

Indirect detection of dark matter



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SAHA Institute of Nuclear Physics



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Syllabus



- Dark matter paradigm and searches (Day 1)
- Indirect searches with gamma rays (Day 2)
- Data analysis: specific methodologies (Day 3)

Credit: NASA, HST, Webb



- Data analysis: specific methodologies
 - Profile likelihood method (IACTs)
 - Event-weighting method (IACTs)
 - Full likelihood method
- Detection prospects
- Proposals for hands-on session

Credit: NASA, HST, Webb

Annihilation (DM DM → SM SM → secondary γ)

$$\frac{\Phi_{\text{ann}}}{dE_\gamma}(E_\gamma, \Delta\Omega) = \underbrace{\frac{\langle\sigma v\rangle}{8\pi m_{\text{DM}}^2} \frac{dN}{dE}}_{\text{Particle Physics factor}} \int_0^{\Delta\Omega} \int_{\text{l.o.s.}} \rho(l, \Omega)^2 dl d\Omega$$

Astrophysical factor

Decay (DM → $\gamma\gamma$, DM → SM SM → secondary γ)

$$\frac{\Phi_{\text{decay}}}{dE_\gamma}(E_\gamma, \Delta\Omega) = \underbrace{\frac{1}{4\pi t_{\text{DM}} m_{\text{DM}}} \frac{dN}{dE}}_{\text{Particle Physics factor}} \int_0^{\Delta\Omega} \int_{\text{l.o.s.}} \rho(l, \Omega) dl d\Omega$$

Astrophysical factor

Local volume, DM mass below few TeV: attenuation not relevant

Galaxies 2022, 10, 92

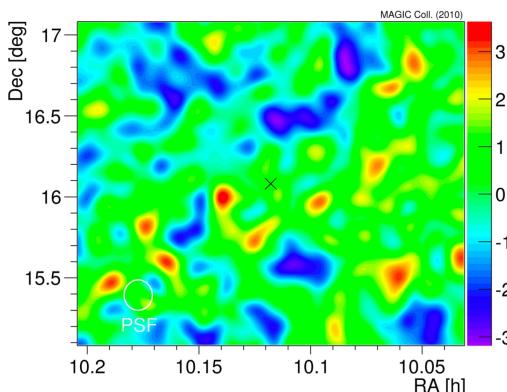
Data analysis: main idea

Annihilation (DM DM \rightarrow SM SM \rightarrow secondary γ)

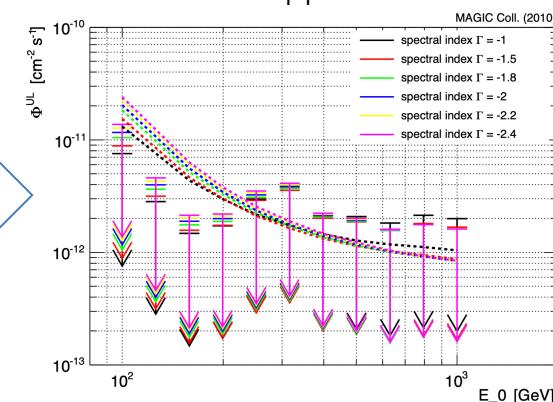
$$\frac{\Phi_{\text{ann}}}{dE_\gamma}(E_\gamma, \Delta\Omega) = \underbrace{\frac{\langle\sigma v\rangle}{8\pi m_{\text{DM}}^2} \frac{dN}{dE}}_{\text{Particle Physics factor}} \int_0^{\Delta\Omega} \int_{\text{l.o.s.}} \rho(l, \Omega)^2 dl d\Omega$$

