

Is DM hidden in the \bar{p} flux?

Pierre

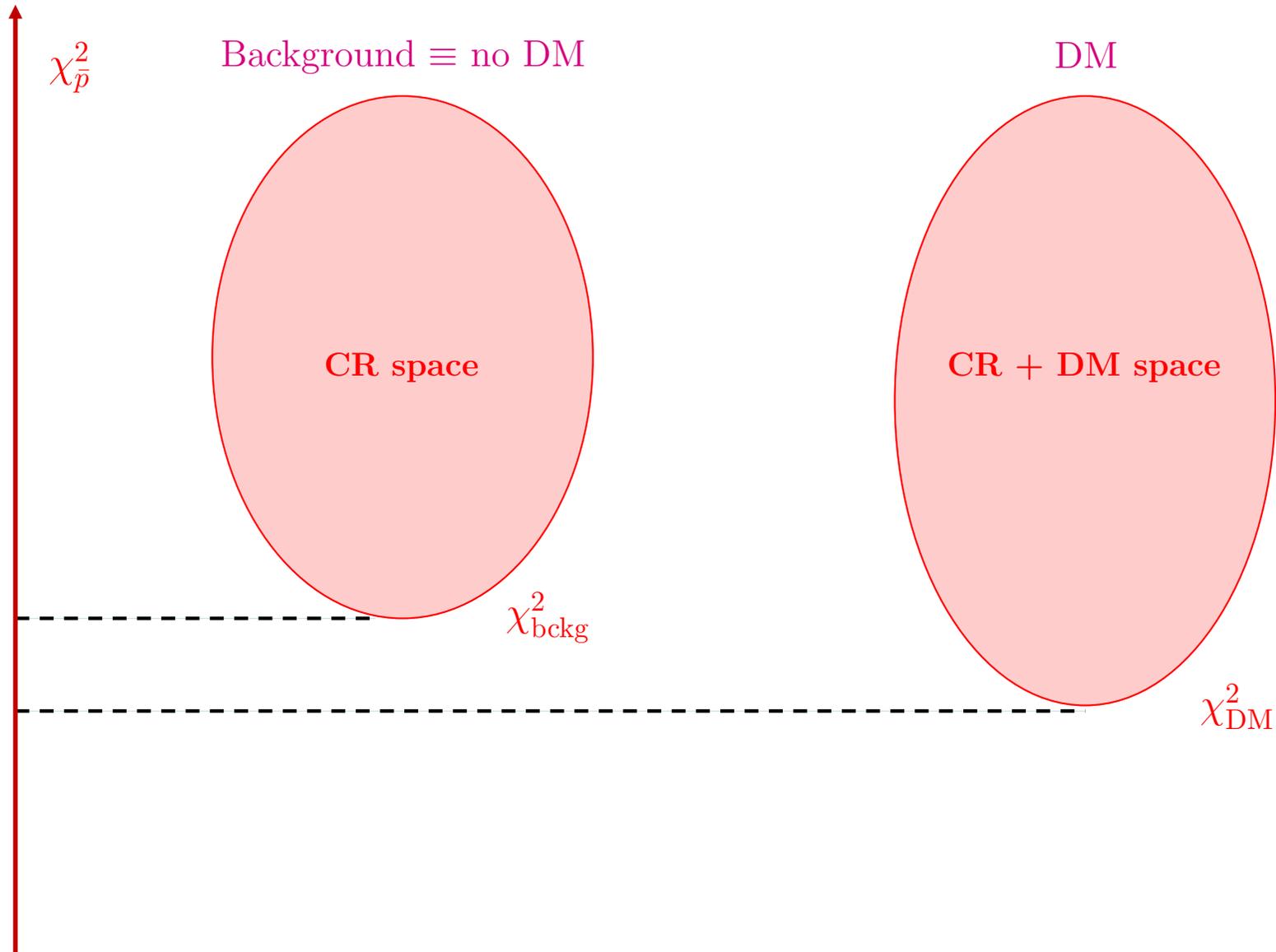
Outline

- 1) Incorporating theoretical systematics in \bar{p} fits
- 2) Spectral anomalies – Frequentist vs Bayesian
- 3) A few recent examples



