

Tracing Dark Matter with Stars

Lina Necib

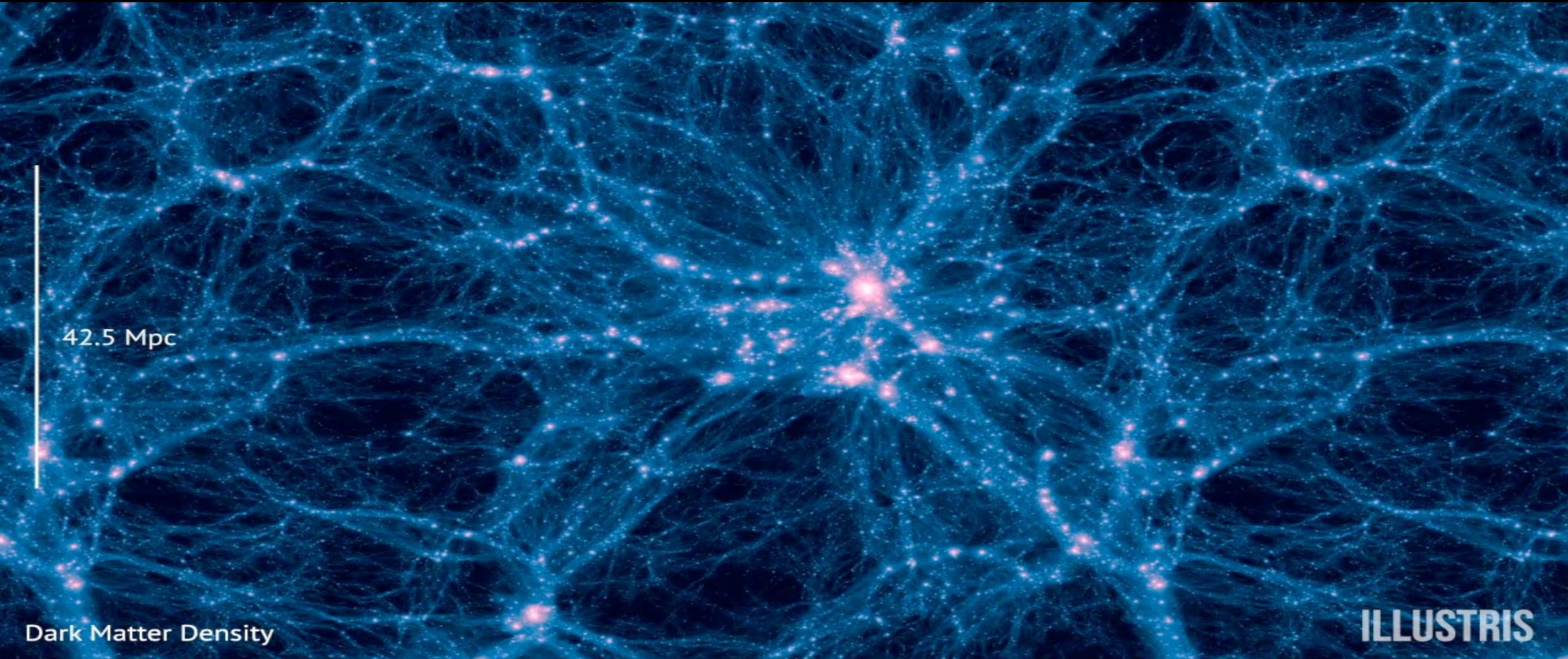
MIT



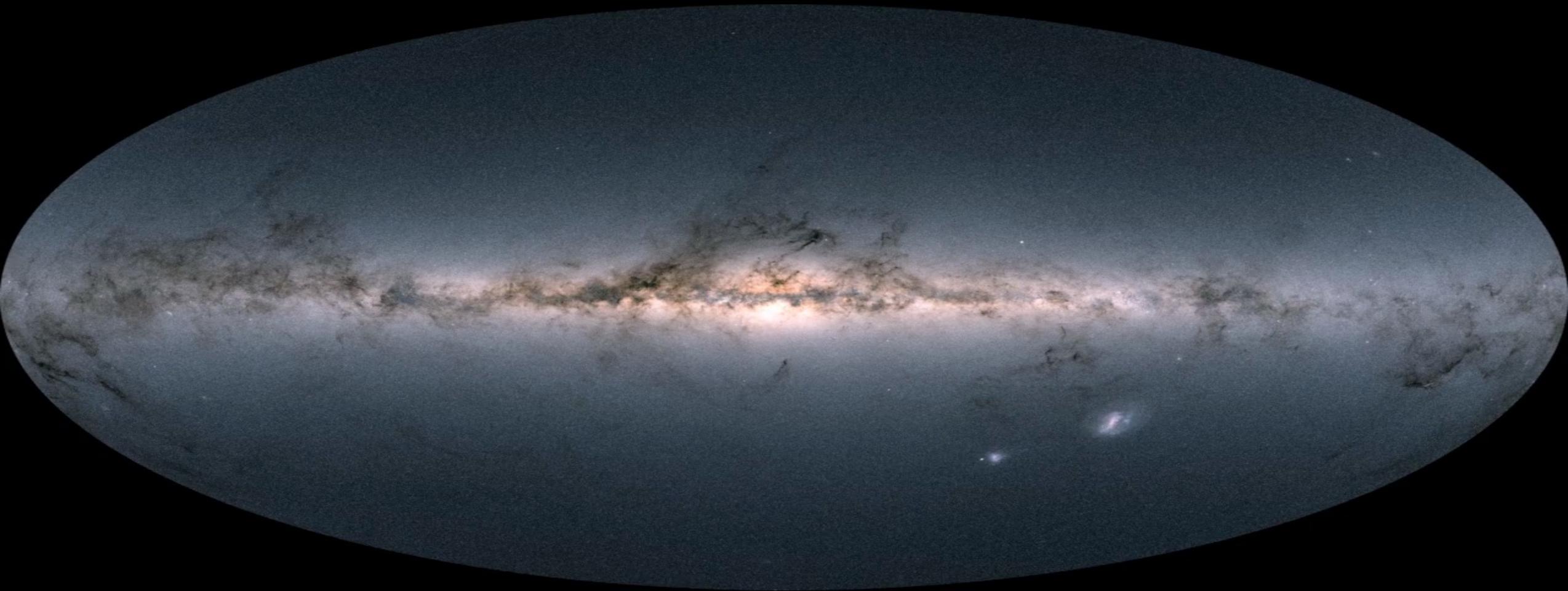
Goal

A Local
Map of
Dark
Matter

Simulations



Gaia



High
Resolution
Simulations

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graph LR; A[High Resolution Simulations] --> B((A Local Map of Dark Matter)); C[Gaia Data] --> B;
```

The diagram features a central orange circle containing the text 'A Local Map of Dark Matter'. To its left is a light gray rounded rectangle with the text 'High Resolution Simulations', and to its right is a dark brown rounded rectangle with the text 'Gaia Data'. Arrows from both rectangles point towards the central circle.

Gaia Data

A Local
Map of
Dark
Matter

This Talk:

How to map out the Dark Matter phase space distribution on Galactic Scales at key locations.

Sun



This Talk:

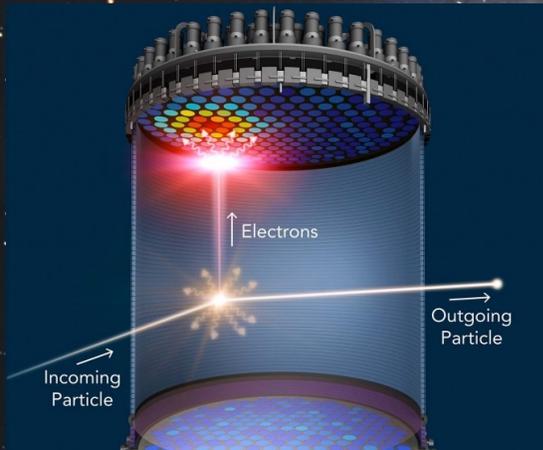
How to map out the Dark Matter phase space distribution on **Galactic Scales** at key locations.