Astrophysical factor

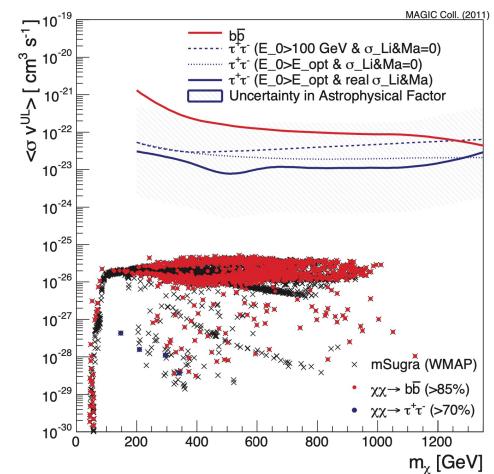
No significant emission



Flux upper limits

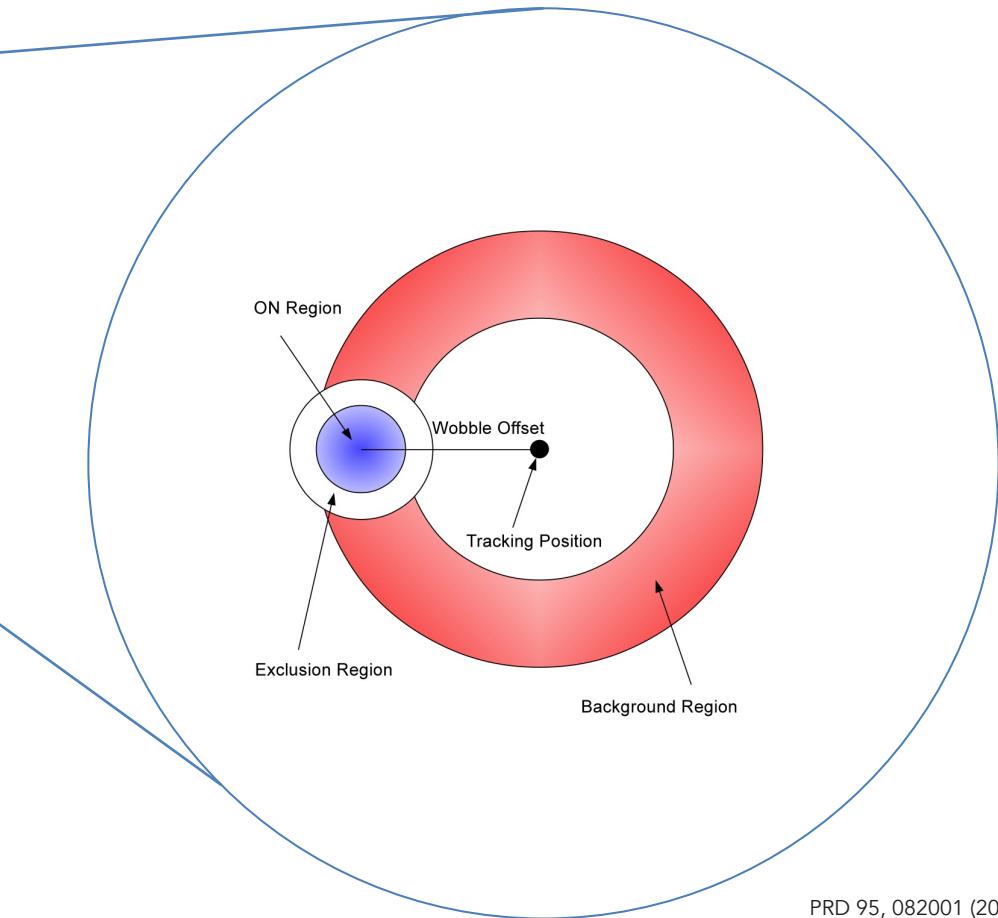
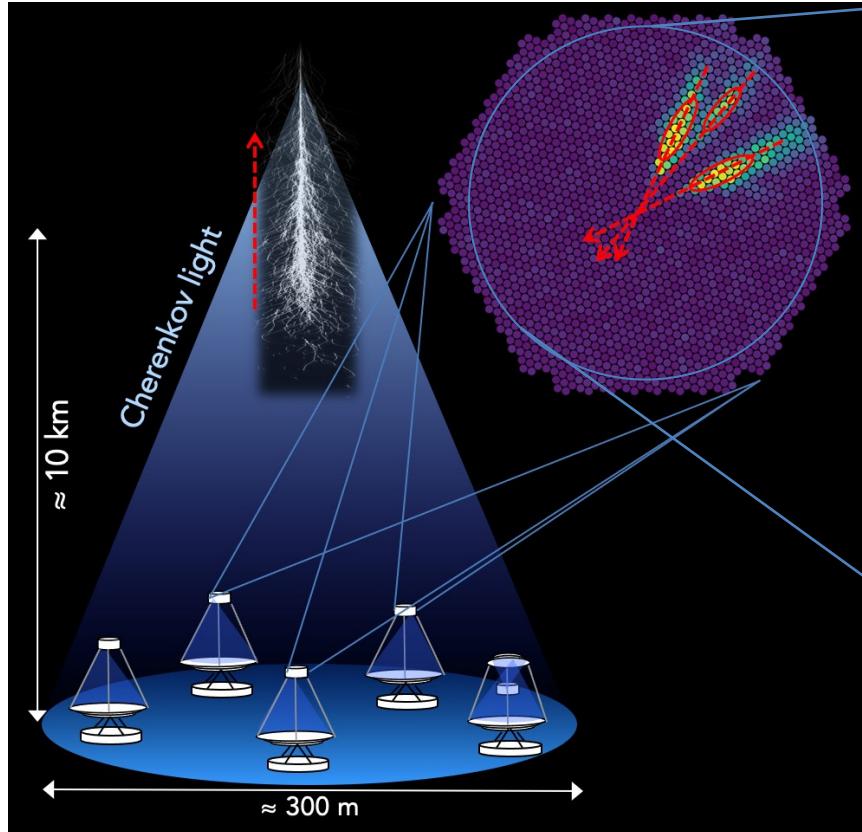


DM model constraints



Segue 1 - MAGIC I - JCAP 1106:035,2011

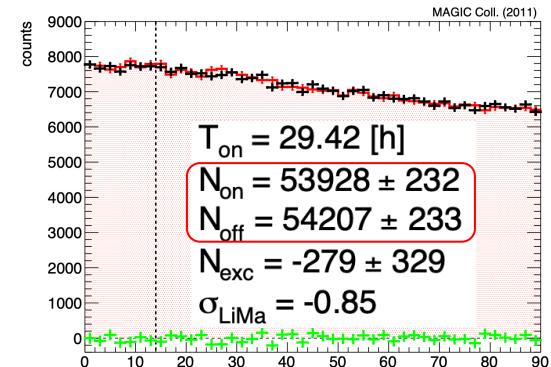
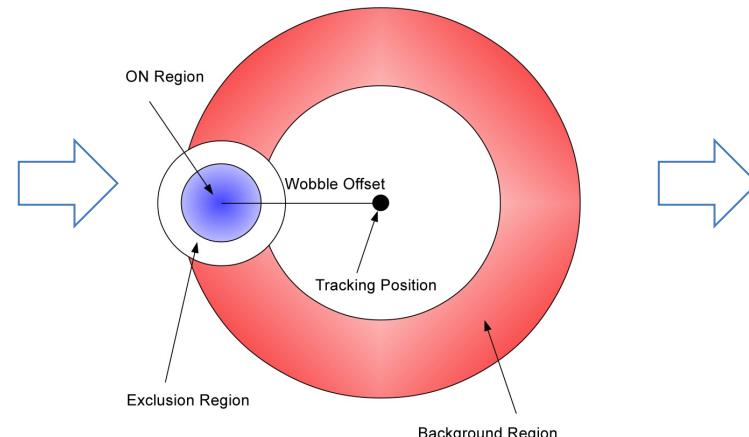
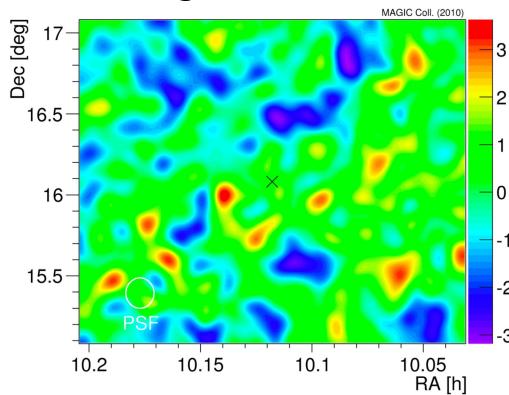
Profile likelihood method (IACTs)