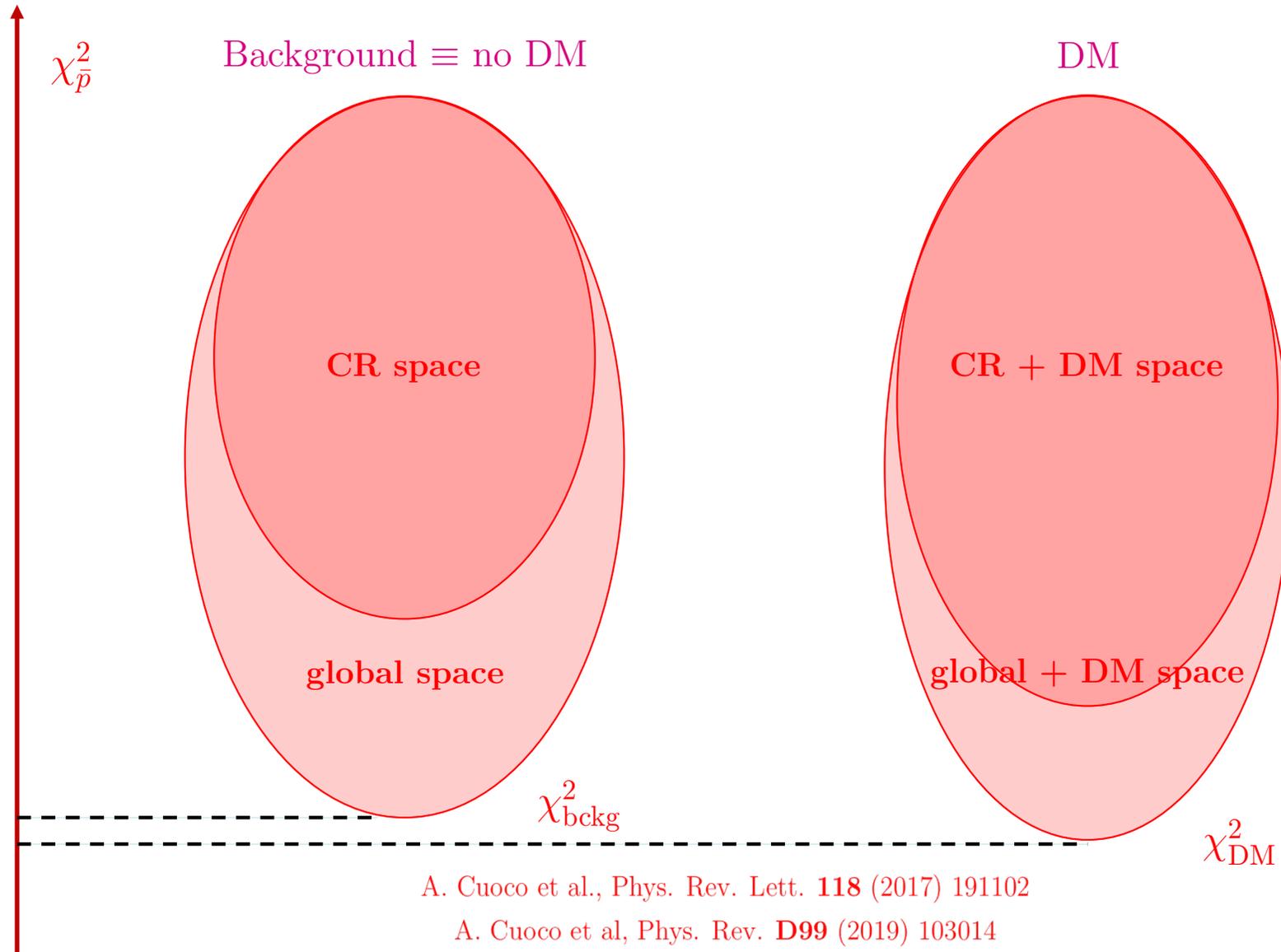
1) Incorporating theoretical systematics in \bar{p} fits

In general **systematics** \equiv **CR** + $\{\Phi_i = \Phi_p + \Phi_\alpha\}$ + $\frac{d\sigma_{ij \rightarrow \bar{p}}}{dE_{\bar{p}}}$



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Incorporating CR systematics in the \bar{p} fit

Background only vs DM hypotheses

(i) Old days agnostic scan of CR parameter space to find $\chi_{\bar{p}, \min}^2$