Sun



This Talk:

How to map out the Dark Matter phase space distribution on **Galactic Scales** at key locations.



1. Terrestrial Experiments

This Talk:

How to map out the Dark Matter phase space distribution on **Galactic Scales** at key locations.

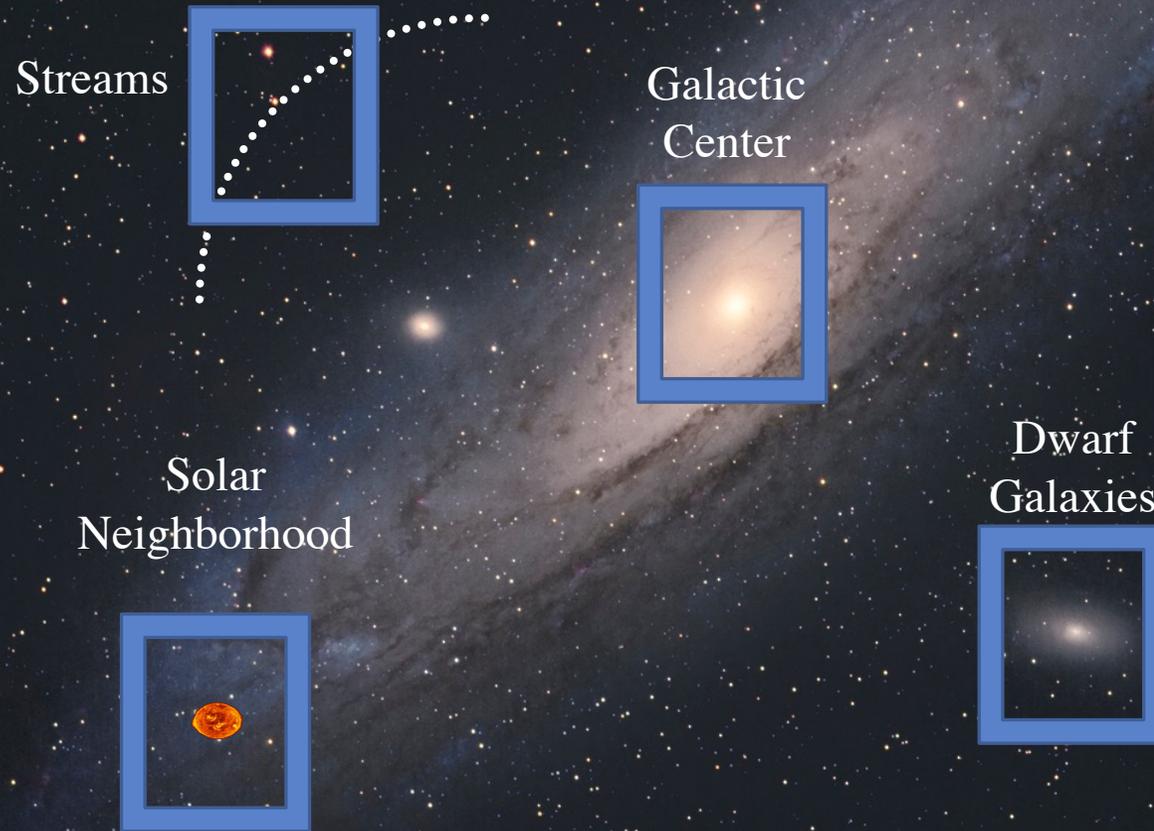
Sun



2. Deviations from Standard
Dark Matter Paradigm at
Galactic Scales

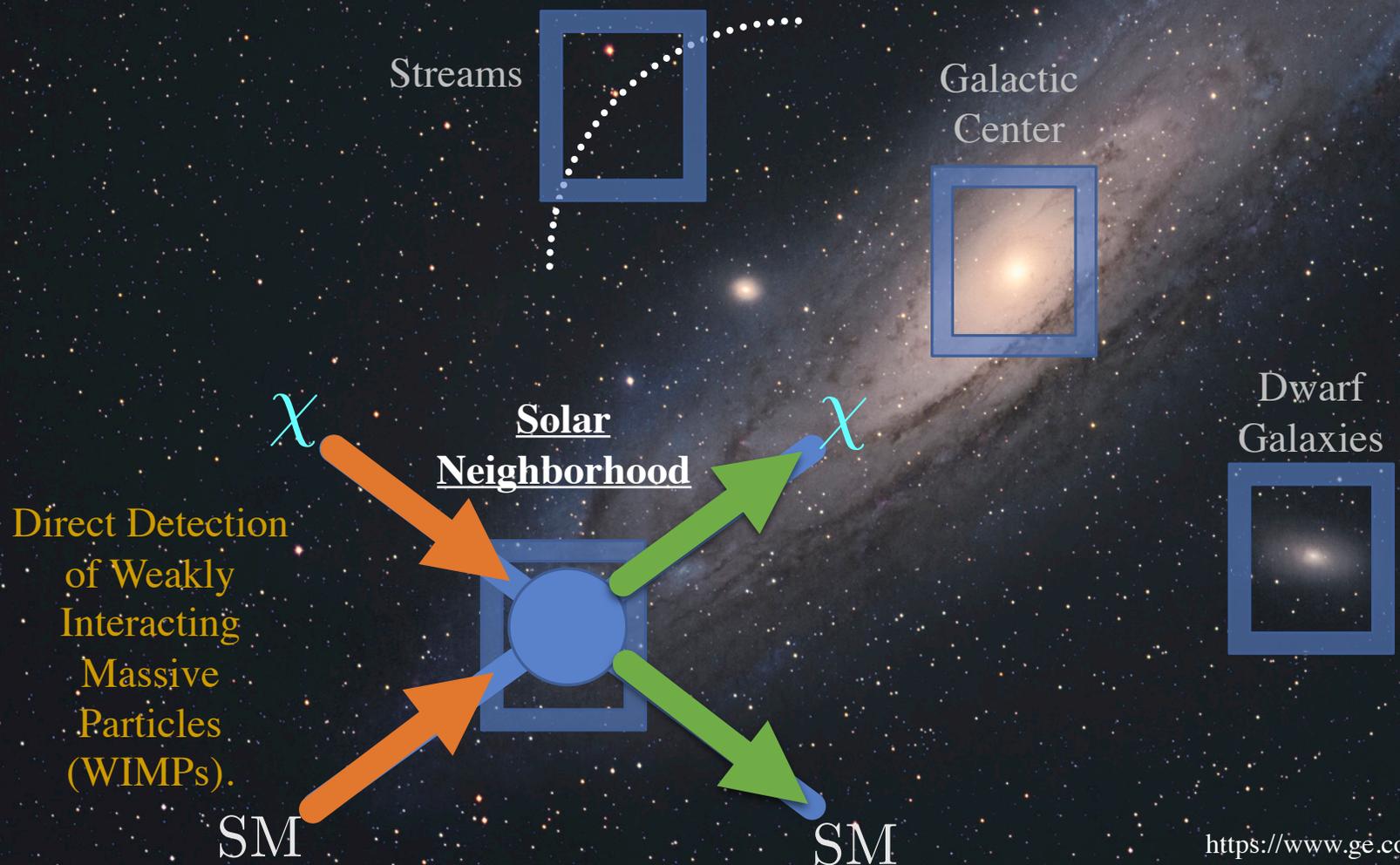
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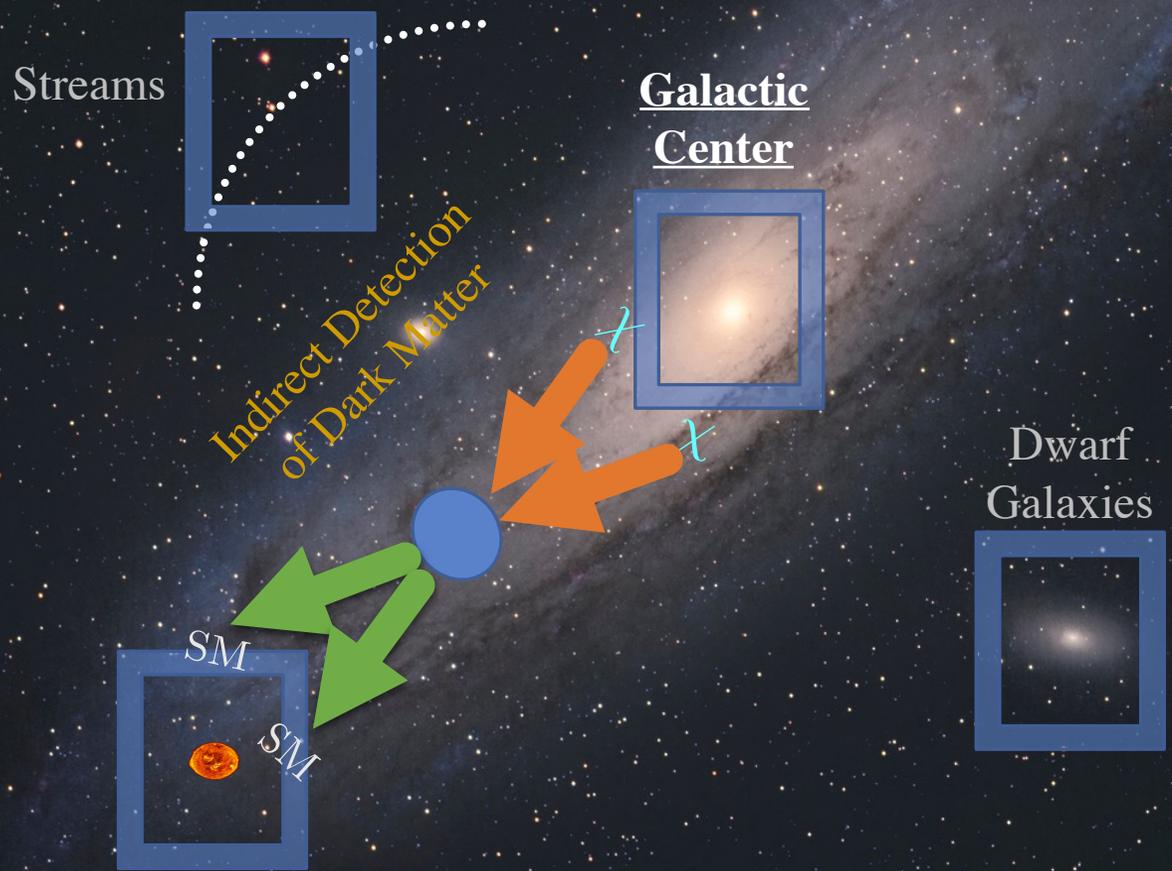
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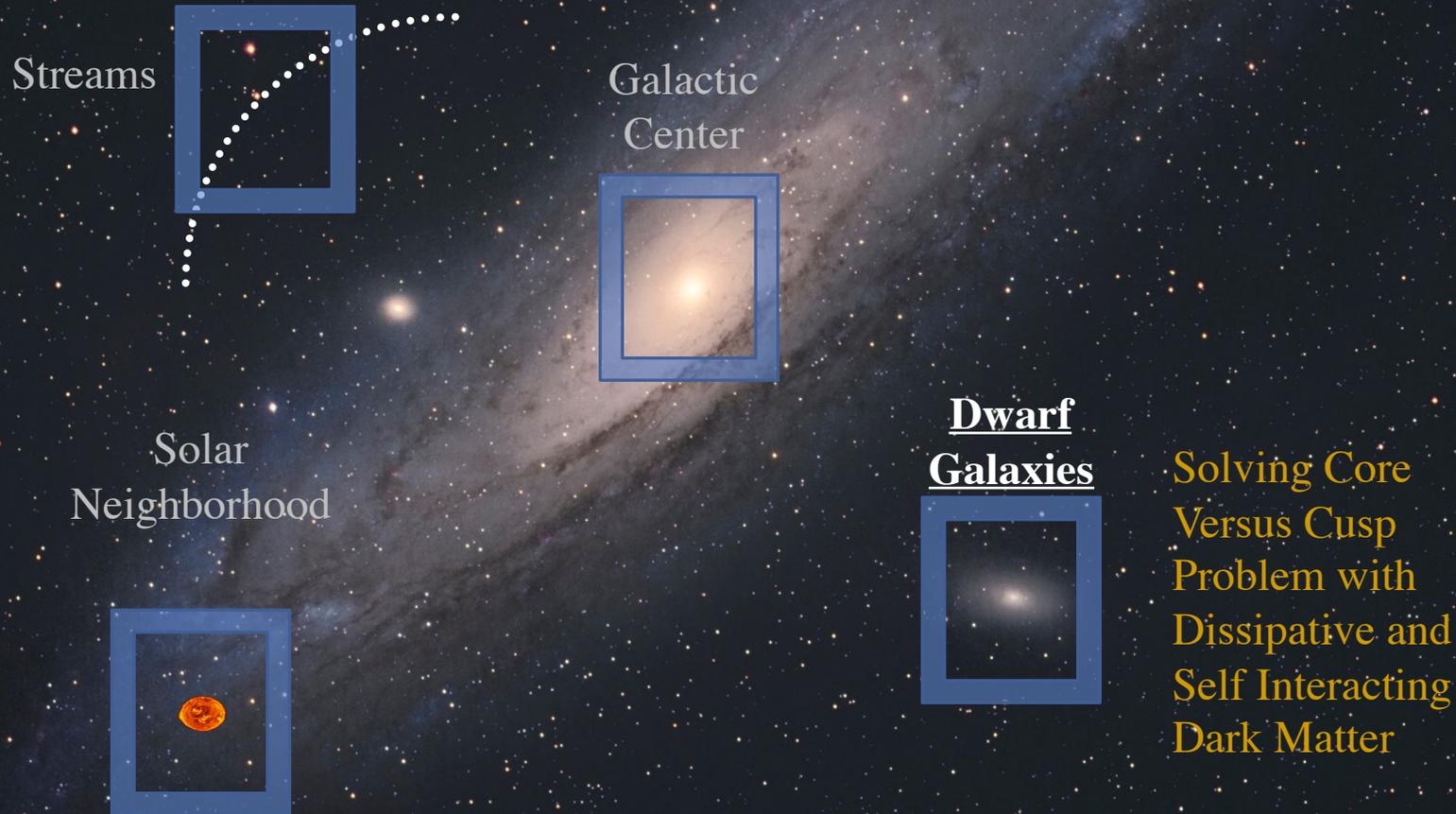
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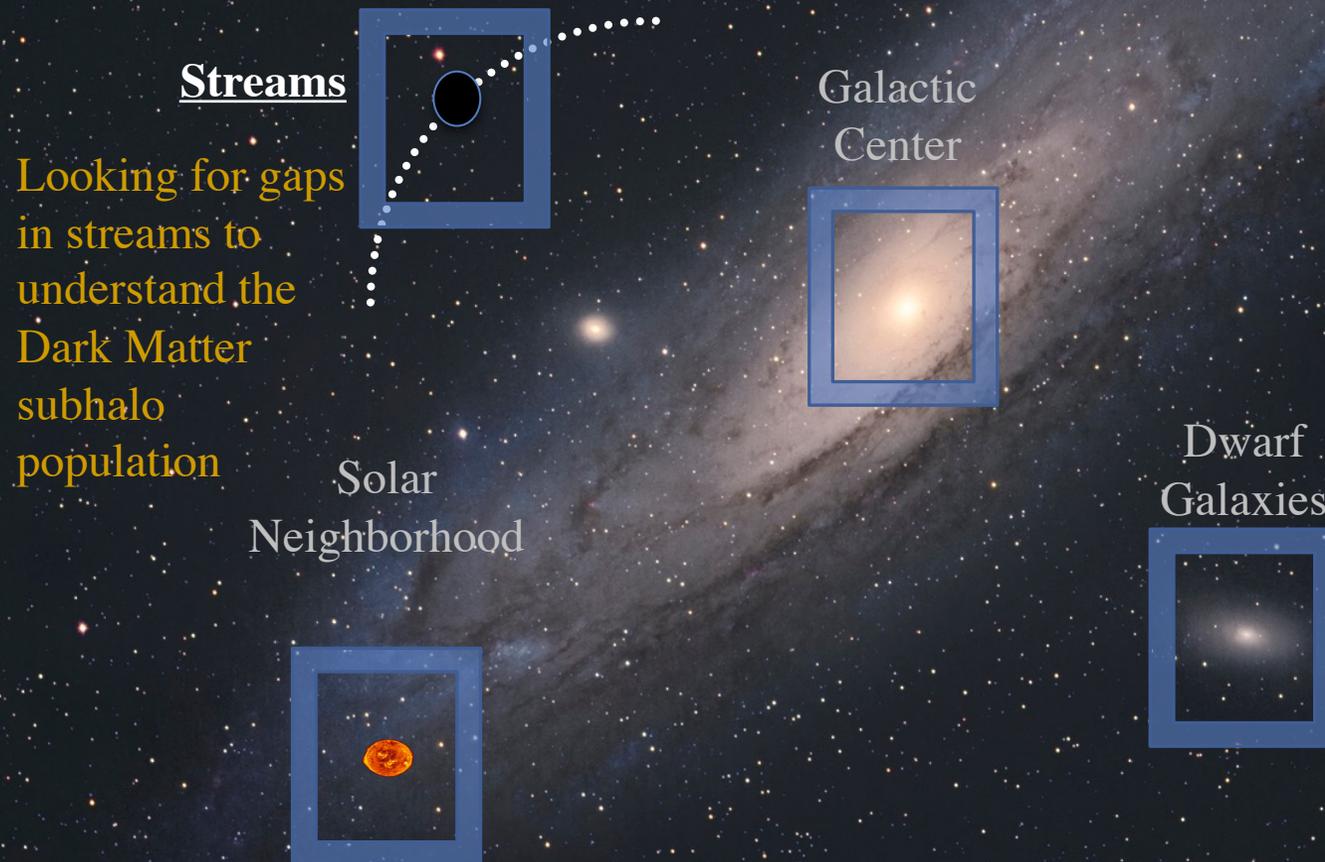
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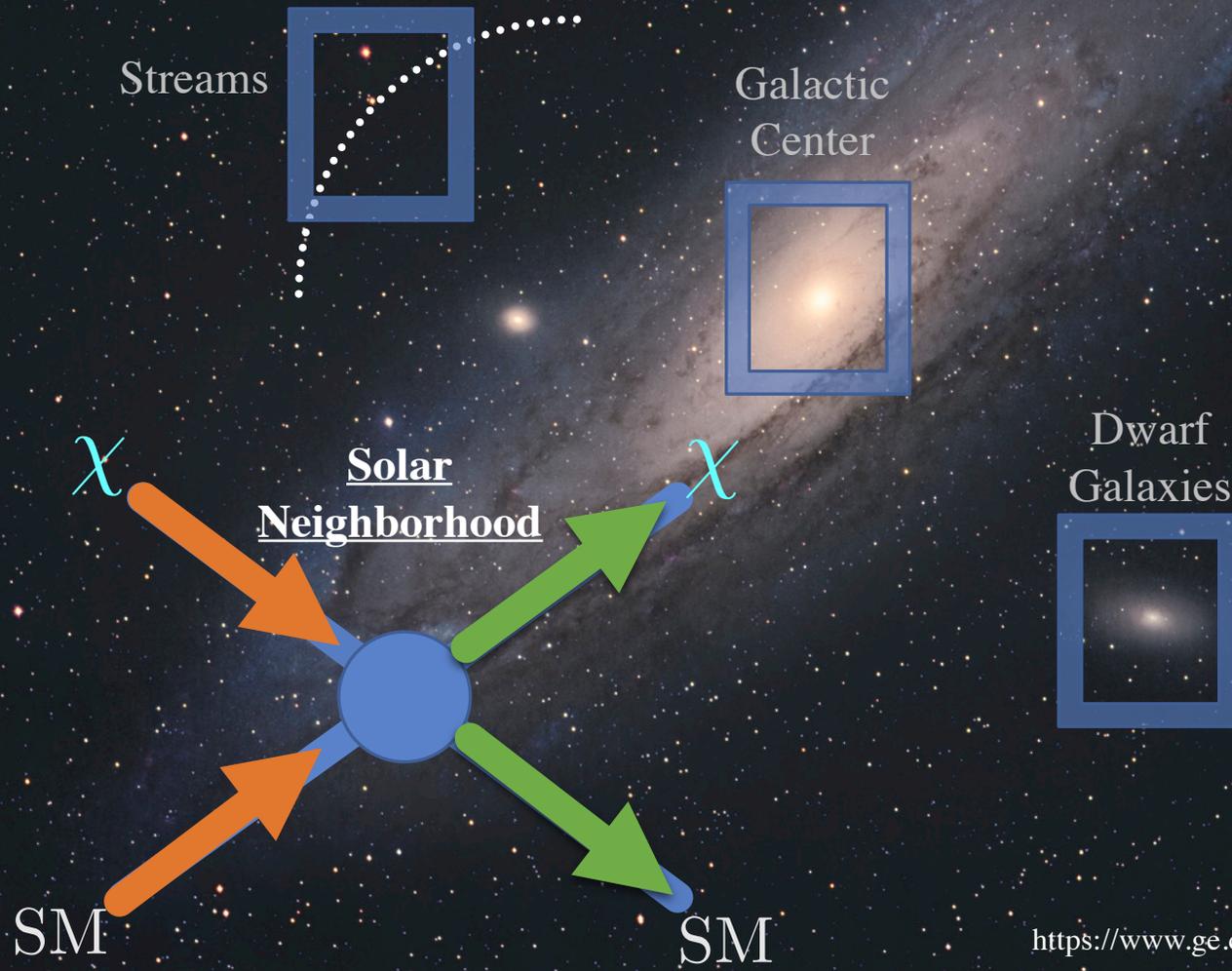


This Talk:

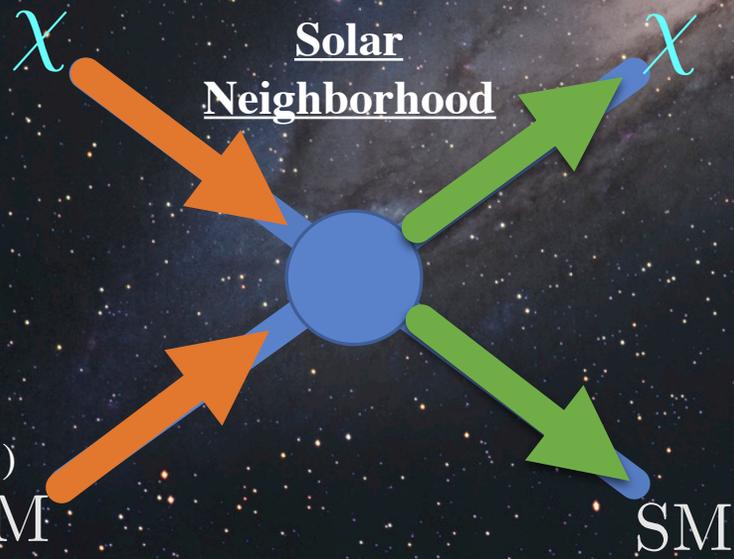
How to map out the Dark Matter phase space distribution on Galactic Scales at key locations.