PRD 95, 082001 (2017)

Profile likelihood method (IACTs)

No significant emission



Integral flux upper limits

$$\Phi^{UL}(> E_0) = \frac{N_{exc}^{UL}(E_{rec} > E_0)}{t_{eff}} \frac{\int_{E_0}^{\infty} S(E) dE}{\int_0^{\infty} A_{eff}(E; E_{rec} > E_0) S(E) dE}$$

Expected differential spectrum

Profile likelihood method (IACTs)

Profile likelihood (Rolle et al. 2009)

$$L(\boldsymbol{\pi}, \boldsymbol{\theta}|X) = \prod_{i=1}^n f(X_i|\boldsymbol{\pi}, \boldsymbol{\theta})$$

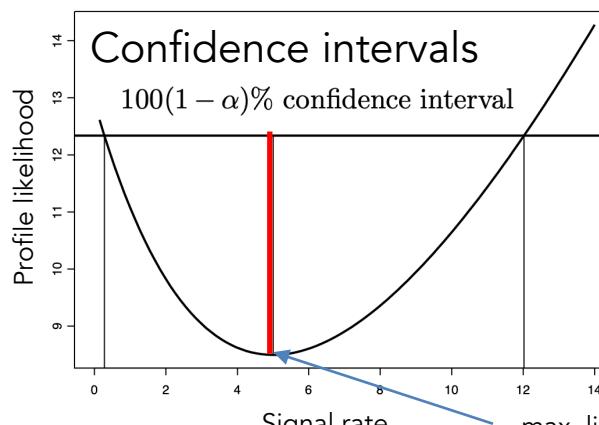
Likelihood
PDF
Nuisance parameters
Indep. observations
Parameters of interest

Profile likelihood

$$\lambda(\boldsymbol{\pi}_0|X) = \frac{\sup \{L(\boldsymbol{\pi}_0, \boldsymbol{\theta}|X); \boldsymbol{\theta}\}}{\sup \{L(\boldsymbol{\pi}, \boldsymbol{\theta}|X); \boldsymbol{\pi}, \boldsymbol{\theta}\}}$$

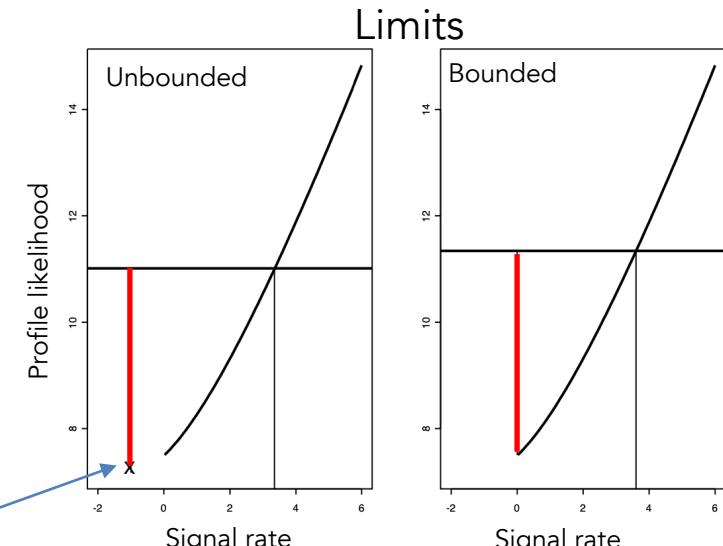
$$-2 \log \lambda \sim \chi^2$$

(with k degrees of freedom, for k parameters of interest)



α percentile of a χ^2

max. likelihood estimator (MLE)



Limits

Profile likelihood (Rolle et al. 2009)

Likelihood

$$L(\boldsymbol{\pi}, \boldsymbol{\theta}|X) = \prod_{i=1}^n f(X_i|\boldsymbol{\pi}, \boldsymbol{\theta})$$

Nuisance parameters

Parameters of interest

PDF

Indep. observations

Profile likelihood

$$\lambda(\boldsymbol{\pi}_0|X) = \frac{\sup \{L(\boldsymbol{\pi}_0, \boldsymbol{\theta}|X); \boldsymbol{\theta}\}}{\sup \{L(\boldsymbol{\pi}, \boldsymbol{\theta}|X); \boldsymbol{\pi}, \boldsymbol{\theta}\}}$$

$$-2 \log \lambda \sim \chi^2$$

(with k degrees of freedom, for k parameters of interest)