$P(\bar{p} | \text{flat } \theta_{\text{CR}}) \Rightarrow$ loss of information from B/C

(ii) Fit together \bar{p} and B/C to get best-fit CR parameters θ_{CR}

$P(\bar{p} + \text{B/C} | \theta_{\text{CR}}) \Rightarrow$ not the same question

Fit could be dominated by B/C and not \bar{p}

(iii) Fit \bar{p} and incorporate a theoretical CR uncertainty matrix such as

$$\mathcal{C}_{\text{nm}}^{\bar{p} \text{ source}} = \left\langle (\Phi_{\bar{p}}^{\text{th}})_n (\Phi_{\bar{p}}^{\text{th}})_m \right\rangle - \langle \Phi_{\bar{p}}^{\text{th}} \rangle_n \langle \Phi_{\bar{p}}^{\text{th}} \rangle_m \text{ with } \mathcal{C}_{\text{B/C}} \text{ matrix}$$

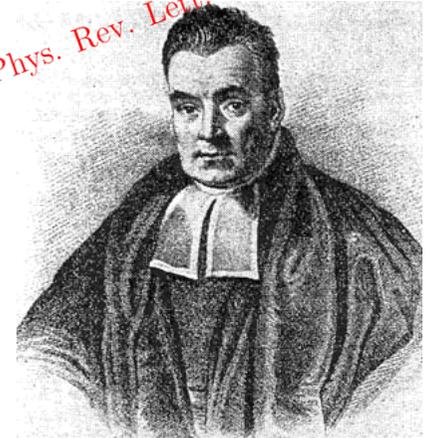
$$\mathcal{C}_{\text{nm}}^{\bar{p}} = \mathcal{C}_{\text{nm}}^{\text{AMS}} + \mathcal{C}_{\text{nm}}^{\bar{p} \text{ source}}$$

(iv) Use θ_{CR} as nuisance parameters when fitting \bar{p} data

$$\chi^2 = \chi_{\bar{p}}^2 + \{ \chi_{\text{CR}}^2 \equiv (\theta_p - \bar{\theta}_p) \mathcal{C}_{\text{pq}}^{\text{CR}} (\theta_q - \bar{\theta}_q) \}$$

How is $\mathcal{C}_{\text{pq}}^{\text{CR}}$ defined? If $\mathcal{C}^{\text{CR}} \equiv \mathcal{C}_{\text{B/C}} \Rightarrow$ back to (iii)

Bayesian with $P(\theta_{\text{CR}} | \bar{p}) = \mathcal{L}(\bar{p} | \theta_{\text{CR}}) \times \Pi(\theta_{\text{CR}})$



A. Cuoco et al., Phys. Rev. Lett. **118** (2017) 191102
 A. Cuoco et al, Phys. Rev. **D99** (2019) 103014

A. Reinert & M.W. Wise, JCAP **1801** (2018) 055

M.-Y. Cui et al., Phys. Rev. Lett. **118** (2017) 191101

2) Spectral anomalies – Frequentist vs Bayesian

(a) Model M_i is right or wrong depending on its reduced χ^2 per d.o.f.

Frequentist approach – models M_1 and M_2 are both true

Maximum likelihood method if $M_1 \subset M_2$

(b) Generate **Mock data sets** with best-fit model M_1 , fit them and look for the Monte Carlo distribution of $\Delta\chi^2 = \chi_1^2 - \chi_2^2$.

Conclude from $P(\Delta\chi^2 | M_1)$

Mock data to be generated – lengthy procedure

Frequentist approach – may be the best one

(c) Use the **Akaike information criterion** (AIC) and compute the difference $\Delta\text{AIC} = \text{AIC}_2 - \text{AIC}_1$.