Dark Matter in the Solar Neighborhood



Dark Matter in the Solar Neighborhood



$$\text{Rate} \propto \rho_{\odot} \int \frac{f(v)}{v} dv$$

ρ_{\odot} : Local Dark Matter Density

$f(v)$: Local Dark Matter Velocity Distribution

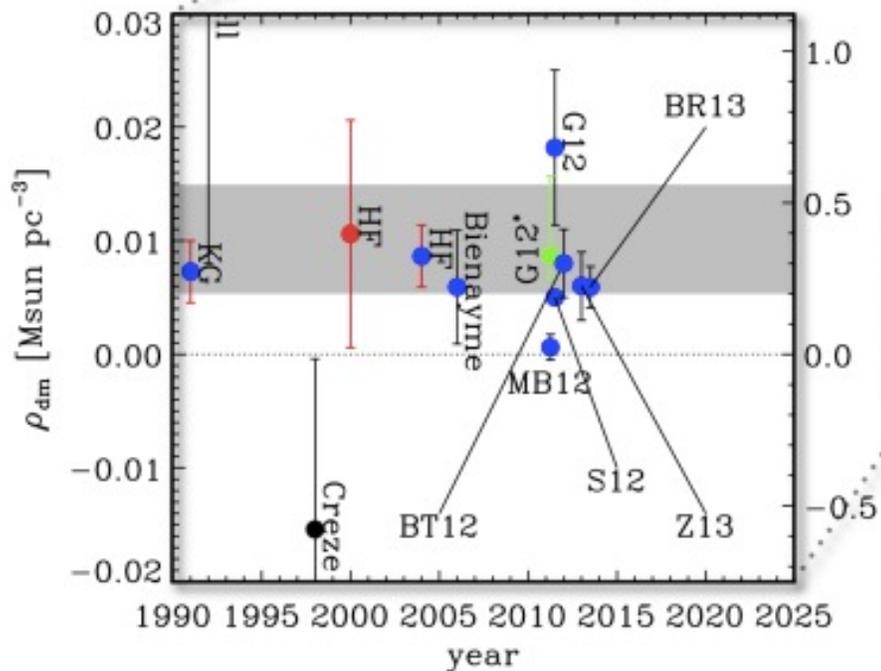
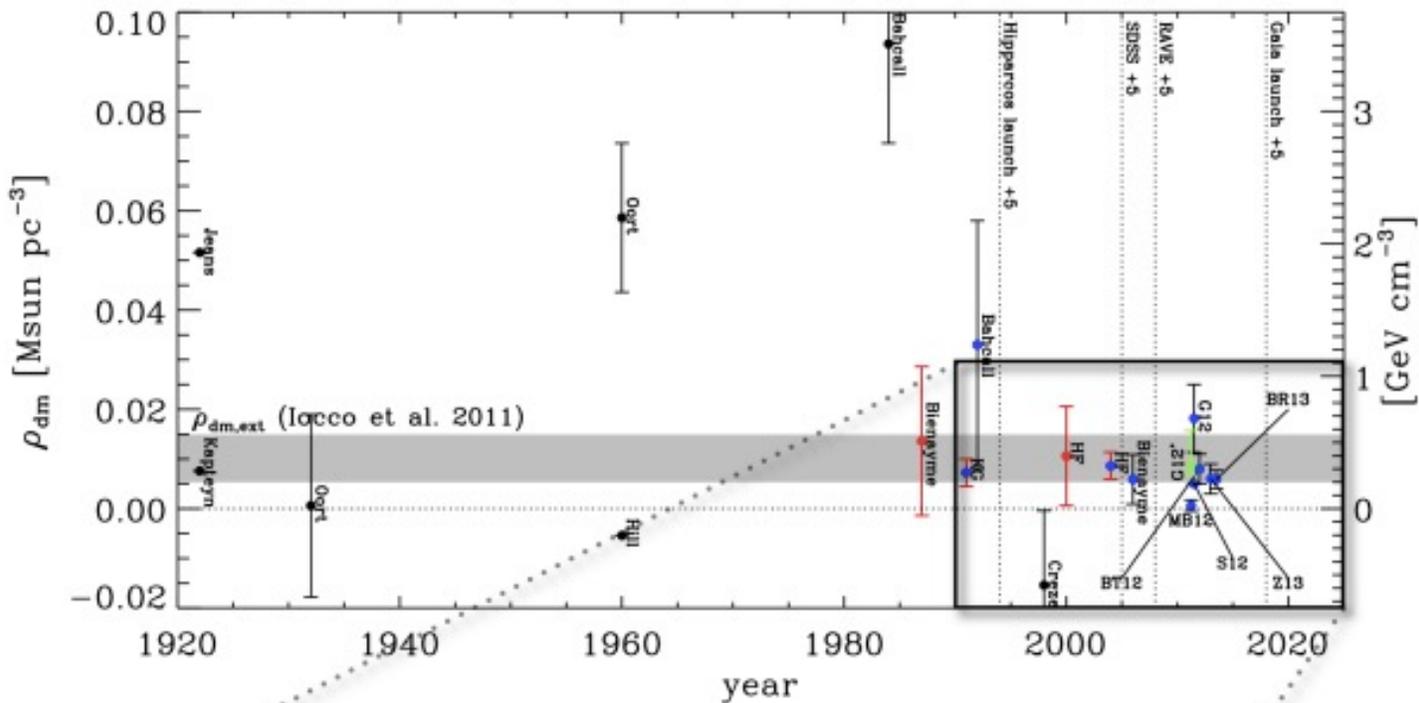
Goodman & Witten (1985)

Freese et al. (1986)

SM

SM

Dark Matter in the Solar Neighborhood



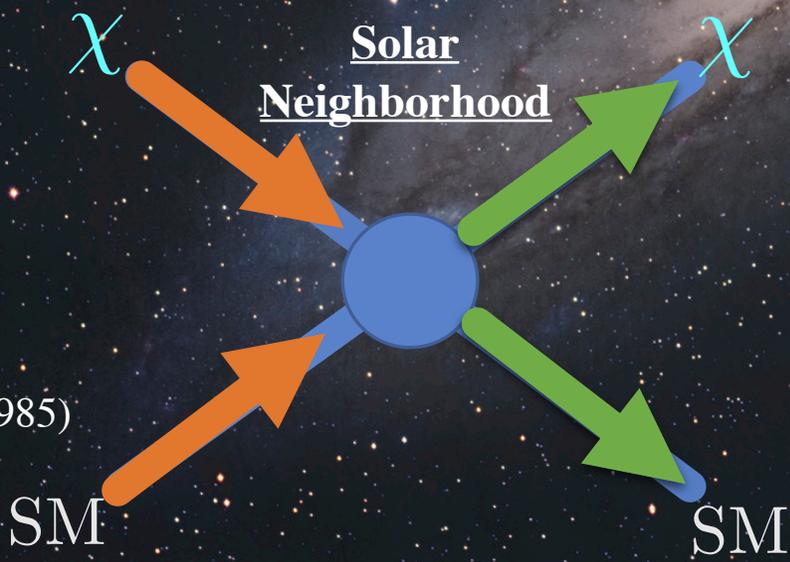
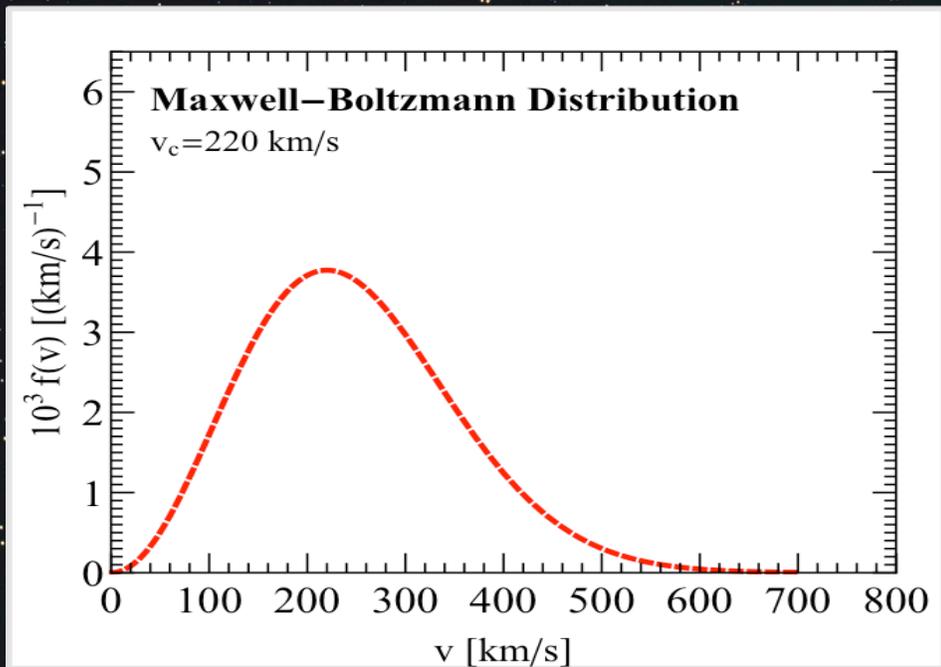
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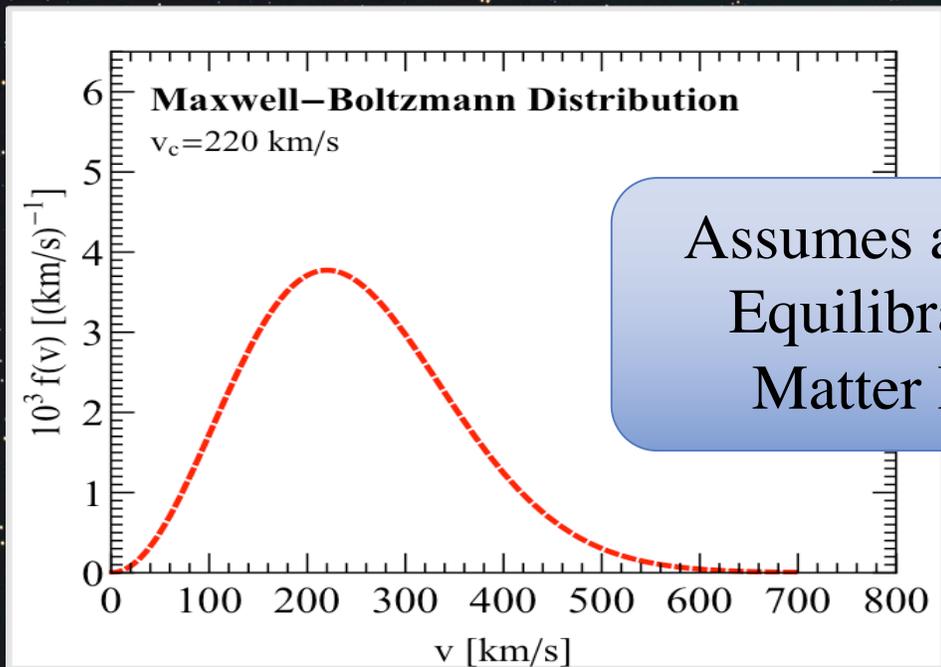
Freese et al. (1986)

Drukier et al. (1987)

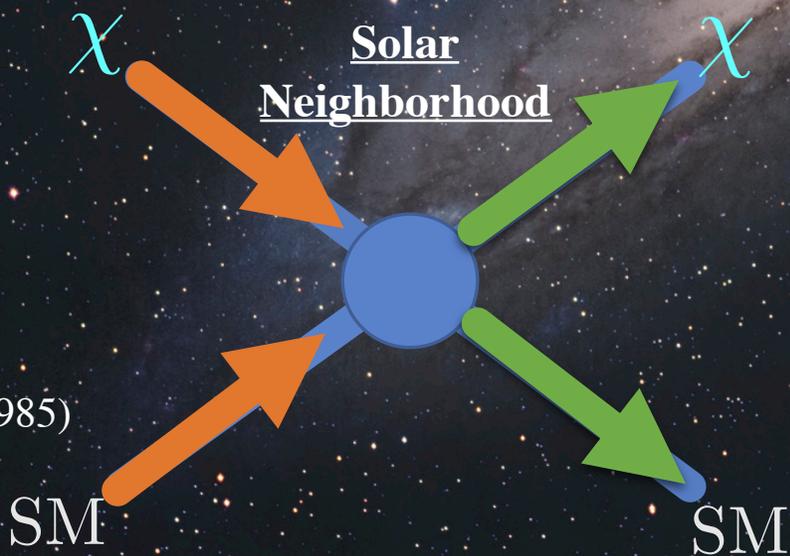
SM

SM

Dark Matter in the Solar Neighborhood



Assumes an Isotropic Equilibrated Dark Matter Potential



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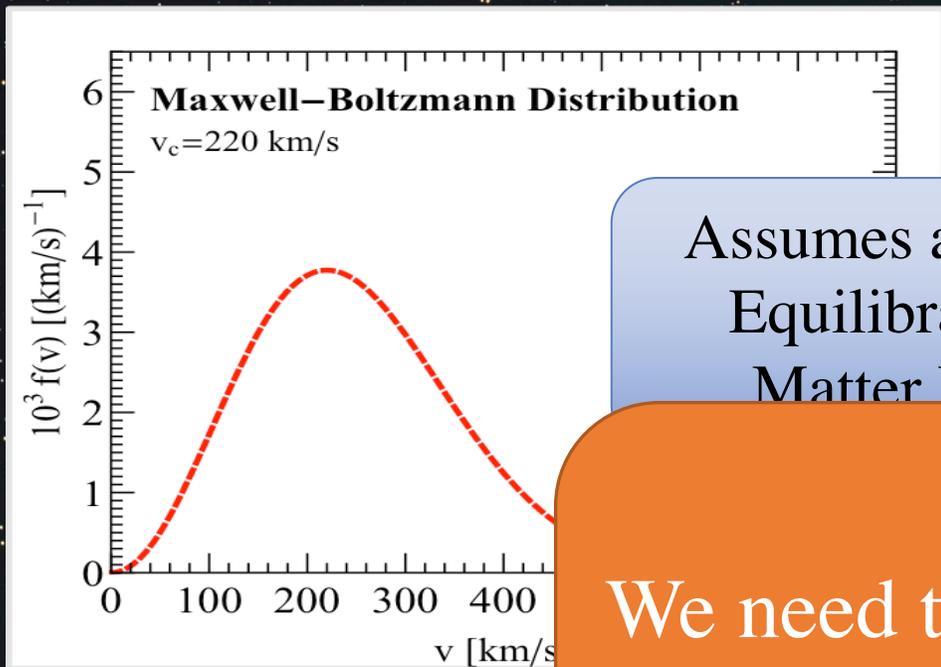
Freese et al. (1986)

Drukier et al. (1987)

SM

SM

Dark Matter in the Solar Neighborhood



Assumes an Isotropic Equilibrated Dark Matter Potential

We need to build an empirical velocity distribution of Dark Matter.

χ

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Goodman & Witten (1985)

Freese et al. (1986)

Drukier et al. (1987)

SM

SM

From
simulations:

Correlate
the dark
matter to
the stars

From *Gaia*:

Measure the
phase space
map of the
stars

Therefore:

Empirically
obtain the
local phase-
space
distribution
of dark
matter

Herzog-Arbeitman, Lisanti, Madau, **Necib** (2018)

Herzog-Arbeitman, Lisanti, **Necib** (2018)

Necib, Lisanti, Belokurov (2019)

Necib, Lisanti, Garrison-Kimmel, et al. (2018)

Building the Dark Matter Map

Strategy: Reconstruct the Dark Matter from the distributions of the stars.



