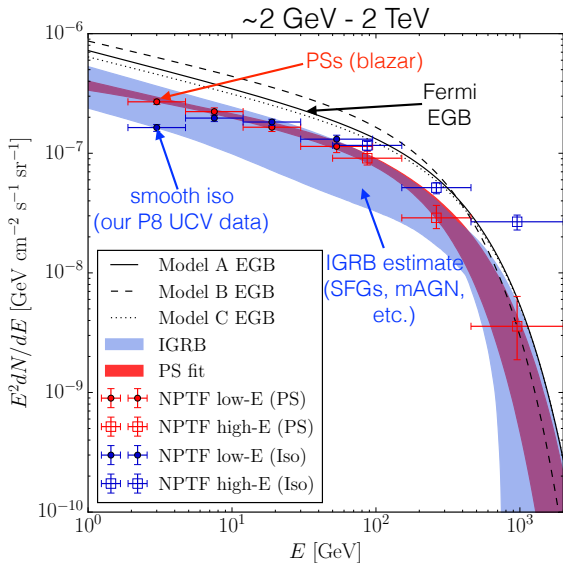
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- ▶ **L. Necib (MIT), N. Rodd (MIT), B.S., Siddharth Sharma (Princeton)**

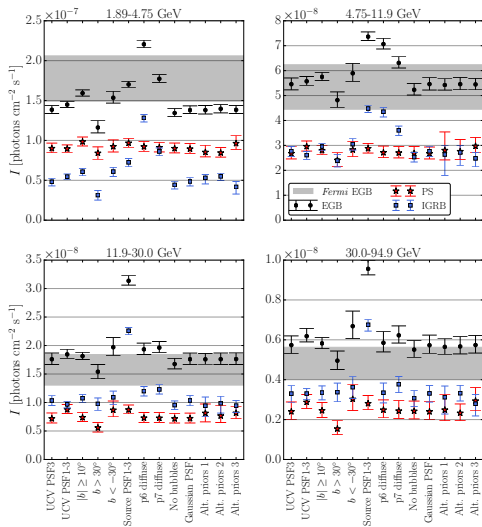
NPTF at high-lat. constrains extragal. PSs (blazars)



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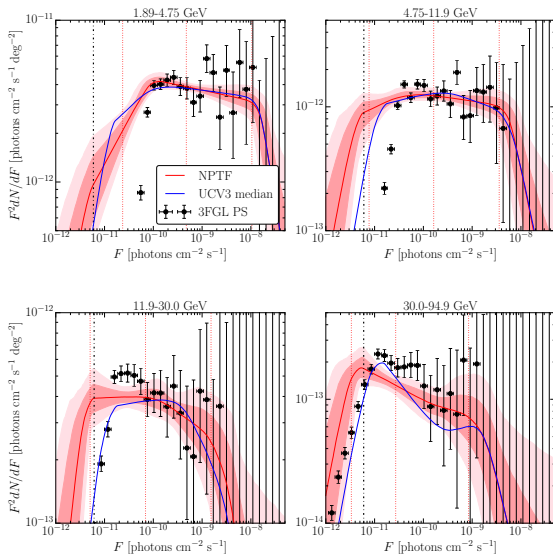


NPTF at high-lat. constrains extragal. PSs (blazars)



1606.04101: M. Lisanti, S. Mishra-Sharma, L. Necib, **B.S.**

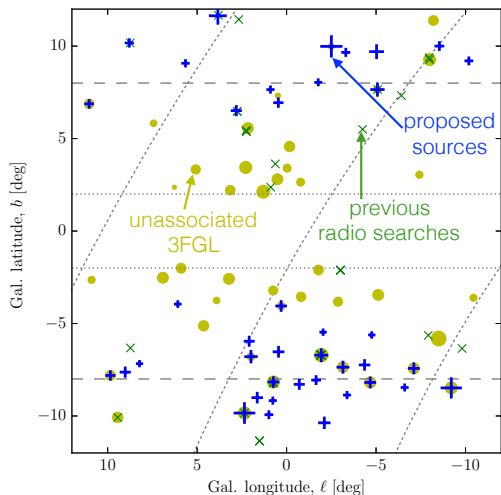
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Questions?

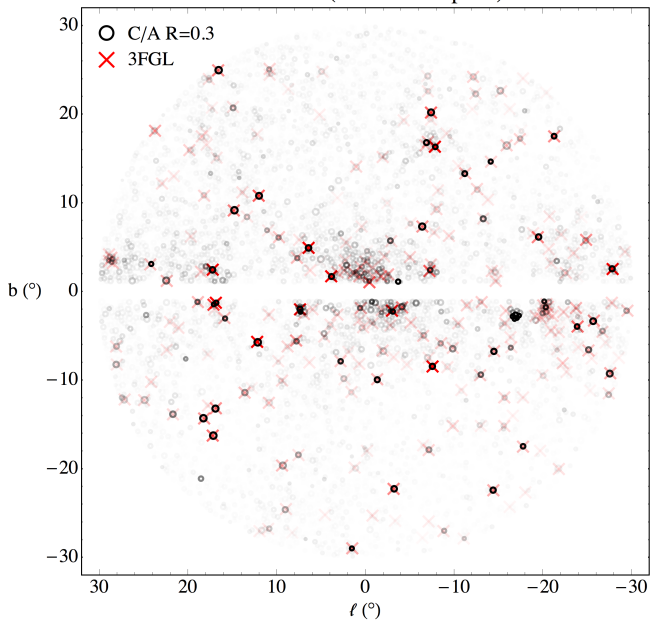
Radio followup survey: where to look

- Candidates identified through wavelet analysis + modified jet clustering analysis (N. Rodd, **B.S.**, J. Thaler) of *Fermi* data



PS candidates from jet clustering

Fermi Jets (no GCE template)



Statistics of PS candidates

- In each jet: $\epsilon^{(p)} \equiv 1 - \text{CDF}(\text{data}; \text{background model})$

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