IACTs

$X \sim Pois(\mu + b)$, $Y \sim Pois(\tau b)$

N_{on}

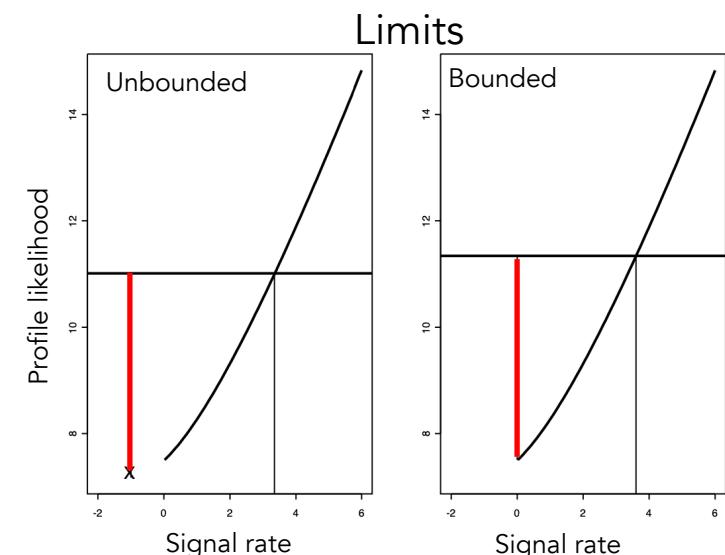
N_{off}

Signal rate

Bkg rate

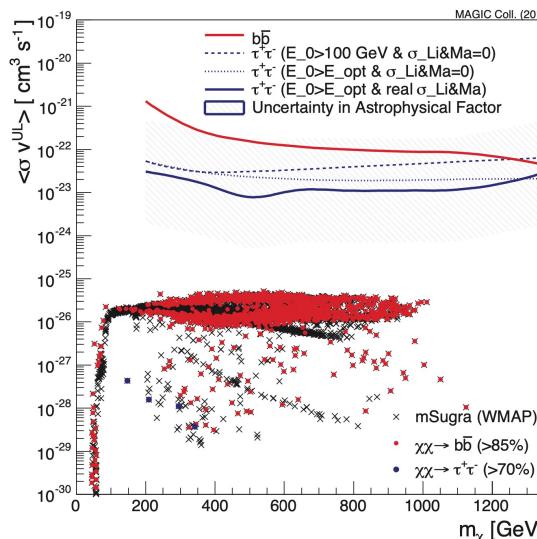
$$f(x, y|\mu, b) = \frac{(\mu + b)^x}{x!} e^{-(\mu+b)} \cdot \frac{(\tau b)^y}{y!} e^{-\tau b}$$

$$\lambda(\mu|x, y) = \frac{L(\mu, \hat{b}(\mu)|x, y)}{L(\hat{\mu}, \hat{b}|x, y)}$$

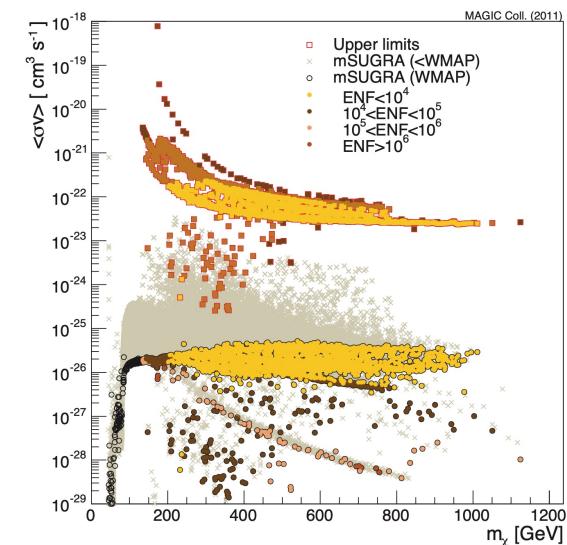


$$\left. \begin{aligned} \Phi(> E_0, \Delta\Omega) &= \Phi_{\epsilon}^{PP}(> E_0) J(\Delta\Omega) \\ \Phi_{\epsilon}^{PP}(> E_0) &= \frac{1}{4\pi} \frac{\langle \sigma_{\text{ann}} v \rangle}{2m_{\chi}^2} \int_{E_0}^{m_{\chi}} \sum_{i=1}^n B^i \frac{dN_{\gamma}^i}{dE} dE \\ J(\Delta\Omega) &= \int_{\Delta\Omega} \int_{los} \rho^2(r(s, \Omega)) ds d\Omega \end{aligned} \right\} \langle \sigma_{\text{ann}} v \rangle^{UL} = \frac{8\pi m_{\chi}^2 \Phi^{UL}(> E_0)}{\tilde{J}(\Delta\Omega) \int_{E_0}^{m_{\chi}} \frac{dN_{\gamma}}{dE} dE}$$

Model-agnostic limits



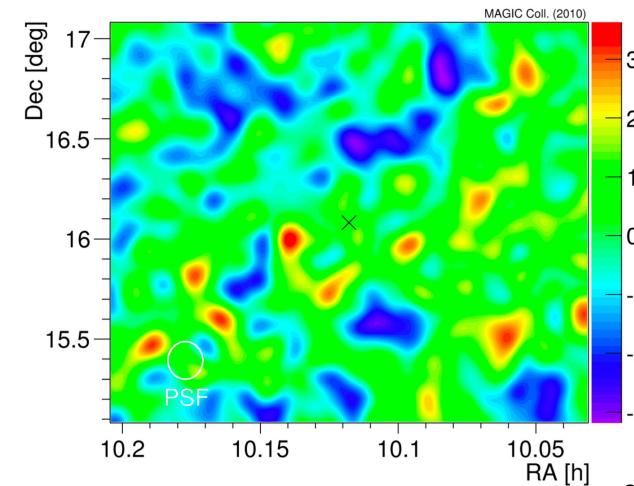
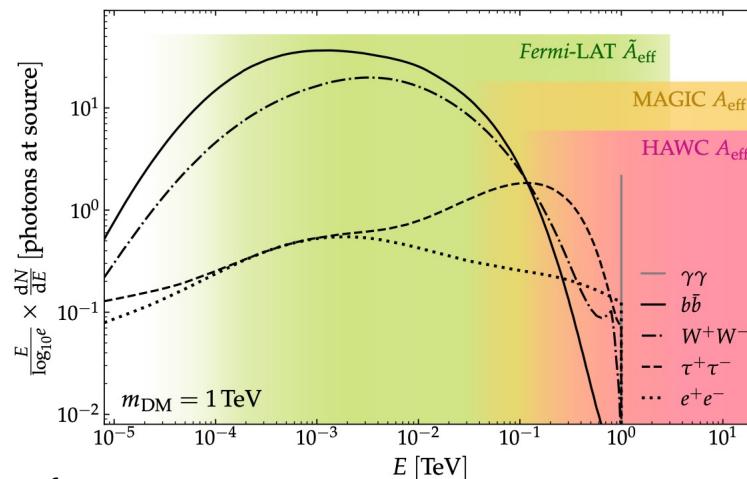
Model-dependent limits