Hopkins et al. (2014)

Wetzel et al. (2016)

Hopkins et al. (2017)

Necib, Lisanti, Garrison-Kimmel, et al. (2018)

Feedback in Realistic Environments (FIRE)

$z=9.9$

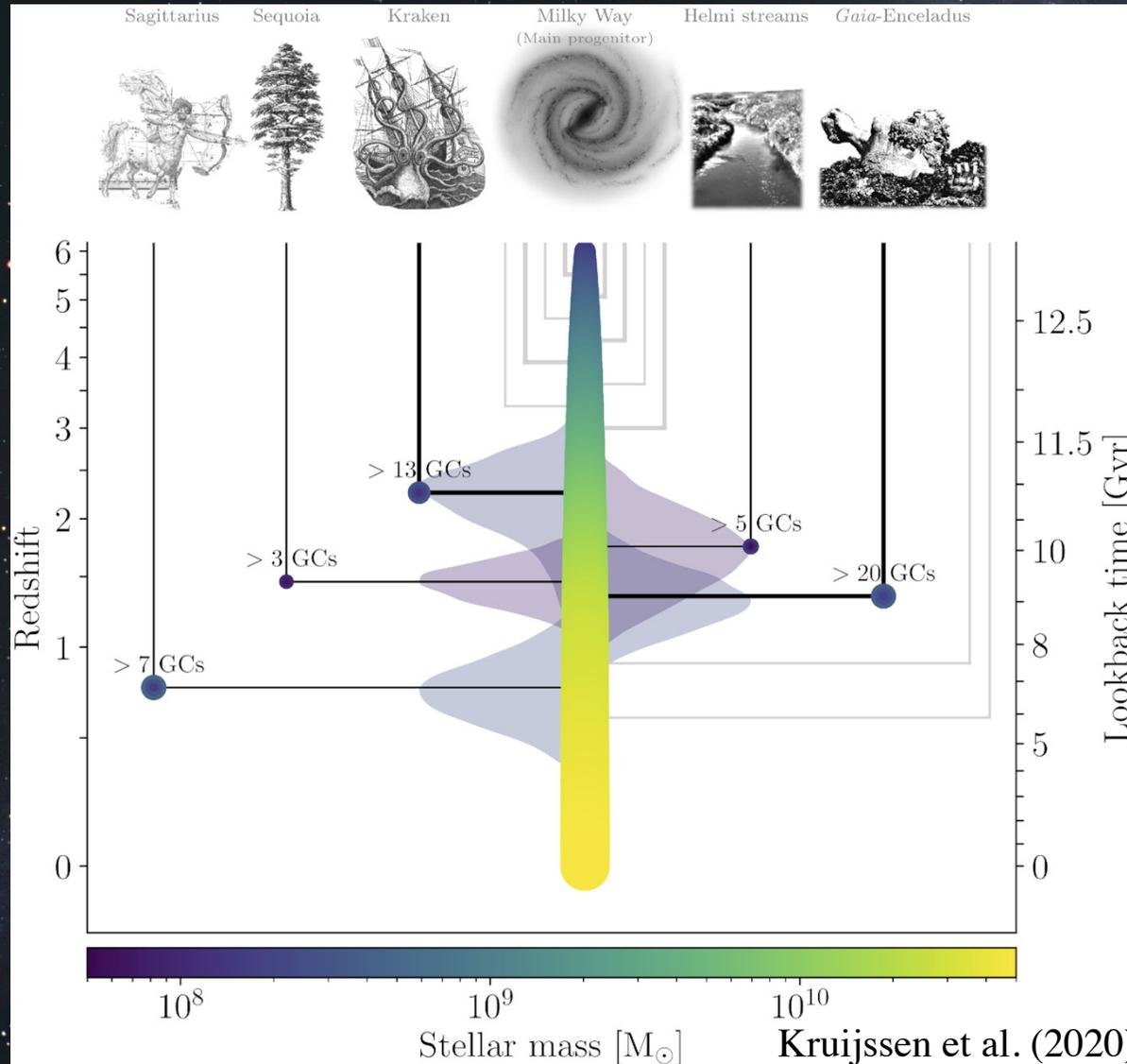
10 kpc



Hopkins et al. (2014)
Wetzel et al. (2016)
Hopkins et al. (2017)

Building the Dark Matter Map

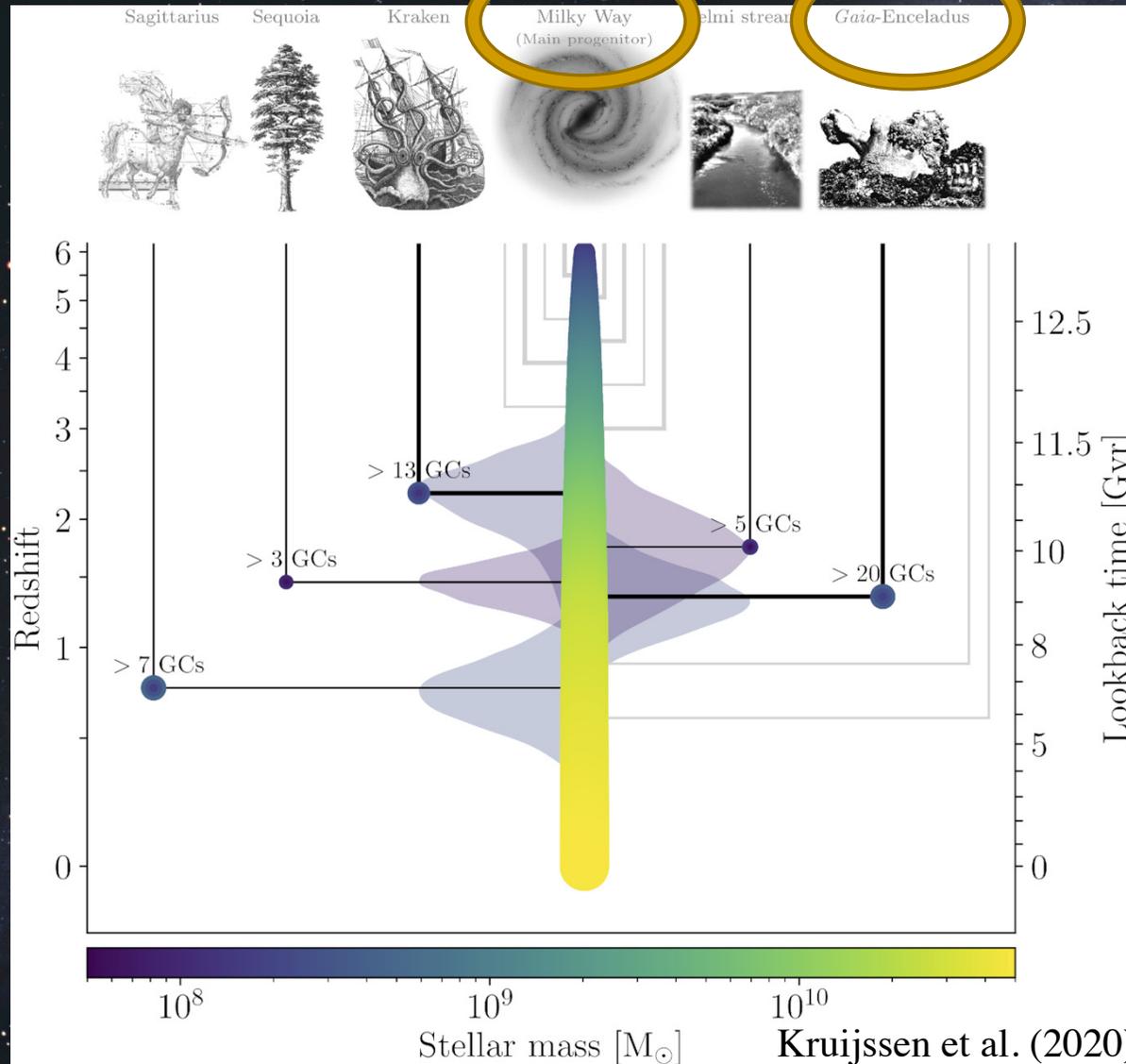
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Belukorov et al. (2018)
Helmi et al. (2018)
Kruijssen et al. (2020)

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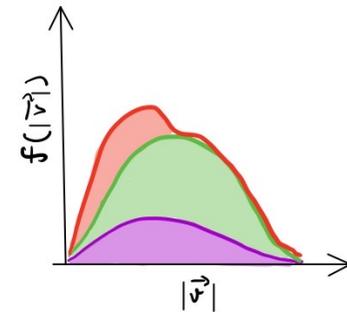
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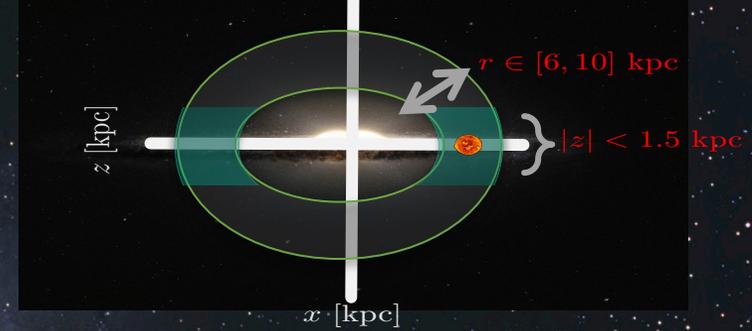
Different Components:

1. Old Halo
2. Sausage/ Gaia Enceladus
3. Nyx



Building the Dark Matter Map

Strategy: Reconstruct the Dark Matter from the distributions of the stars.

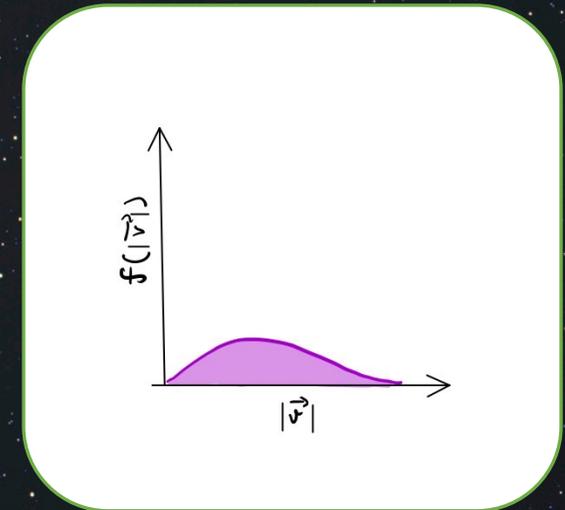
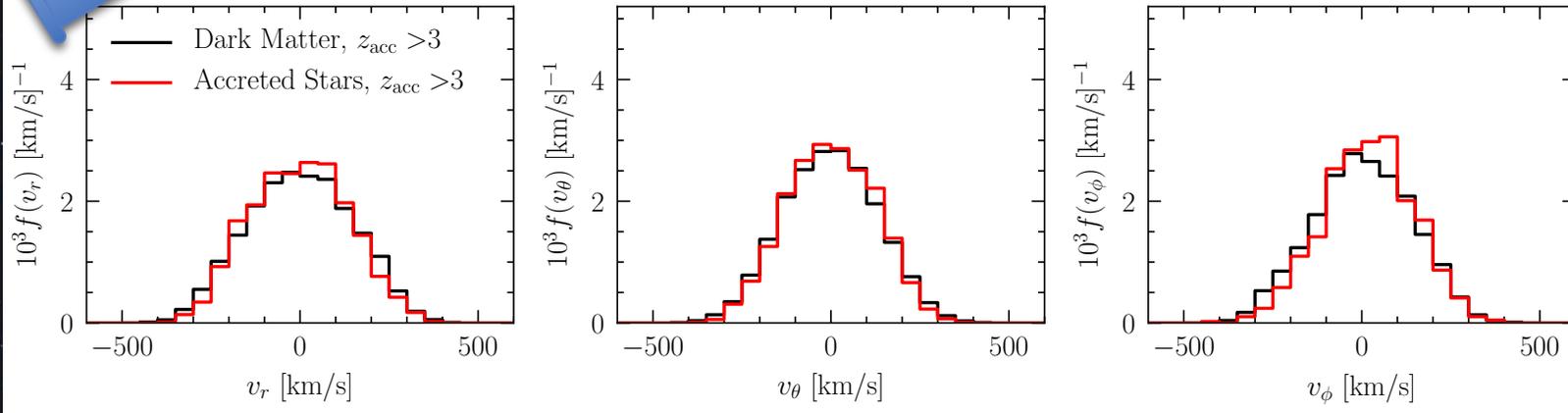


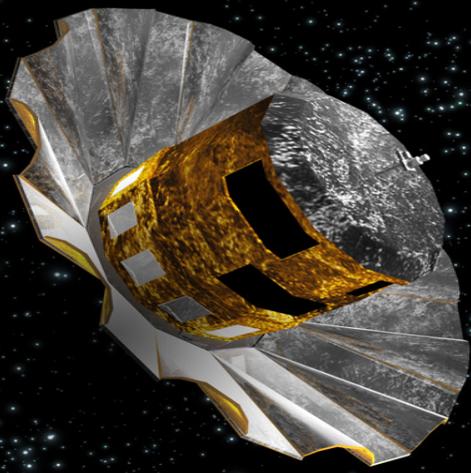
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Simulation

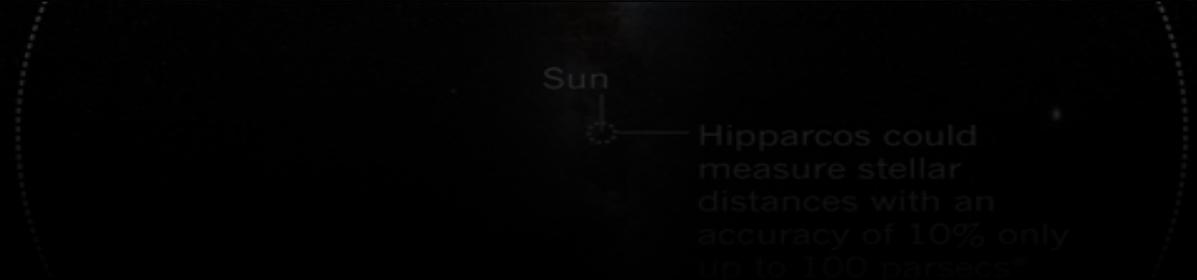
FIRE Host Halo m12i, Relaxed Component

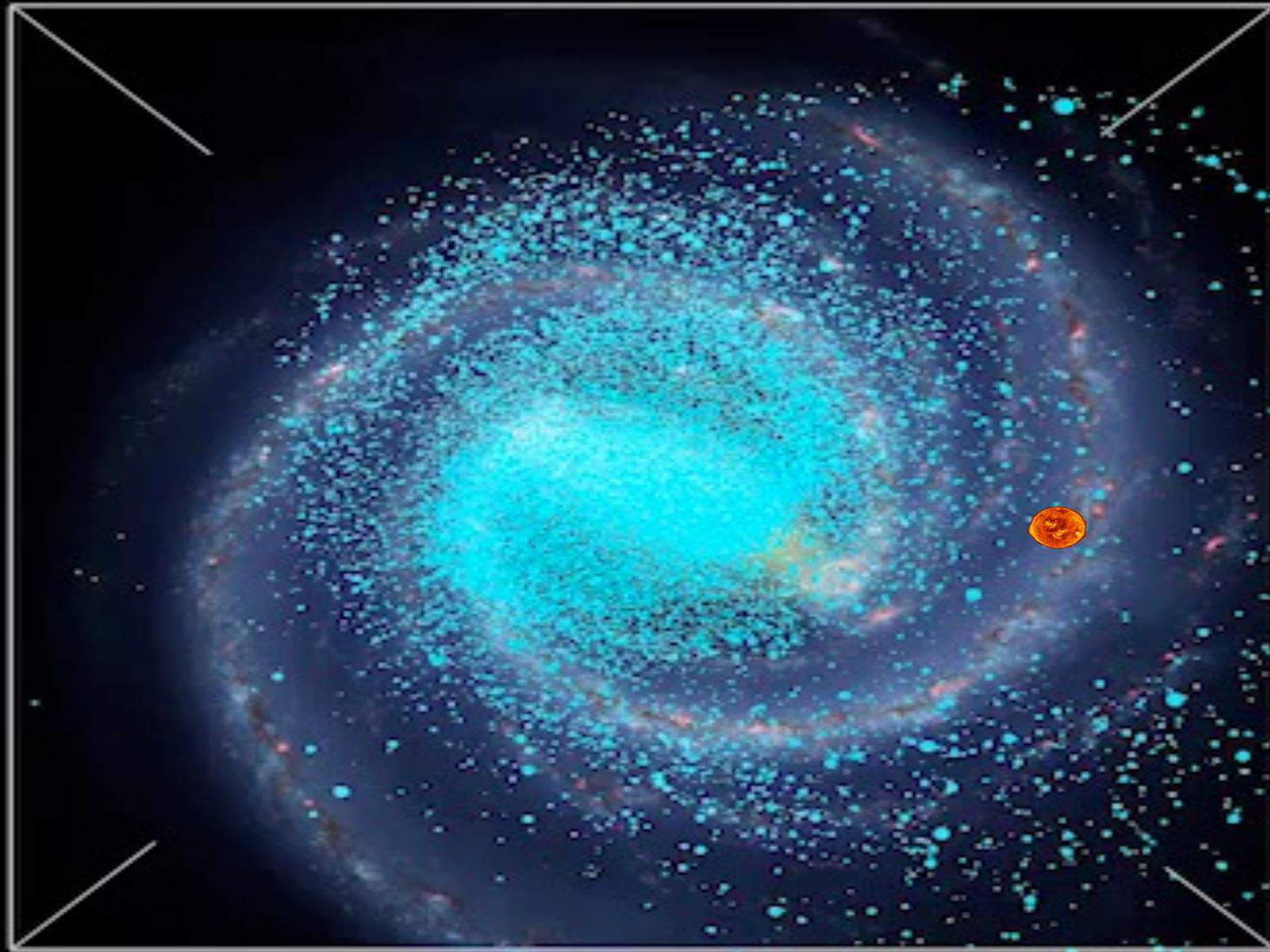




Gaia

- 🎯 Launched December 2013
- 🎯 Goal: Positional and kinematic measurement of 1 billion stars (1% of the Milky Way)
- 🎯 DR2: 7 million radial velocities



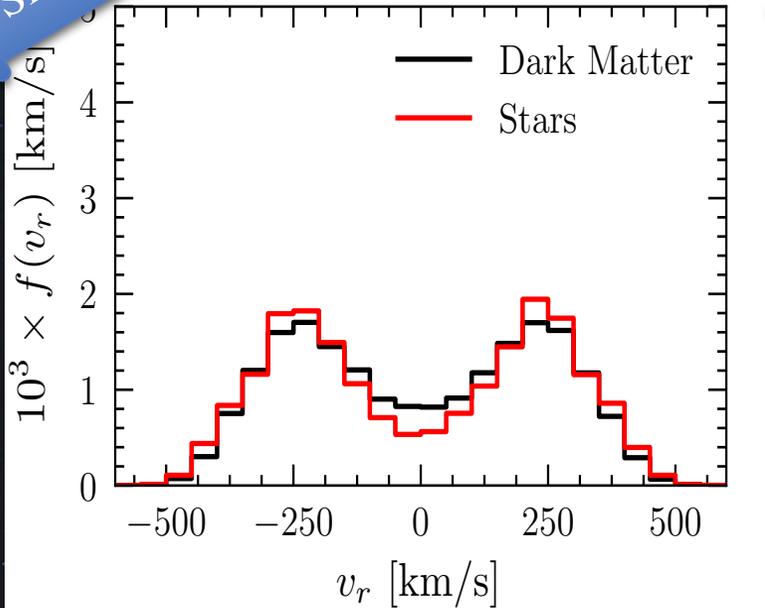


Credit: H.H. Koppelman, A. Villanlobos, A. Helmi

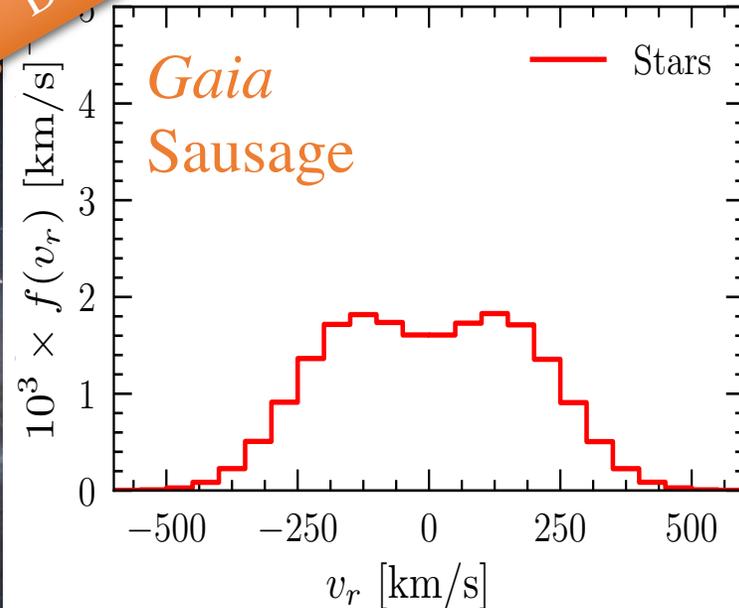
Building the Dark Matter Map

Strategy: Reconstruct the Dark Matter component by component from the distributions of the stars.