$\text{AIC} = 2p - 2 \ln \mathcal{L}_{\max}$ with p parameters

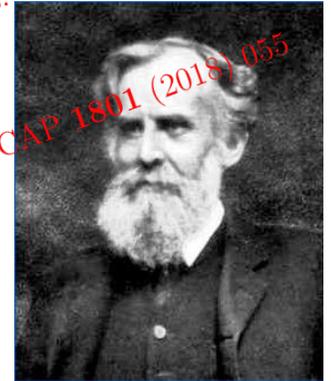
$\text{AIC}_{\text{cor}} = \text{AIC} + \frac{2p^2 + 2p}{N - p - 1}$ with N data points

In our example, N is large and $\Delta\text{AIC} = 4 - \Delta\chi^2$

$$\frac{P(M_1)}{P(M_2)} = e^{\Delta\text{AIC}/2}$$

In Cuoco et al. (2019) $P(\text{no DM})/P(\text{DM}) = 1.3 \times 10^{-2}$

A. Cuoco et al., Phys. Rev. Lett. **118** (2017) 191102
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A. Reinert & M.W. Winkler, JCAP **1801** (2018) 055



(d) **Bayesian** approach in the spirit of method **iv** with nuisance parameters. In this example $\theta_1 = \{\theta_{\text{CR}}\}$ while $\theta_2 = \{\theta_{\text{CR}}, \langle \sigma_{\text{ann}} v \rangle, m_\chi\}$. Compute the probability $P(\bar{p} | M_i)$ and use **Bayes factor** K .

$$P(\theta_1 | \bar{p}, M_1) = \frac{\mathcal{L}(\bar{p} | \theta_1, M_1) \Pi(\theta_1 | M_1)}{P(\bar{p} | M_1)}$$

$$P(\bar{p} | M_1) = \int \mathcal{L}(\bar{p} | \theta_1, M_1) \Pi(\theta_1 | M_1) d\theta_1$$

$$\frac{P(M_2 | \bar{p})}{P(M_1 | \bar{p})} = \left\{ K \equiv \frac{P(\bar{p} | M_2)}{P(\bar{p} | M_1)} \right\} \left\{ \frac{P(M_2)}{P(M_1)} \simeq 1 \right\}$$

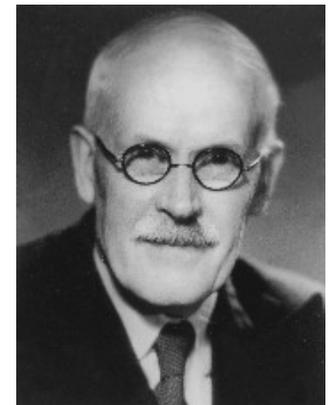
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(e) Use the **Schwarz/Bayes information criterion** (SBIC) and compute the difference $\Delta\text{SBIC} = \text{SBIC}_2 - \text{SBIC}_1$.

Jeffreys' scale

	$2 \ln K$	K	Strength of evidence
$\text{SBIC} = p \ln(N)$	0 to 2	1 to 3	not worth more than a bare mention
In our example,	2 to 6	3 to 20	positive
	6 to 10	20 to 150	strong
$\frac{P(M_1 \bar{p})}{P(M_2 \bar{p})} = e^{\Delta\text{SBIC}}$	>10	>150	very strong



In Cuoco et al. conditions of CRs [46]. The logarithmic Bayes factor value ($2 \ln K$) of the DM component is found to be about 11 – 54 for the three cross section parameterizations used. The best fit DM

- (d) **Bayesian** approach in the spirit of method **iv** with nuisance parameters. In this example $\theta_1 = \{\theta_{\text{CR}}\}$ while $\theta_2 = \{\theta_{\text{CR}}, \langle \sigma_{\text{ann}} v \rangle, m_\chi\}$. Compute the probability $P(\bar{p} | M_i)$ and use **Bayes factor** K .

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- (e) Use the **Schwarz/Bayes information criterion** (SBIC) and compute the difference $\Delta\text{SBIC} = \text{SBIC}_2 - \text{SBIC}_1$.

$$\text{SBIC} = p \ln(N) - 2 \ln \mathcal{L}_{\text{max}} \text{ with } p \text{ parameters}$$

$$\text{In our example, } N = 158 \text{ and } \Delta\text{SBIC} = 10.1 - \Delta\chi^2$$