Segue 1 - MAGIC I - JCAP 1106:035,2011

Event-weighting method (IACTs)

Not all events are equally informative → assign relevance based on event weights



(Rec.) Energy for event i

(Rec.) Distance to target for event i

$$w_i = w(\nu_i, E_i, \theta_i)$$

Astro. target for event i

Weight for event i

Test statistic

$$T = \sum_i w_i$$

Likelihood ratio method

Optimized weight

$$w = \log \left[1 + \frac{s}{b} \right]$$

$$s(\nu, E, \theta)$$

$$b(\nu, E, \theta)$$

expected # bkg events

Event-weighting method (IACTs)

$w_i = w(\nu_i, E_i, \theta_i)$ (Rec.) Energy for event i (Rec.) Distance to target for event i Astro. target for event i Weight for event i	Test statistic $T = \sum_i w_i$ Likelihood ratio method	Optimized weight $w = \log \left[1 + \frac{s}{b} \right]$	expected # signal events $s(\nu, E, \theta)$ expected # bkg events $b(\nu, E, \theta)$
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Estimation of s

$$s(\nu, E, \theta) = \frac{dN(\nu, E, \theta)}{dEd\Omega} dE 2\pi \sin(\theta) d\theta$$

$$\frac{dN(E, \hat{\mathbf{n}})}{dEd\Omega} = \int_{E_t} \int_{\Omega_t} dE_t d\Omega_t \frac{dF(E_t, \hat{\mathbf{n}}_t)}{dE_t d\Omega_t} R(E, \hat{\mathbf{n}} | E_t, \hat{\mathbf{n}}_t)$$

$$R(E, \hat{\mathbf{n}} | E_t, \hat{\mathbf{n}}_t) = \sum_{\text{runs}} \tau A_{\text{eff}}(E_t) \text{PSF}(\hat{\mathbf{n}} | E_t, \hat{\mathbf{n}}_t) D(E | E_t)$$

Instrument response runs Effective area

DM spectrum

Obs. time

Angular res.

Energy res.

Event-weighting method (IACTs)

$w_i = w(\nu_i, E_i, \theta_i)$ (Rec.) Energy for event i (Rec.) Distance to target for event i Astro. target for event i Weight for event i	Test statistic $T = \sum_i w_i$ Likelihood ratio method	Optimized weight $w = \log \left[1 + \frac{s}{b} \right]$	expected # signal events $s(\nu, E, \theta)$ expected # bkg events $b(\nu, E, \theta)$
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Estimation of s

$$s(\nu, E, \theta) = \frac{dN(\nu, E, \theta)}{dEd\Omega} dE 2\pi \sin(\theta) d\theta$$

Estimation of b

$$b(\nu, E, \theta)$$

empirical energy distribution of the OFF events
+ assuming isotropy within the ON region

$$\frac{dN(E, \hat{\mathbf{n}})}{dEd\Omega} = \int_{E_t} \int_{\Omega_t} dE_t d\Omega_t \frac{dF(E_t, \hat{\mathbf{n}}_t)}{dE_t d\Omega_t} R(E, \hat{\mathbf{n}} | E_t, \hat{\mathbf{n}}_t)$$

DM spectrum

$$R(E, \hat{\mathbf{n}} | E_t, \hat{\mathbf{n}}_t) = \sum_{\text{runs}} \tau A_{\text{eff}}(E_t) \text{PSF}(\hat{\mathbf{n}} | E_t, \hat{\mathbf{n}}_t) D(E | E_t)$$

Instrument response

Energy res.

Angular res.

Effective area

Obs. time

Computation of upper limits

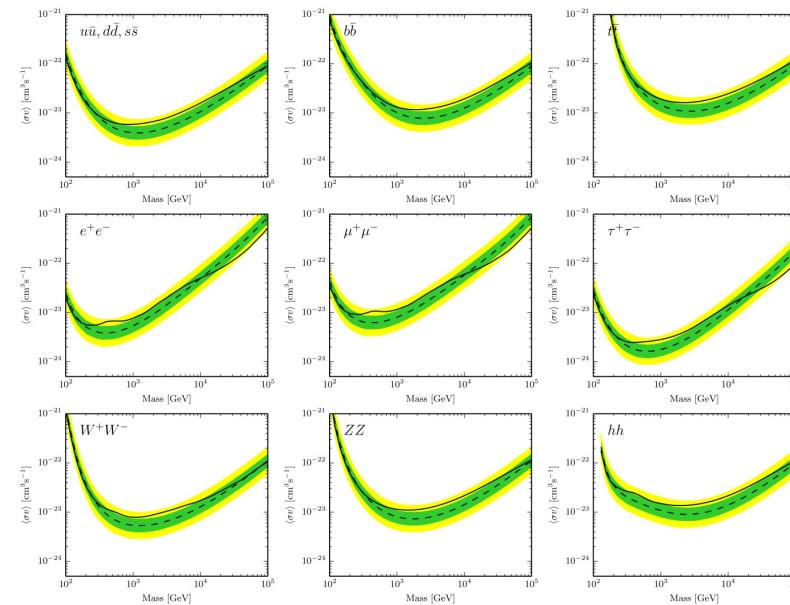
Hypothesis test is performed at every value of $\langle \sigma v \rangle$, for each DM mass and channel:

$$P(T < T_{obs}) < \alpha \rightarrow \text{hypothesis rejected at } 1 - \alpha \text{ confidence level} \rightarrow \langle \sigma v \rangle^{UL}$$