Simulation

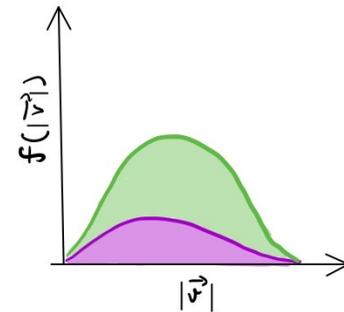


Data



Different Components:

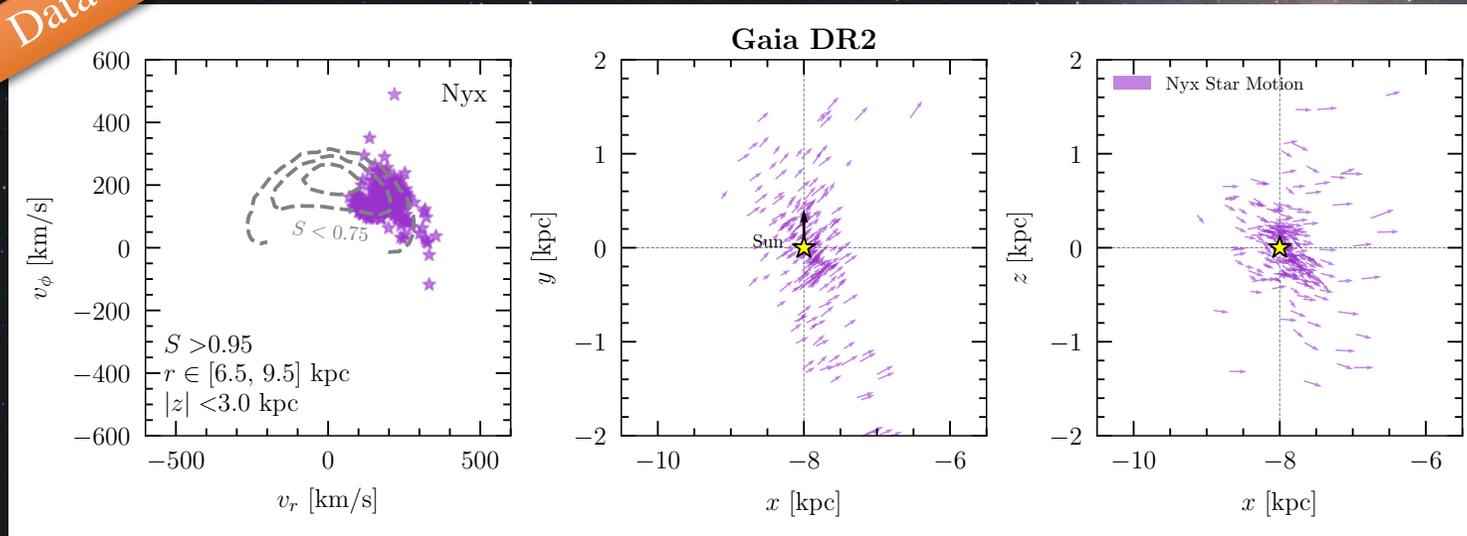
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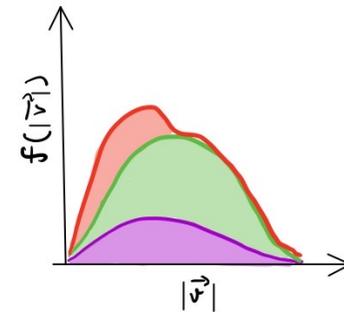
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Data

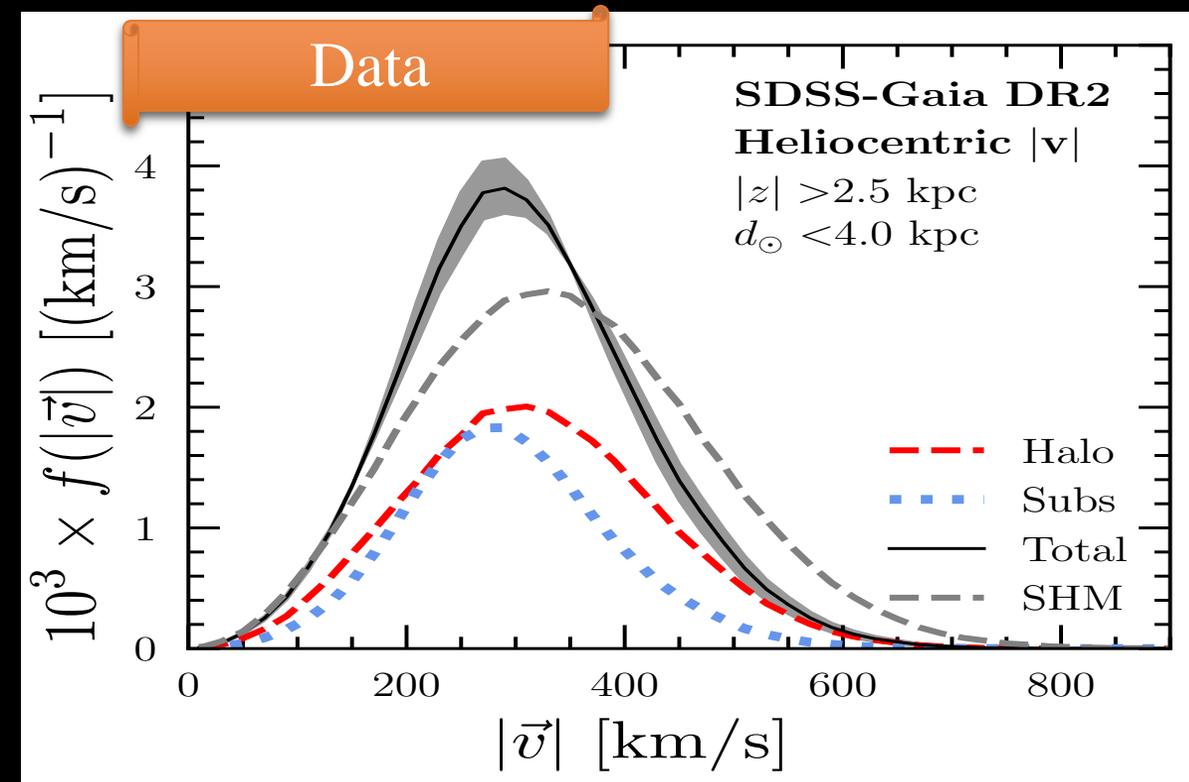


- Different Components:
1. Old Halo
 2. Sausage/ Gaia Enceladus
 3. Nyx



First Empirical Dark Matter Distribution

Sausage



$$f_{\text{total}}(v) = c_{\text{halo}} f_{\text{halo}}(v) + c_{\text{subs}} f_{\text{subs}}(v)$$

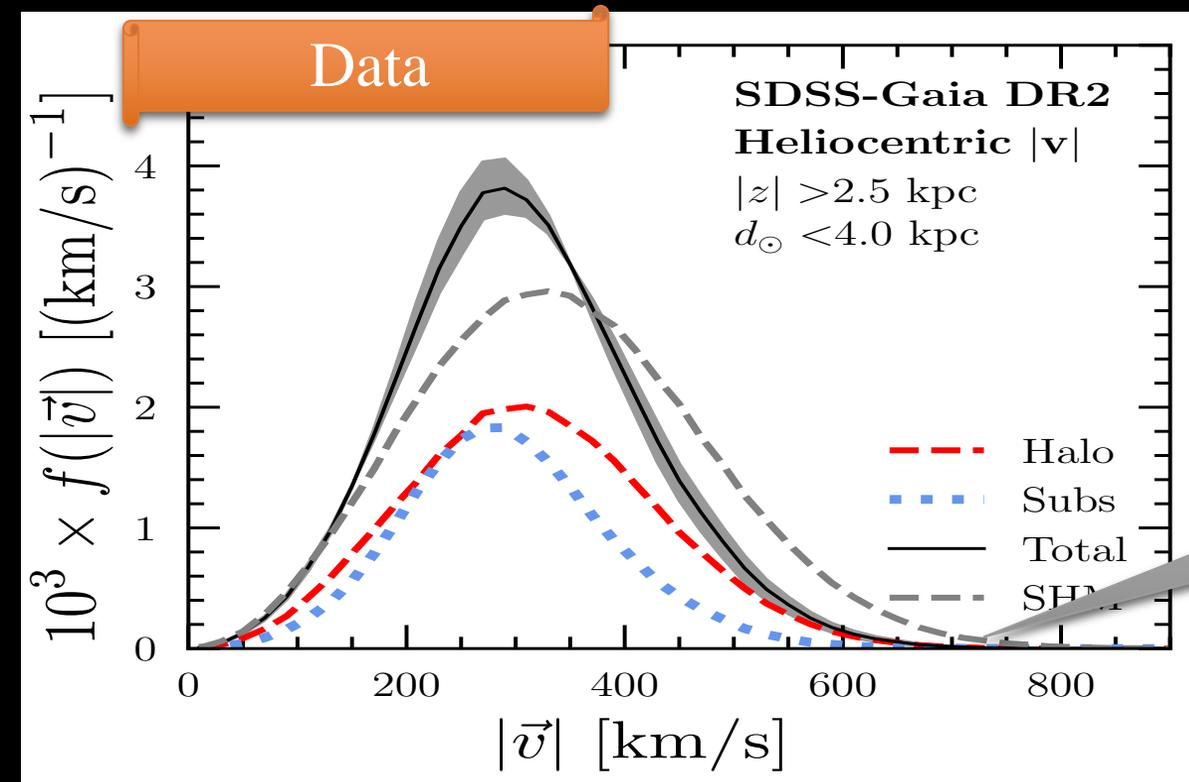
Kirby et al. (2013)

Garrison-Kimmel et al. (2015)

Necib, Lisanti, Belokurov (2018)

Necib, Lisanti, Garrison-Kimmel et al. (2018)

First Empirical Dark Matter Distribution



$$f_{\text{total}}(v) = c_{\text{halo}} f_{\text{halo}}(v) + c_{\text{subs}} f_{\text{subs}}(v)$$

$$c_{\text{subs}} = 0.42^{+0.26}_{-0.22}$$

Final distribution different from the assumed Maxwell Boltzmann distribution

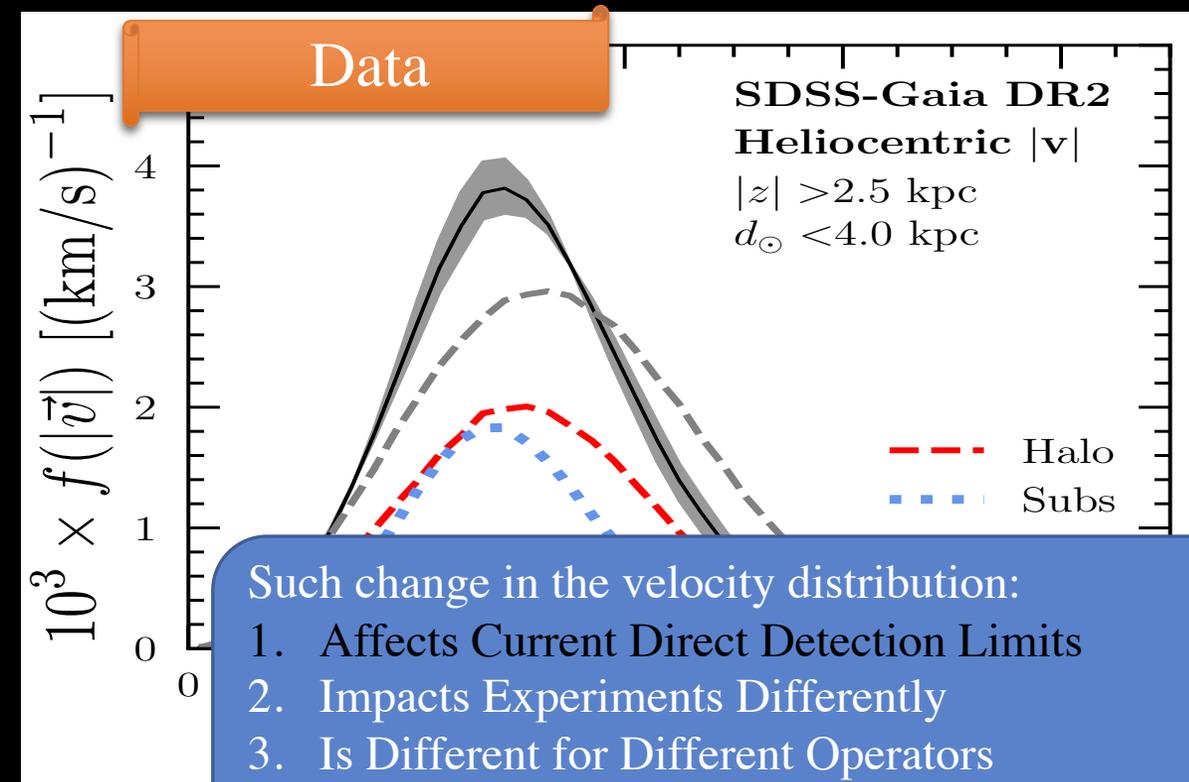
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Necib, Lisanti, Garrison-Kimmel et al. (2018)

First Empirical Dark Matter Distribution