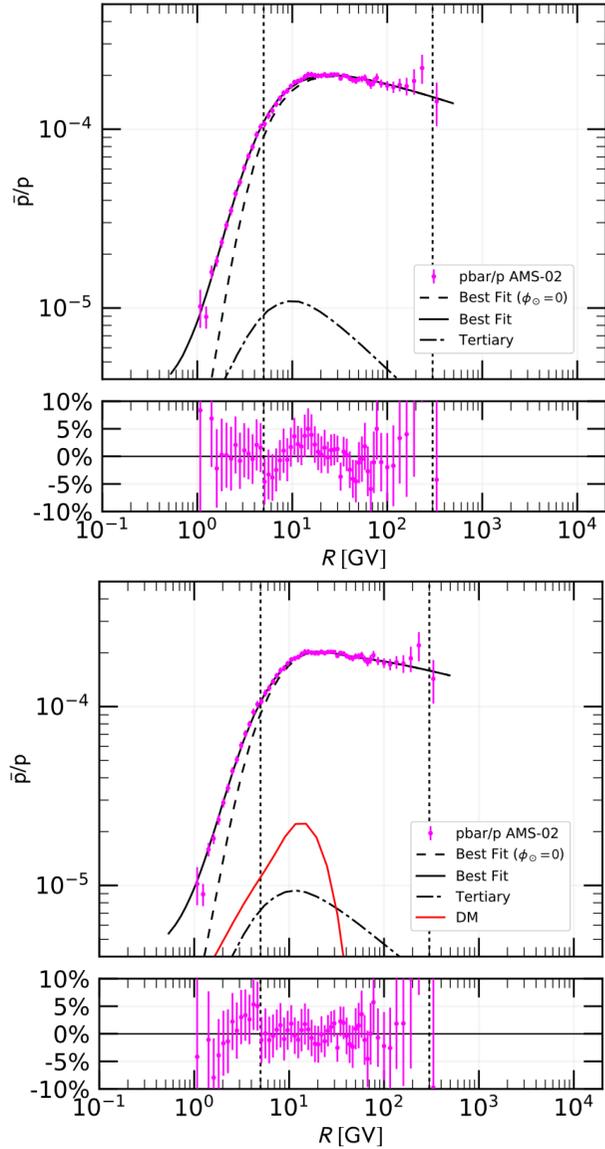
$$\frac{P(M_1 | \bar{p})}{P(M_2 | \bar{p})} = e^{\Delta\text{SBIC}/2}$$

In Cuoco et al. (2019) $P(\text{no DM})/P(\text{DM}) = 0.28$ (no comment)