\uparrow
bkg only

VERITAS

Dwarf	Zenith [deg]	Azimuth [deg]	Exposure [hours]	Energy Range [GeV]
Segue 1	15–35	100–260	92.0	80–50000
Draco	25–40	320–40	49.8	120–70000
Ursa Minor	35–45	340–30	60.4	160–93000
Boötes 1	15–30	120–249	14.0	100–41000
Willman 1	20–30	340–40	13.6	100–43000



Profile likelihood for N_{targets} and $N_{\text{instruments}}$

$$\lambda_P(\langle \sigma v \rangle | \mathcal{D}) = \frac{\mathcal{L}(\langle \sigma v \rangle; \hat{\nu} | \mathcal{D})}{\mathcal{L}(\widehat{\langle \sigma v \rangle}; \hat{\nu} | \mathcal{D})}$$

Dataset Value maximizing joint likelihood
for a given $\langle \sigma v \rangle$

Nuisance parameters

$$\mathcal{L}(\langle \sigma v \rangle; \boldsymbol{\nu} | \mathcal{D}) = \prod_{i=1}^{N_{\text{target}}} \mathcal{L}_i(\langle \sigma v \rangle; J_i, \boldsymbol{\mu}_i | \mathcal{D}_i) \cdot \mathcal{J}(J_i | J_{\text{obs},i}, \sigma_i)$$

Likelihood for J Observed J

Uncertainty on J_{obs}

$$\mathcal{L}_i(\langle\sigma v\rangle; J_i, \boldsymbol{\mu}_i \mid \mathcal{D}_i) = \prod_{j=1}^{N_{\text{instrument}}} \mathcal{L}_{ij}(\langle\sigma v\rangle; J_i, \boldsymbol{\mu}_{ij} \mid \mathcal{D}_{ij})$$

$$\mathcal{J}(J_i | J_{\text{obs,i}}, \sigma_i) = \frac{1}{\ln(10)J_{\text{obs,i}}\sqrt{2\pi}\sigma_i} \times e^{-\left(\log_{10}(J_i) - \log_{10}(J_{\text{obs,i}})\right)^2 / 2\sigma_i^2}$$

Likelihood for J

Full likelihood method

Example: MAGIC (Segue 1) + Fermi-LAT (15 dSph)

Fermi-LAT ($j \equiv F$)

$$\mathcal{L}_{iF}(\langle\sigma v\rangle; J_i, \hat{\mu}_{iF} | \mathcal{D}_{iF}) = \prod_{k=1}^{N_{E-\text{bins}}} \mathcal{L}_{iFk}(\bar{E}\Phi_k(\langle\sigma v\rangle, J_i))$$

$$\bar{E}\Phi_k(\langle\sigma v\rangle, J_i) = \int_{E_{\min,k}}^{E_{\max,k}} dE E \frac{d\Phi}{dE}(\langle\sigma v\rangle, J_i)$$

MAGIC ($j \equiv M$)

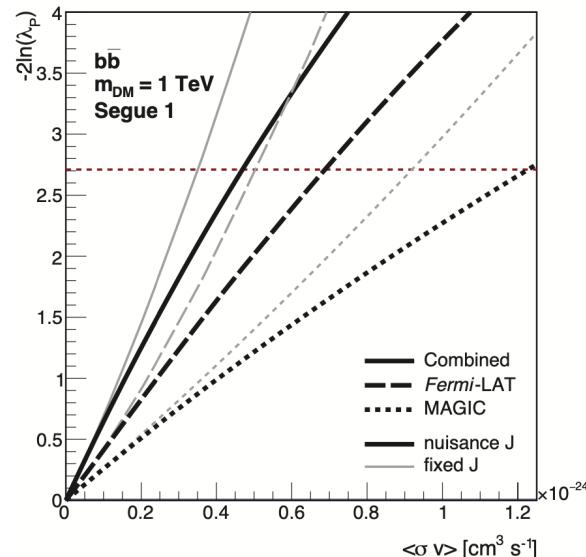
$$\mathcal{L}_{iM}(\langle\sigma v\rangle; J_i, \mu_{iM} | \mathcal{D}_{iM}) = \prod_{k=1}^N \mathcal{L}_{iMk}(\langle\sigma v\rangle; J_i, \mu_{iMk} | \mathcal{D}_{iMk})$$

$$\mathcal{L}(\langle\sigma v\rangle; J, \mu | \mathcal{D}) = \frac{(g + b/\tau)^{N_{\text{ON}}}}{N_{\text{ON}}!} e^{-(g+b/\tau)} \frac{b^{N_{\text{OFF}}}}{N_{\text{OFF}}!} e^{-b} \prod_{l=1}^{N_{\text{ON}}} f(g; b, \tau | E'_l) \prod_{m=1}^{N_{\text{OFF}}} h(E'_m)$$

$$f(g; b, \tau | E') = \frac{\frac{b}{\tau} h(E') + g p(E')}{\frac{b}{\tau} + g}$$