Such change in the velocity distribution:

1. Affects Current Direct Detection Limits
2. Impacts Experiments Differently
3. Is Different for Different Operators
4. Affects Yearly Modulations

$$f_{\text{total}}(v) = c_{\text{halo}} f_{\text{halo}}(v) + c_{\text{subs}} f_{\text{subs}}(v)$$

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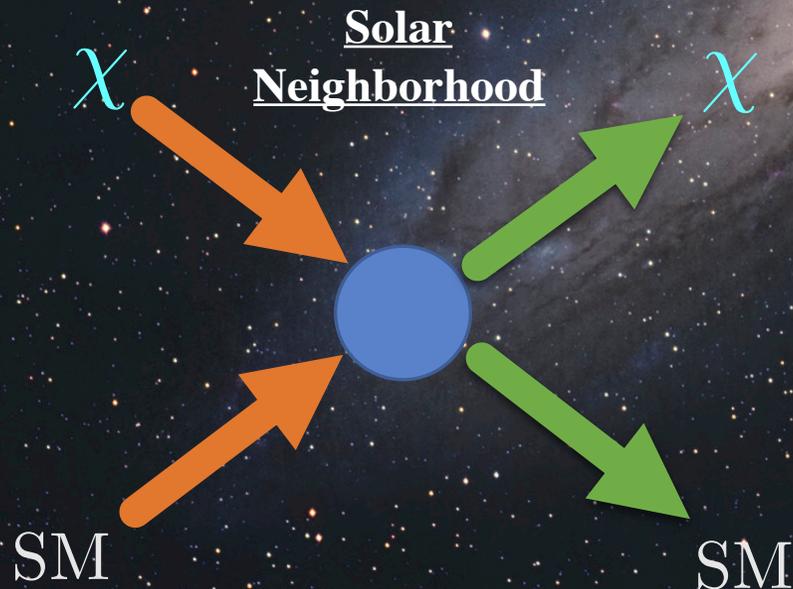
Necib, Lisanti, Belokurov (2018)

Necib, Lisanti, Garrison-Kimmel et al. (2018)

Direct Detection

Direct detection depends on astrophysical parameters:

- Dark matter density
- **Dark Matter velocity**

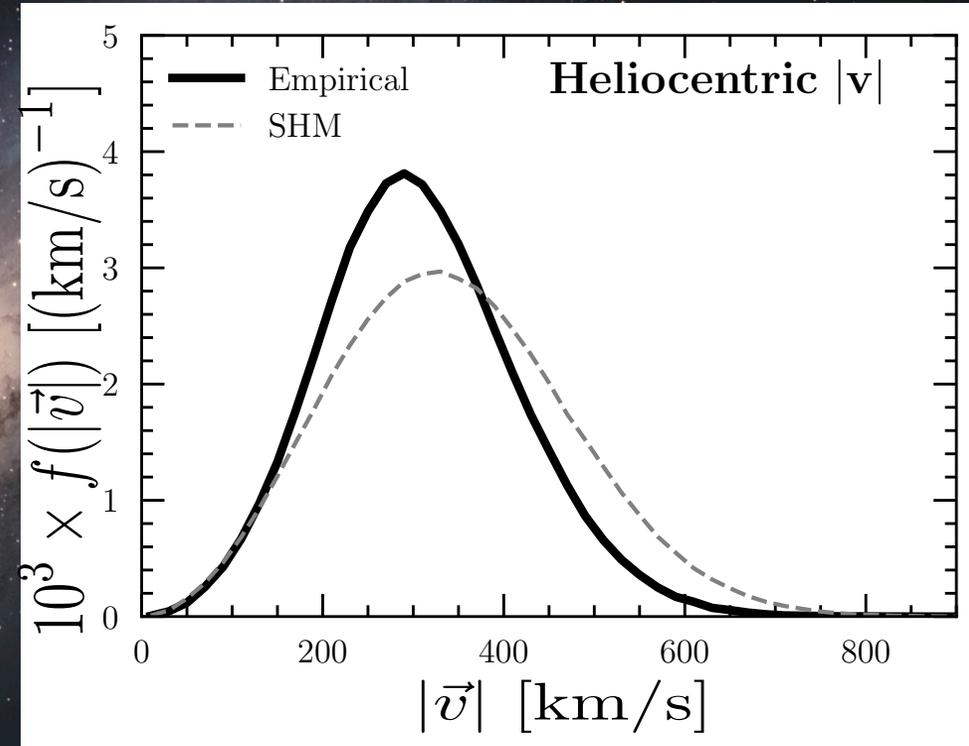
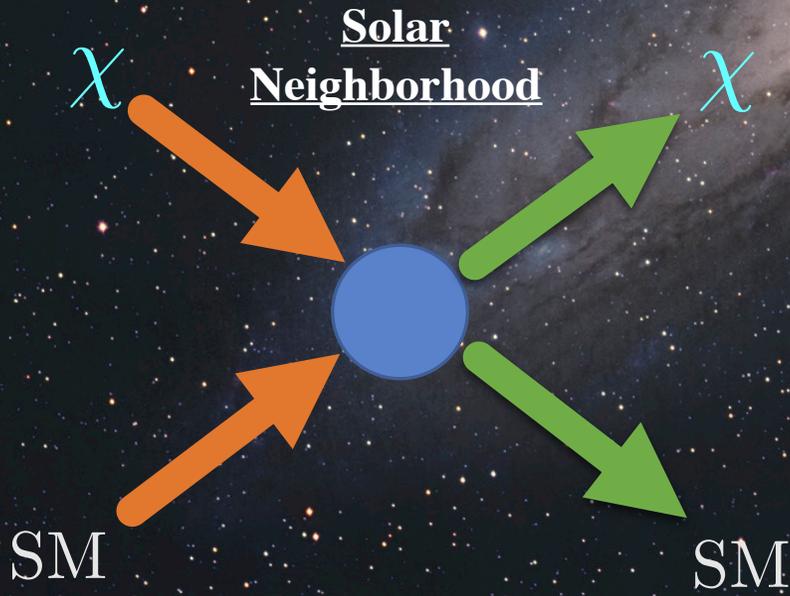


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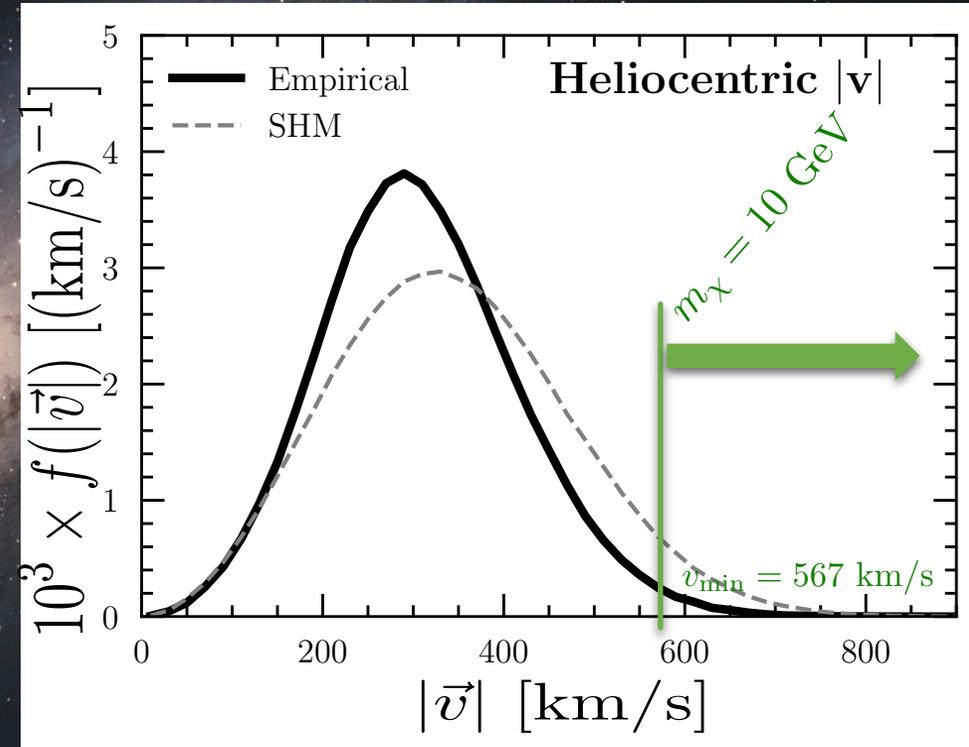
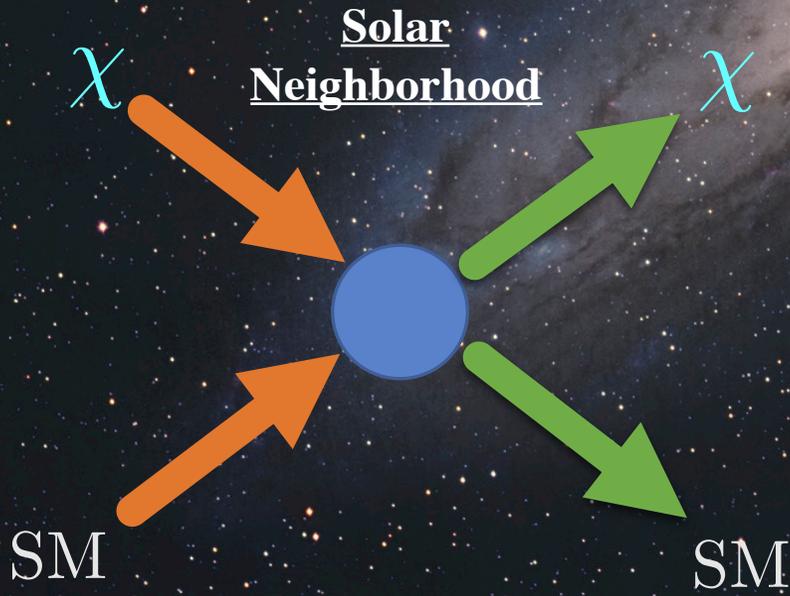
$$\text{Rate} \propto \rho_{DM} \int_{v_{\min}}^{\infty} \frac{f(v)}{v} dv$$

$$v_{\min} = v_{\min}(E_{\text{thresh}}, m_{\chi})$$

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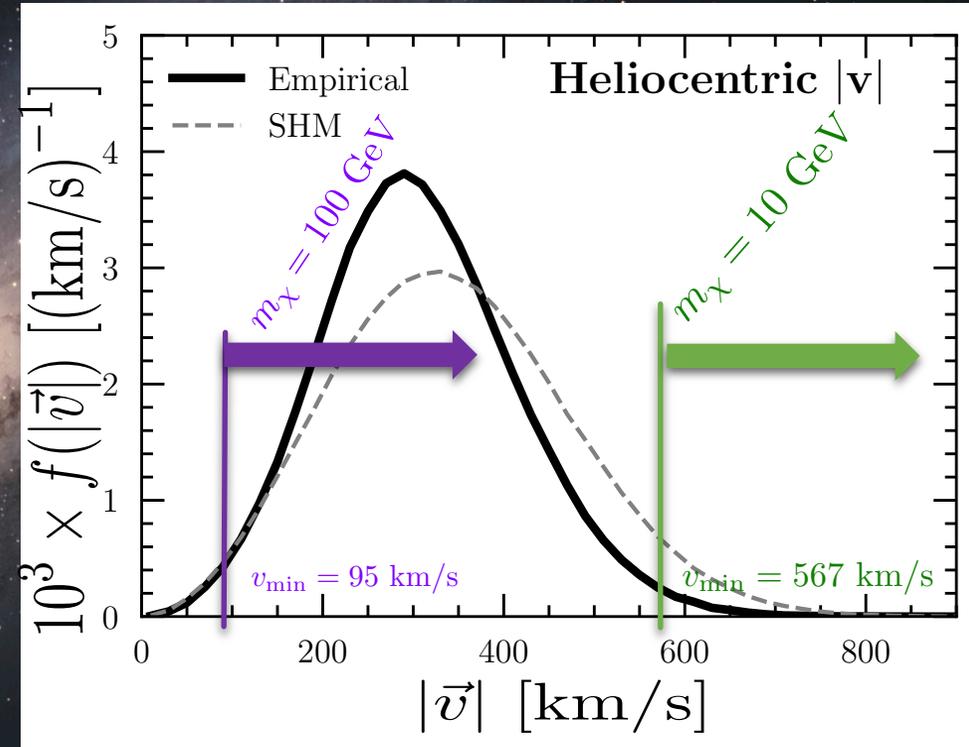
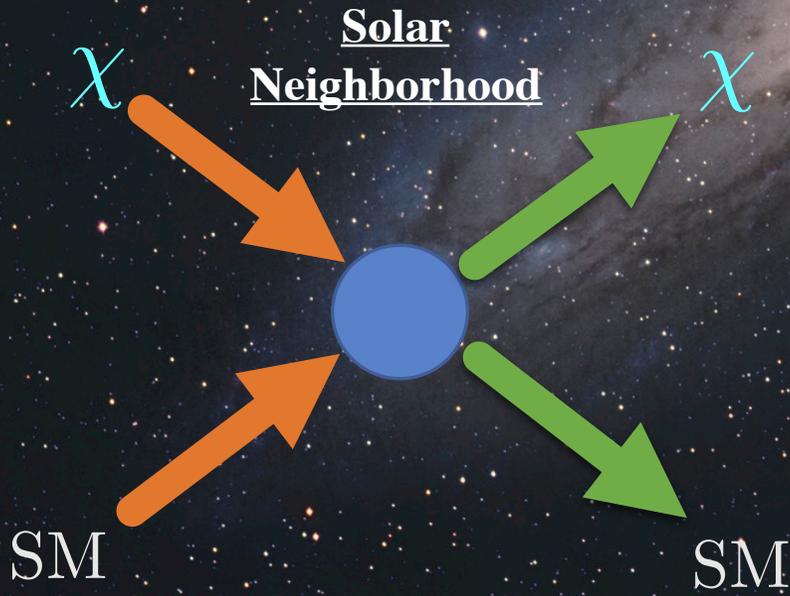
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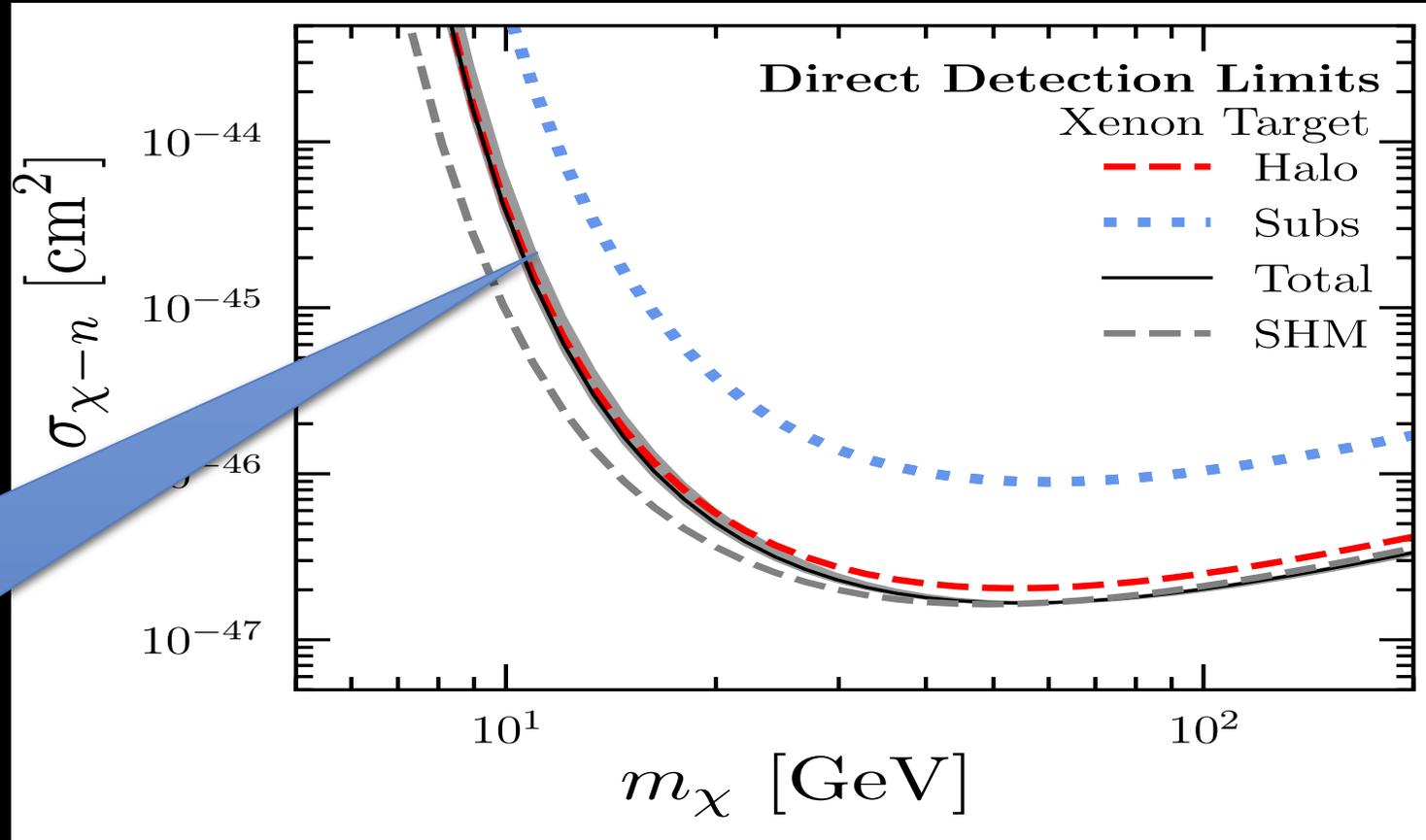