3) A few recent examples

A. Cuoco et al, Phys. Rev. **D99** (2019) 103014

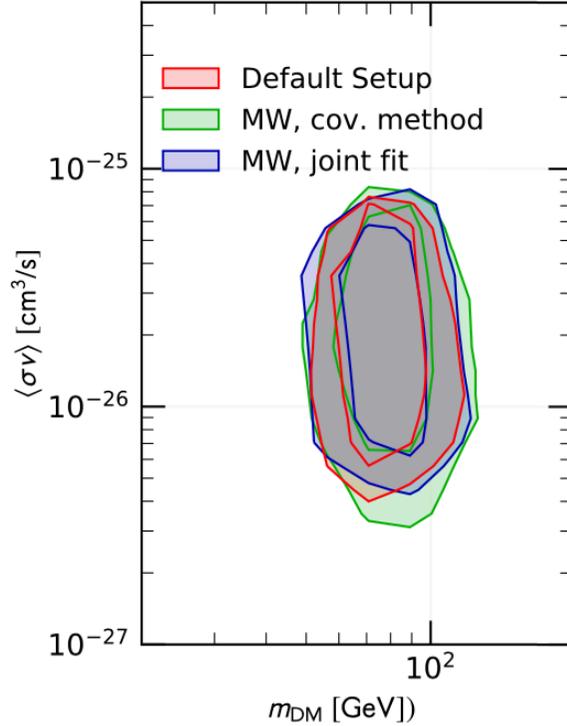


parameter	default setup	
XS parametrization	Param. MW	
DM	incl.	excl.
γ_1	$1.71^{+0.02}_{-0.25}$	$1.72^{+0.04}_{-0.12}$
$\gamma_{1,p}$	$1.78^{+0.003}_{-0.19}$	$1.75^{+0.03}_{-0.10}$
γ_2	$2.41^{+0.03}_{-0.002}$	$2.38^{+0.01}_{-0.02}$
$\gamma_{2,p}$	$2.45^{+0.03}_{-0.002}$	$2.42^{+0.01}_{-0.02}$
R_0 , [MV]	6950^{+330}_{-1640}	7380^{+910}_{-1450}
s_0	$0.38^{+0.06}_{-0.04}$	$0.34^{+0.05}_{-0.04}$
D_0 , [10^{28} cm ² /s]	$5.43^{+0.45}_{-3.17}$	$2.90^{+1.33}_{-1.21}$
δ	$0.38^{+0.01}_{-0.03}$	$0.42^{+0.02}_{-0.01}$
v_A , [km/h]	$18.0^{+2.1}_{-1.4}$	$16.2^{+1.0}_{-2.5}$
$v_{0,c}$, [km/h]	$0.08^{+9.09}_{-0.08}$	$0.52^{+2.32}_{-0.51}$
z_h , [kpc]	$6.45^{+0.30}_{-4.26}$	$3.58^{+2.36}_{-1.52}$
$\log(m_{DM}/[\text{GeV}])$	$1.89^{+0.03}_{-0.08}$	
$\log(\langle\sigma v\rangle/[\text{s}/\text{cm}^3])$	$-26.16^{+0.78}_{-0.04}$	
δ_2		
R_1 , [GV]		
$\varphi_{SM,AMS-02,p,He}$, [MV]	616^{+71}_{-72}	625^{+55}_{-85}
$\varphi_{SM,AMS-02,\bar{p}}$, [MV]	604^{+112}_{-114}	561^{+135}_{-112}
$\chi^2_{AMS-02,p}$	3.2	2.6
$\chi^2_{AMS-02,He}$	4.0	4.8
$\chi^2_{AMS-02,\bar{p}}$	11.1	22.1
$\chi^2_{\text{Voager},p}$	3.2	3.8
$\chi^2_{\text{Voager},He}$	1.3	1.9
$\chi^2_{\text{CREAM},p}$		
$\chi^2_{\text{CREAM},He}$		
$\chi^2_{\varphi_{SM}}$	0.0	0.4
χ^2_{CR}	22.9	35.6
χ^2/dof	22.9/143	35.6/145
$\Delta\chi^2$		12.7
DM significance		3.1 σ

Method (ii) used to look for DM with $P(p + \alpha + \bar{p}/p | \theta_{CR})$

3) A few recent examples

A. Cuoco et al, Phys. Rev. **D99** (2019) 103014



	χ^2		$\Delta\chi^2$
	no DM	with DM	
Param. MD	34.2	22.2	12.0
Param. MW	35.6	22.9	12.7
Cov-mat (Param. MD)	34.2	22.0	12.2
Cov-mat (Param. MW)	33.9	23.0	10.9
Joint fit (Param. MW)	825.2	814.5	10.7

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3) A few recent examples

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rigidity cut [GV]	χ^2/ndf		$\Delta\chi^2$	DM significance
	excl. DM	incl. DM		
5	35.6/145 = 0.245	22.9/143 = 0.160	12.7	3.1 σ
3	52.7/160 = 0.329	34.2/158 = 0.216	18.5	3.9 σ
2	68.2/172 = 0.396	57.1/170 = 0.336	11.1	2.9 σ
1	105.4/182 = 0.579	105.6/180 = 0.586	-0.2	-

Method (e), i.e., Schwarz/Bayes information criterion, yields

$$\frac{P(\text{no DM} | \text{data})}{P(\text{DM} | \text{data})} = e^{\Delta\text{SBIC}/2} = 0.28$$