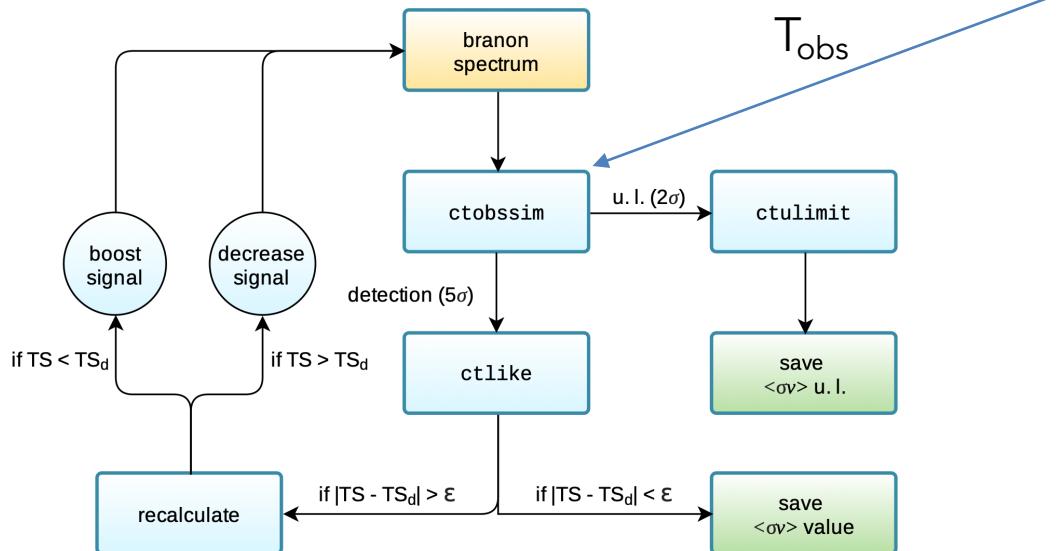
$$p(E') = \frac{T_{\text{obs}} \int_{E'_{\min}}^{E'_{\max}} dE' \int_0^\infty dE \frac{d\Phi}{dE}(\langle\sigma v\rangle, J) A(E) G(E; E')}{g}$$

$$-2 \ln \lambda_P(\langle\sigma v\rangle_{2.71} | \mathcal{D}) = 2.71$$

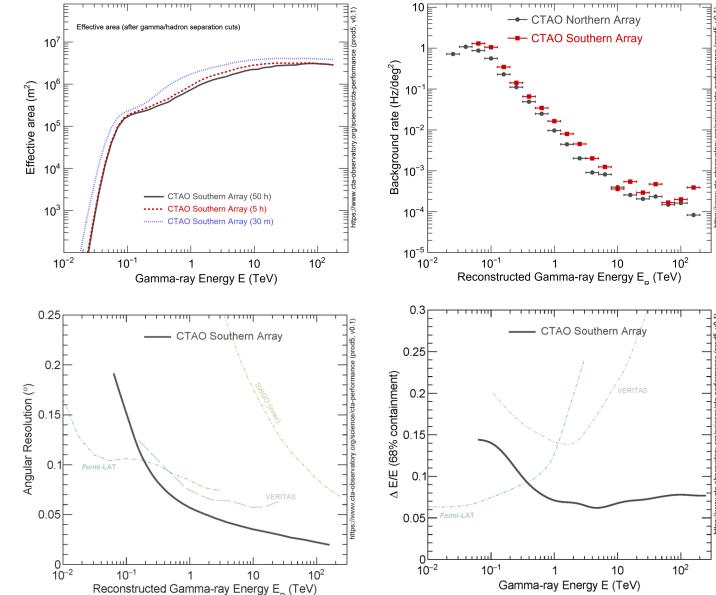


Detection prospects

$$\frac{\Phi_{\text{ann}}}{dE_\gamma}(E_\gamma, \Delta\Omega) = \frac{\langle\sigma v\rangle}{8\pi m_{\text{DM}}^2} \frac{dN}{dE} \int_0^{\Delta\Omega} \int_{\text{l.o.s.}} \rho(l, \Omega)^2 dl d\Omega$$