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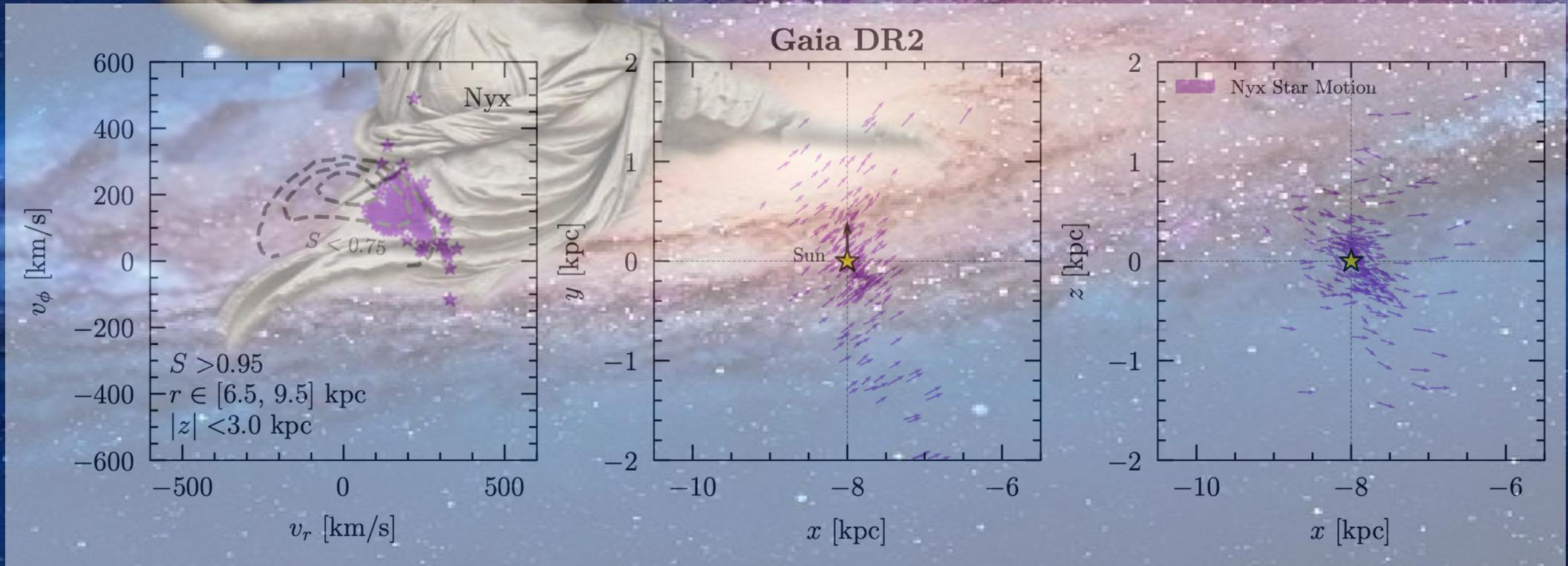
Implications for Direct Detection



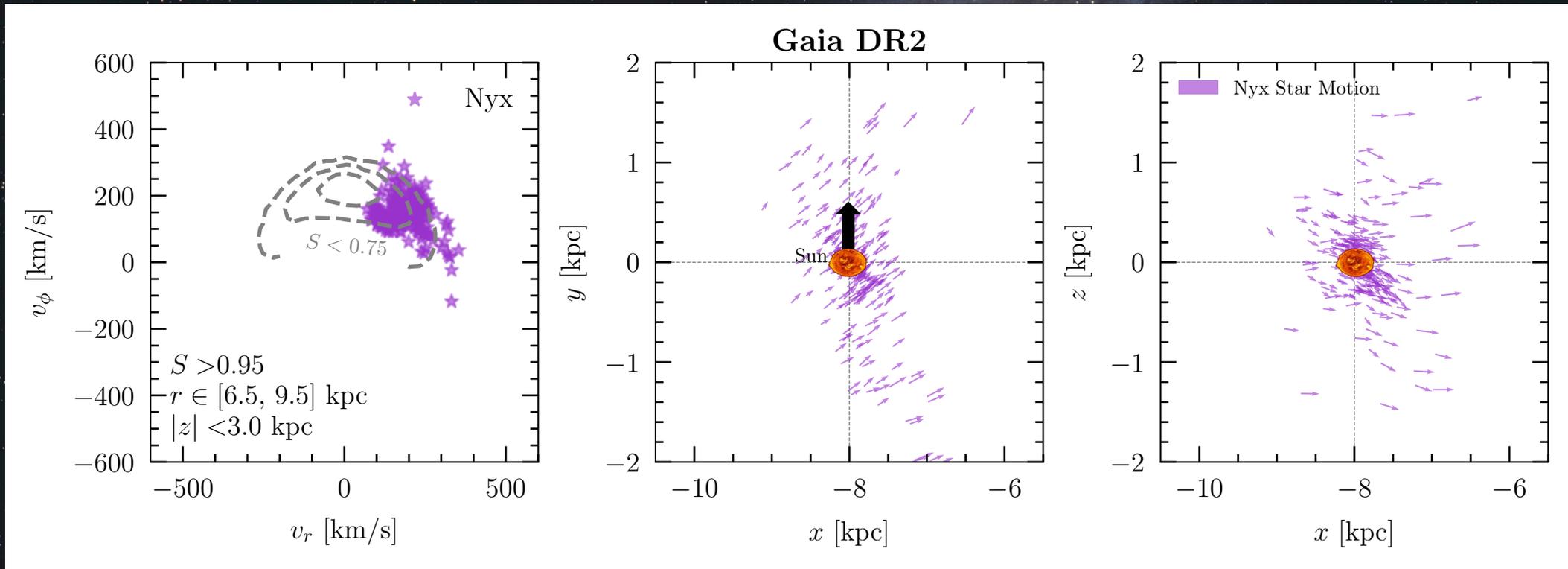
Largest changes are at low dark matter masses

Other theory models might have stronger effects. See for example Buch et al. 2019

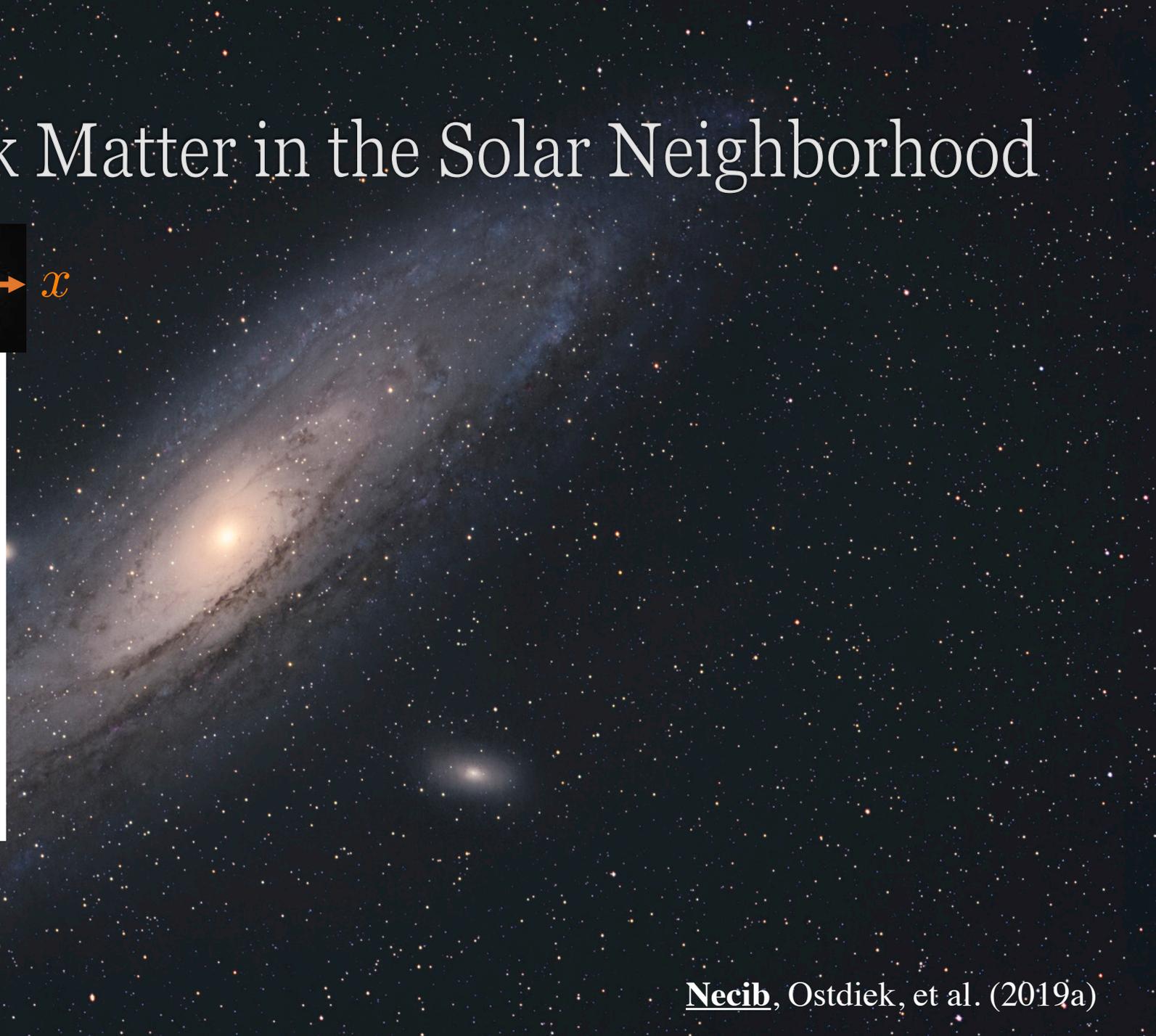
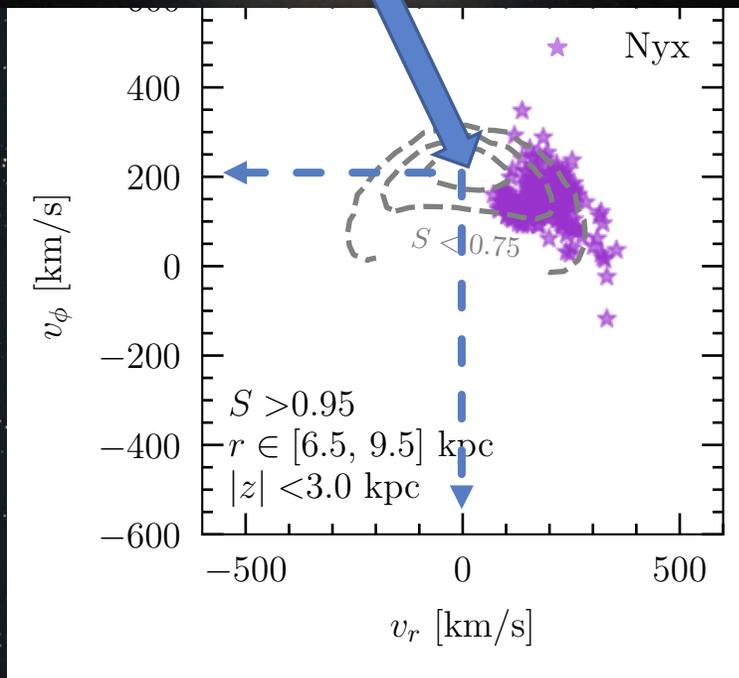
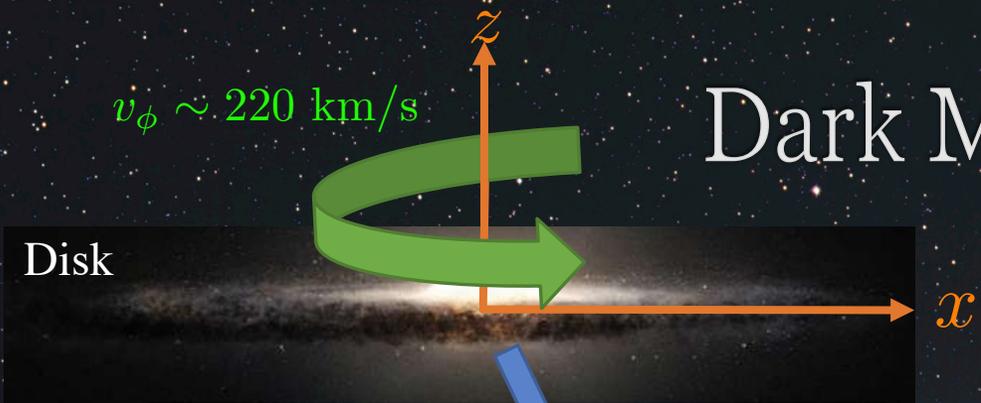
Nyx: Greek Goddess of the Night



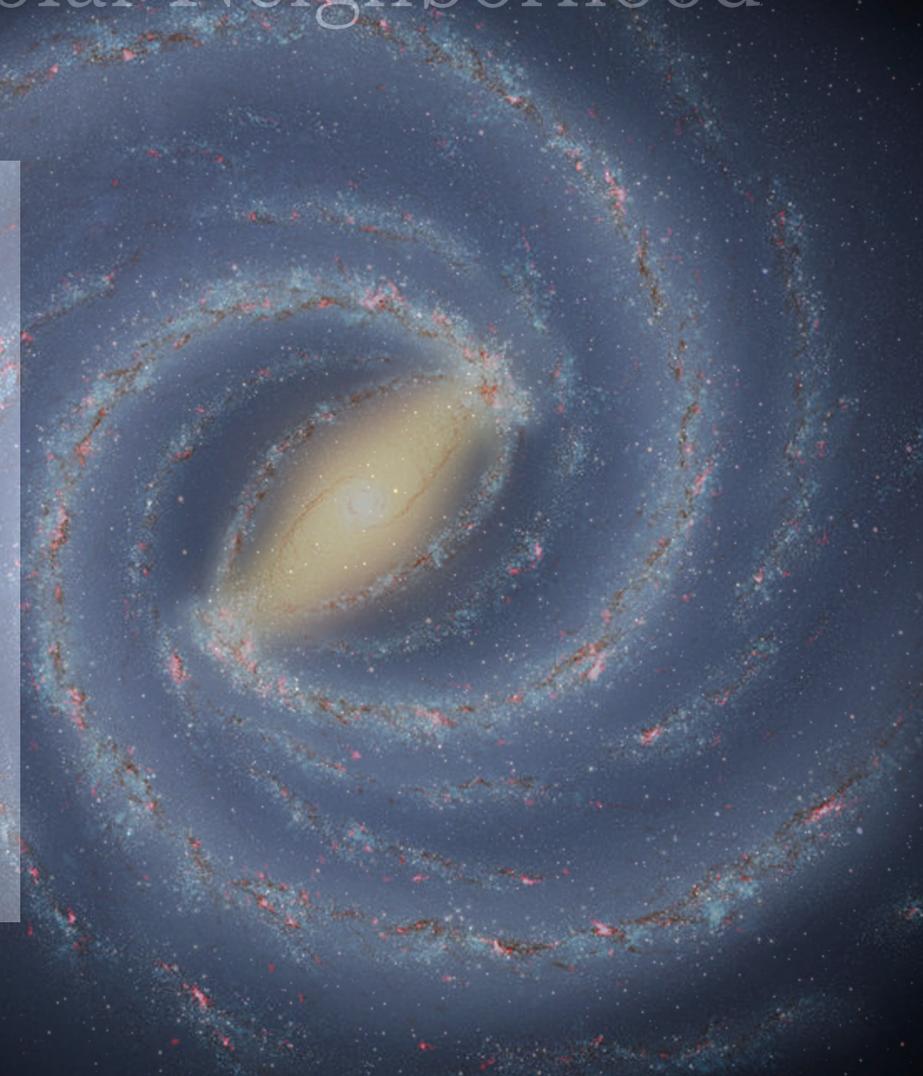
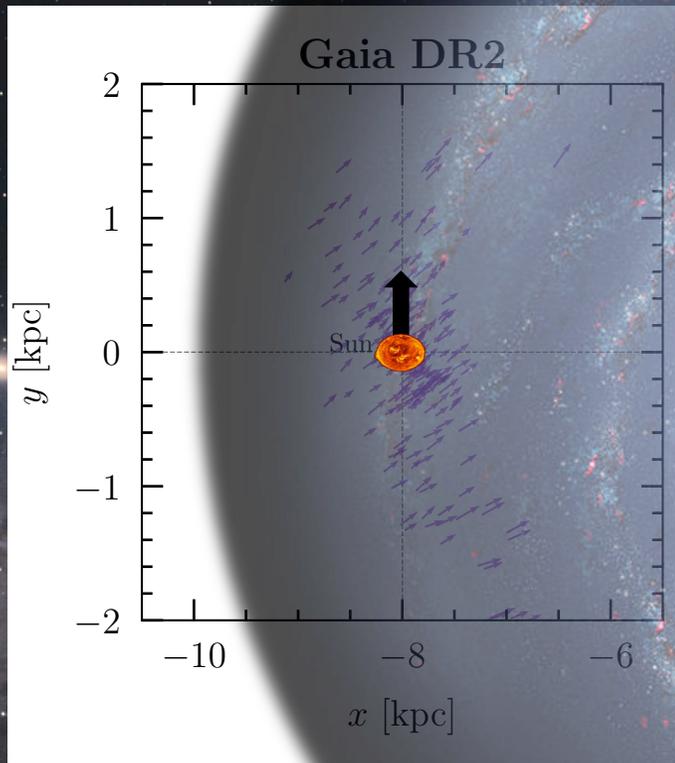
Dark Matter in the Solar Neighborhood



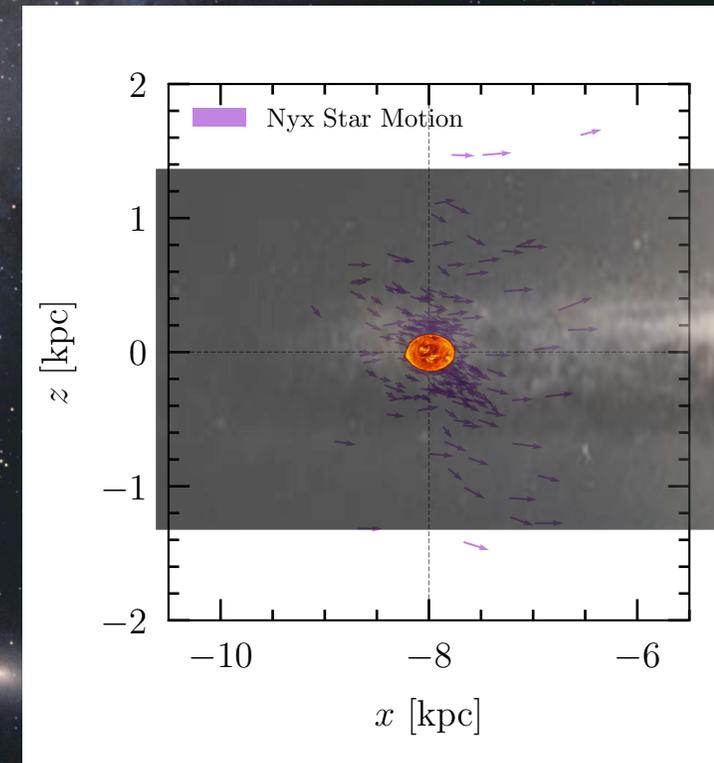
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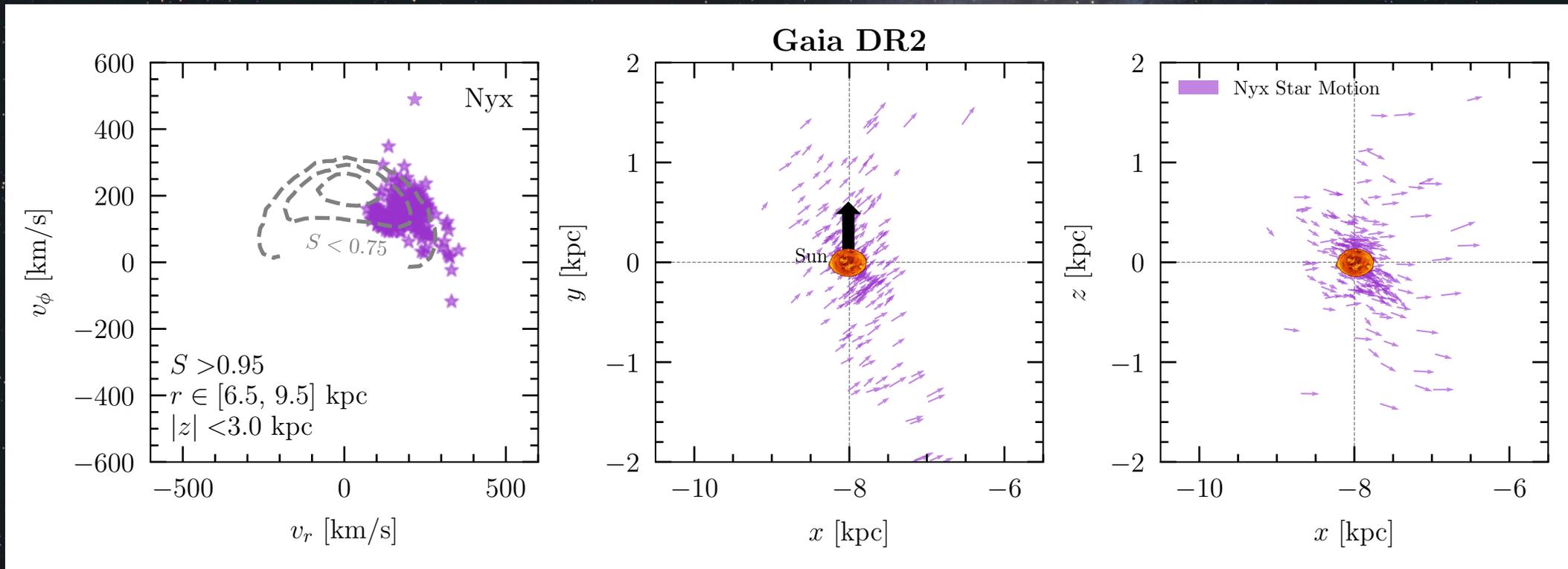
Dark Matter in the Solar Neighborhood



Dark Matter in the Solar Neighborhood



Dark Matter in the Solar Neighborhood



Simulation Analog

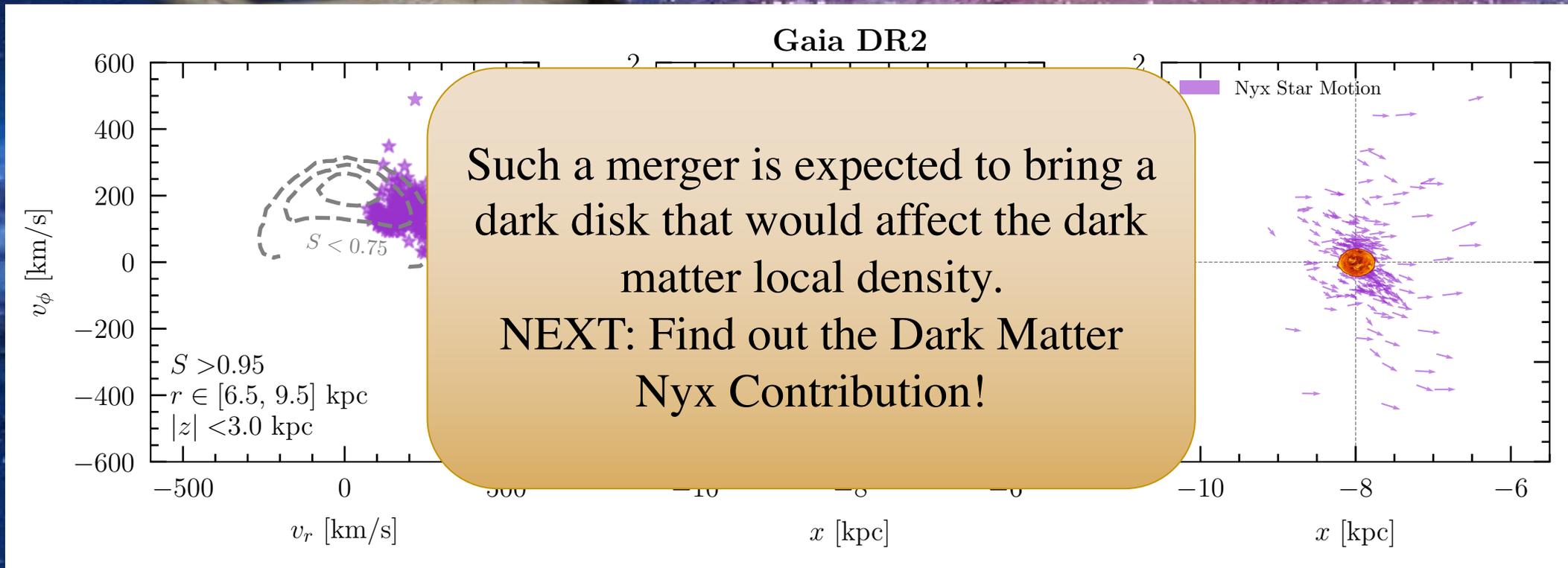
$z=0.12$



Hopkins et al. (2014)
Wetzel et al. (2016)
Hopkins et al. (2017)

Video by Shea Garisson-Kimmel,
<http://www.tapir.caltech.edu/~sheagk/firemovies.html>

Nyx: Greek Goddess of the Night



Dark Disk

As satellites are torn apart by tidal forces, they deposit both their stars *and their dark matter* into a thick disc (Lake 1989). The latter point is the key new idea presented in this work: a dark matter disc must form in a Λ CDM cosmology