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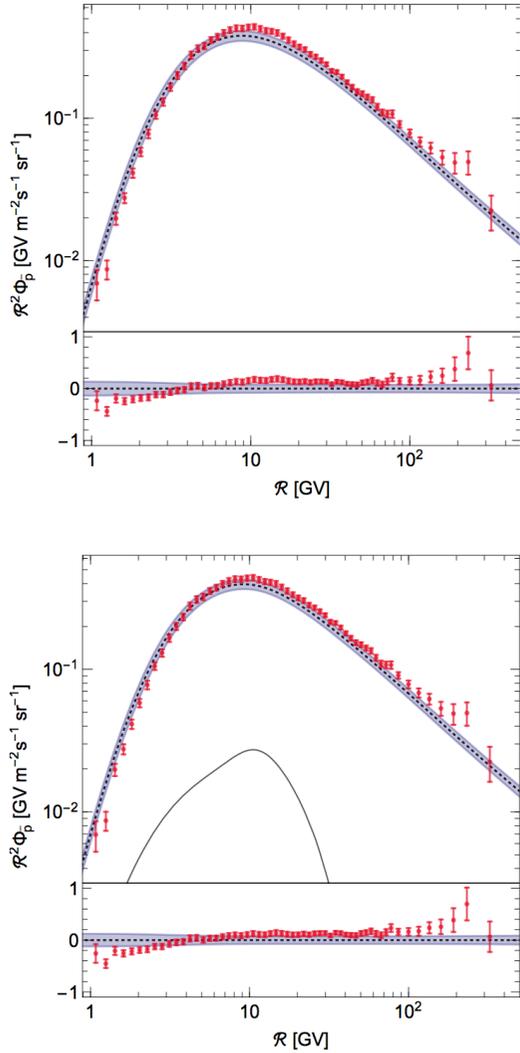
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Mixed methods (ii) + (iii) used to look for DM

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Uncertainties on XS treated with $\sum_{nm}^{\bar{p}} \text{source} + \sum_{nm}^{B/C} \text{source}$

$P(\bar{p} + B/C + \text{AMS/PAMELA} | \theta_{CR} + \phi_{\bar{p}}^F)$



Best Fit	B/C (w/o break)	B/C (w/ break)	\bar{p} (w/ break)	B/C + \bar{p} (w/ break)
K_0 [$\frac{\text{kpc}^2}{\text{Gyr}}$]	$39.6 \cdot L_{4.1}$	$34.3 \cdot L_{4.1}$	$39.5 \cdot L_{4.1}$	$32.5 \cdot L_{4.1}$
δ	0.479	0.507	0.446	0.506
V_a [$\frac{\text{km}}{\text{s}}$]	0	0	$59.7 \cdot \sqrt{L_{4.1}}$	$15.6 \cdot \sqrt{L_{4.1}}$
V_c [$\frac{\text{km}}{\text{s}}$]	0	1.3	0	0
$\Delta\delta$		0.157	0.157	0.157
\mathcal{R}_b [GV]		275	275	275
s		0.074	0.074	0.074
ϕ_0 [GV]	0.72	0.72	0.72	0.72
ϕ_1 [GV]			0.66	0.84
$\chi_{B/C}^2$ (67 bins)	64.2	48.0		55.1
$\chi_{\bar{p}}^2$ (57 bins)			21.3	47.9
$\chi_{\text{AMS/PAM}}^2$ (17 bins)			10.9	12.6

Channel	$\bar{b}\bar{b}$	WW
m_{DM}	78.7 GeV	85.2 GeV
$\langle\sigma v\rangle$ [$\frac{\text{cm}^3}{\text{s}}$]	$0.91 \cdot 10^{-26}$	$1.0 \cdot 10^{-26}$
K_0 [$\frac{\text{kpc}^2}{\text{Gyr}}$]	34.0	33.7
L [kpc]	4.1	4.1
δ	0.499	0.500
V_a [$\frac{\text{km}}{\text{s}}$]	15.0	15.1
V_c [$\frac{\text{km}}{\text{s}}$]	0	0
$\Delta\delta$	0.157	0.157
\mathcal{R}_b [GV]	275	275
s	0.074	0.074
ϕ_0 [GV]	0.72	0.72
ϕ_1 [GV]	0.95	0.96
$\Delta\chi^2$	4.7	2.4
p_{local}	0.015 (2.2 σ)	0.061 (1.6 σ)
p_{global}	0.14 (1.1 σ)	0.25 (0.7 σ)

Method (b) used

Mock data $\Rightarrow \Delta\chi^2$ & $P(\Delta\chi^2 | M_1)$

$$\frac{P(\text{no DM} | \text{data})}{P(\text{DM} | \text{data})} = e^{\Delta\text{SBIC}/2} = 13.5 \text{ \& } 42.5$$

Method (e), i.e., Schwarz/Bayes information criterion, yields $\Delta\text{SBIC} = 2 \ln(N) - 2 \Delta \ln \mathcal{L}_{\text{max}} = 9.9 - \Delta\chi^2$

DM is 7.4% ($\bar{b}\bar{b}$) and 2.4% (WW) as probable as no DM!