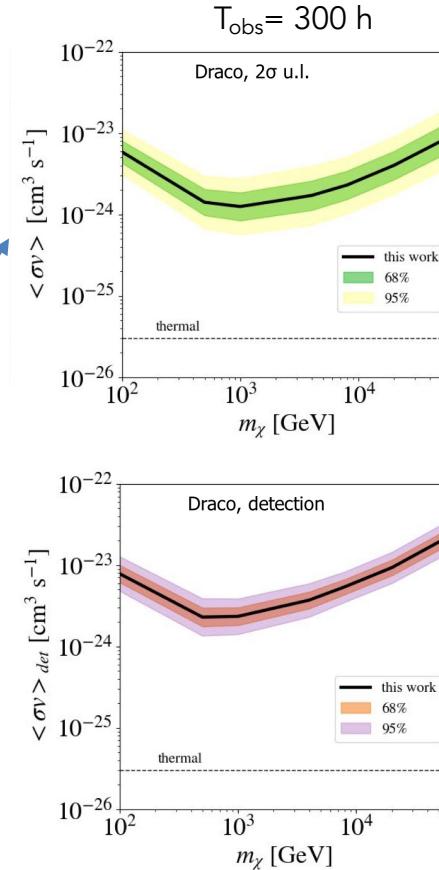
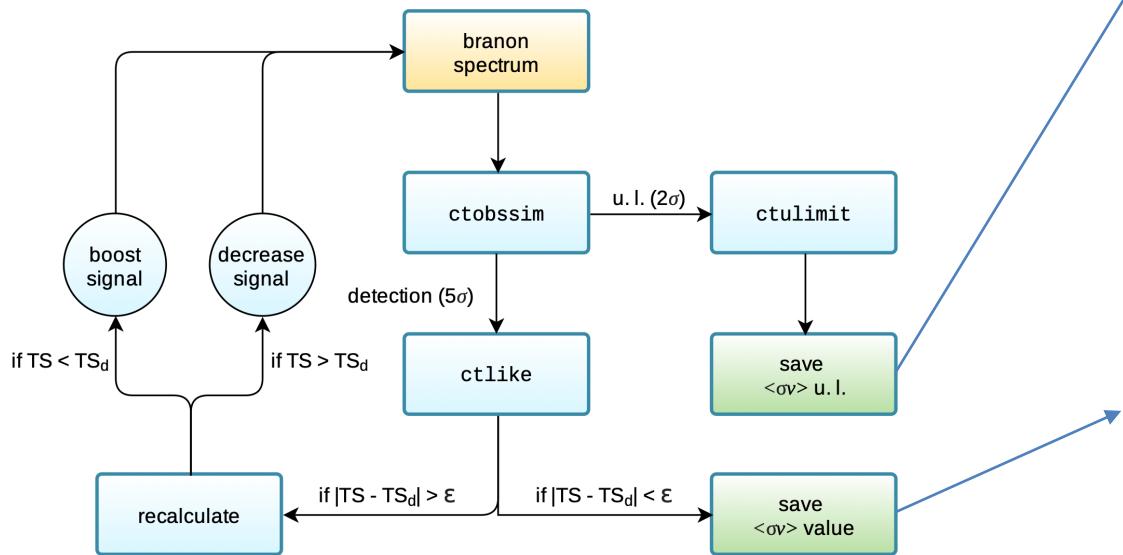
Instrument response functions



cherenkov
telescope
array

the observatory for
ground-based
gamma-ray astronomy

Detection prospects



Analyze Fermi-LAT data on Draco dSph and obtain $\langle \sigma v \rangle^{UL}$

You will be using Fermipy and this tutorial(*):

<https://github.com/fermiPy/fermipy-extra/blob/master/notebooks/draco.ipynb>

Exercises:

- Compute $\langle \sigma v \rangle^{UL}$ vs DM mass for different annihilation channels
- Increase exposure from 6 years to 14 years

(*) For recent installations of Fermipy one should update gll_iem_v06.fits to gll_iem_v07.fits in the config.yaml

Compute photon yield for branon DM using PPPC 4 DM ID:

Photon yields per annihilation channel can be found here:

<http://web.mit.edu/lns/research/CascadeSpectra.tar.gz>

Read `CascadeSpectraExample.py` to learn how to use the Python interface

Branching ratios can be computed following equations 2.2-2.4 in [JCAP05\(2022\)005](#)

Then, one can compute the model-dependent photon yield per annihilation like:

$$\frac{dN}{dE} = \sum_i B_i \left. \frac{dN}{dE} \right|_i$$

Some examples of the photon yield from branon annihilation can be found in Fig. 1 in [JCAP05\(2022\)005](#)

Compute astrophysical factors using pyCLUMPY:

The goal would be to compute the astrophysical factor of a user-defined halo and test how it depends on the integration angle

- 1.- Install CLUMPY: <https://clumpy.gitlab.io/CLUMPY/v3.1.1/install.html>
- 2.- Load Tutorial.ipynb in [CLUMPY/python_helperscripts/pyClumpy_Tutorial](#)
- 3.- Import necessary packages and go to Haloes section
- 4.- Run Haloes section picking your halo preformatted example from [list_generic.txt](#)
- 5.- Using the code in J-Factors (1D) subsection compute J factors and fluxes for an angle of 0.5° (typical for Fermi-LAT) and 0.1° (typical for IACTs). Compare the results

TASI Lectures on the Particle Physics and Astrophysics of Dark Matter

<https://arxiv.org/abs/2303.02169>

TeV Dark Matter Searches in the Extragalactic Gamma-ray Sky

<https://doi.org/10.3390/galaxies10050092>

Les Houches Lectures on Indirect Detection of Dark Matter

<https://arxiv.org/abs/2109.02696>

TASI Lectures on Indirect Searches For Dark Matter

<https://arxiv.org/abs/1812.02029>

Particle Dark Matter: Observations, Models, and Searches

Bertone et al., Cambridge University Press (2010)

The Review of Particle Physics (2023)

<https://pdg.lbl.gov>

Searches for dark matter annihilation signatures in the Segue 1 satellite galaxy with the MAGIC-I telescope
<https://doi.org/10.1088/1475-7516/2011/06/035>

Optimized dark matter searches in deep observations of Segue 1 with MAGIC
<https://doi.org/10.1088/1475-7516/2014/02/008>

Limits to dark matter annihilation cross-section from a combined analysis of MAGIC and Fermi-LAT observations of dwarf satellite galaxies
<https://doi.org/10.1088/1475-7516/2016/02/039>

Comprehensive search for dark matter annihilation in dwarf galaxies
<https://doi.org/10.1103/PhysRevD.91.083535>

Dark matter constraints from a joint analysis of dwarf Spheroidal galaxy observations with VERITAS
<https://doi.org/10.1103/PhysRevD.95.082001>

Fermi Large Area Telescope Fourth Source Catalog (Appendix B)
<https://doi.org/10.3847/1538-4365/ab6bcb>

Clumpy (Astrophysical factor)
<https://clumpy.gitlab.io/CLUMPY>

PPPC 4 DM (Particle Physics factor)
<http://www.marcocirelli.net/PPPC4DMID.html>

gLike (combined likelihood analysis)
<https://github.com/javierrico/gLike>

LlkCom (combined likelihood analysis)
https://github.com/TjarkMiener/likelihood_combiner