Read et al. (2008)

Lake (1989)

Read et al. (2008)

Bruch et al. (2008)

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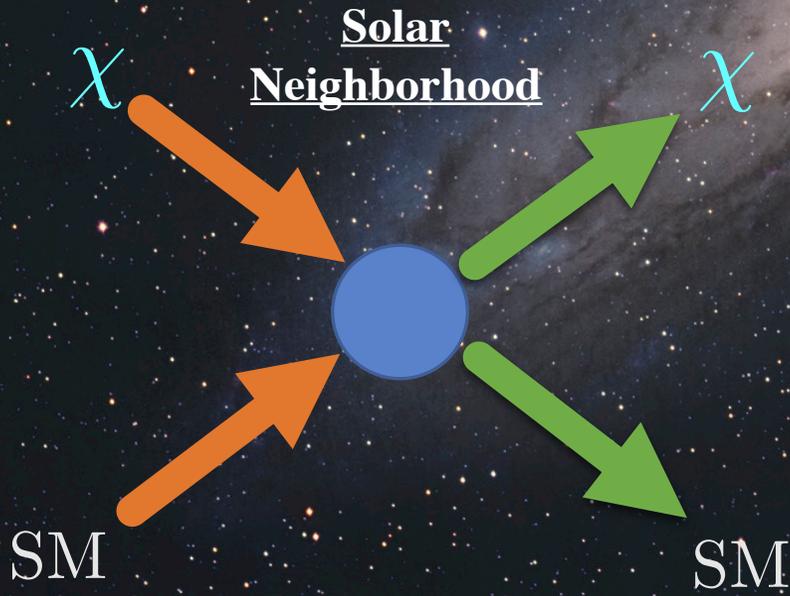
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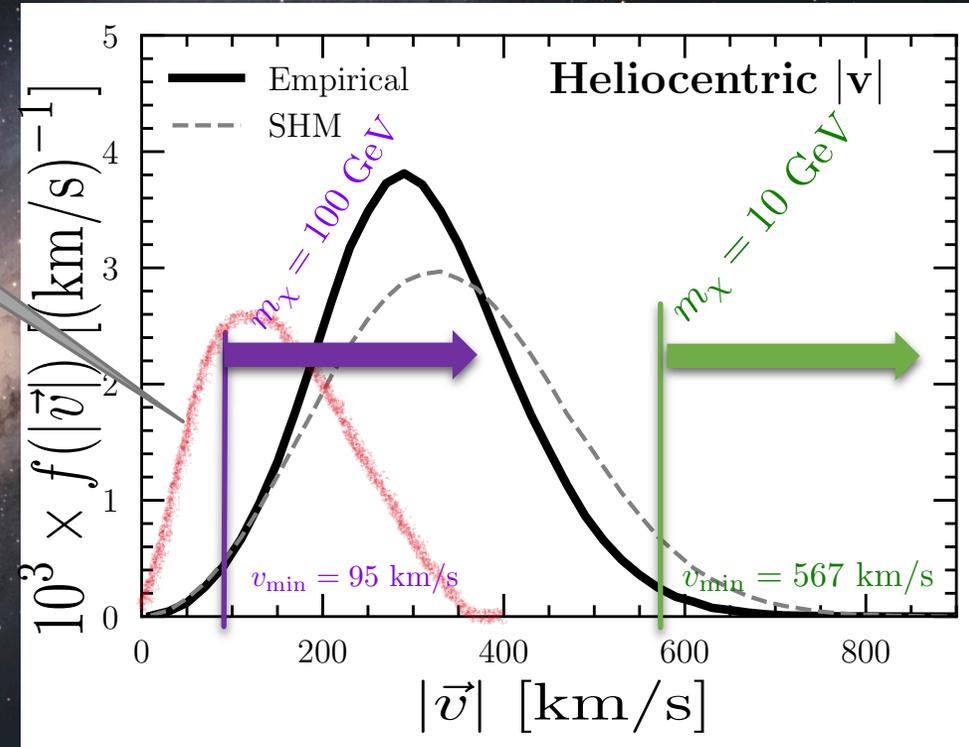
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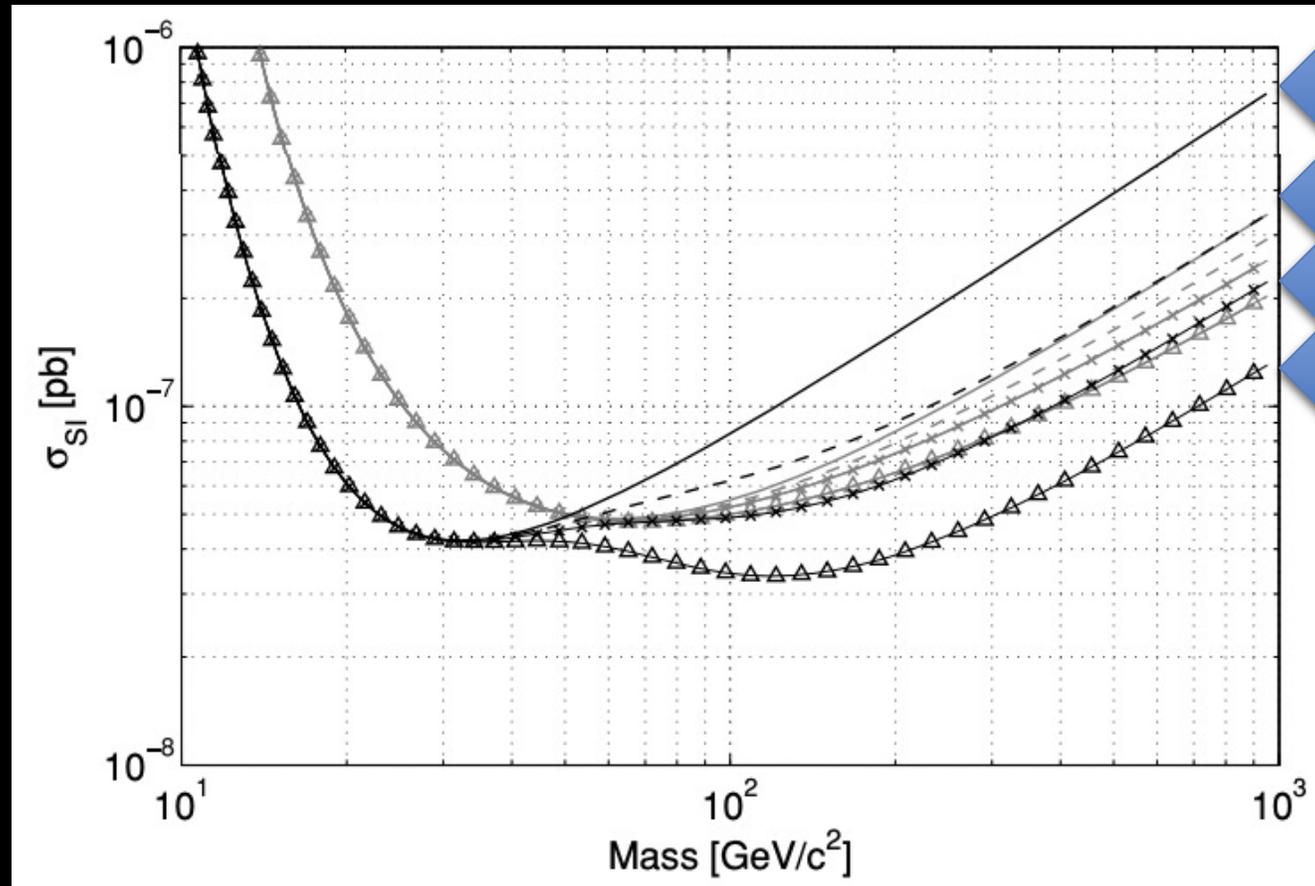
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Effect of Dark Disk on Direct Detection



No Dark Disk

$\rho_d/\rho_h = 0.5$

$\rho_d/\rho_h = 1$

$\rho_d/\rho_h = 2$

ρ_d : Dark Disk Density

ρ_h : Dark Halo Density

Simulation Analog

NEXT:

- Observational study of Nyx
- Correlating dark matter and stars in streams
- Estimating relative contribution of dark matter in Nyx
- Estimating relative contribution of Dark Subhalos to the local Dark Matter



Simulation Analog

NEXT:

- **Observational study of Nyx**

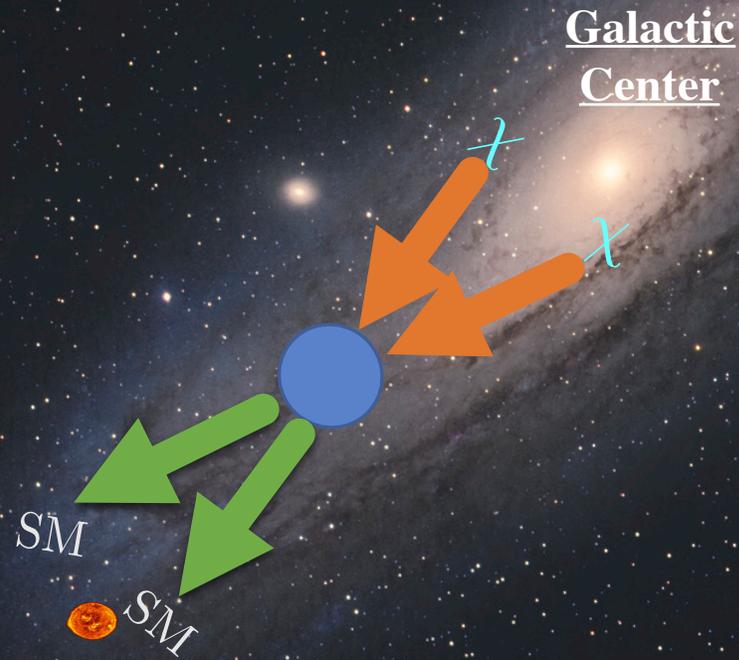
- Correlating dark matter and stars in
Using Keck and Magellan, with Alex Ji, Mimi Truong, Allison Strom, and Mia de los Reyes, we observed a selection of Nyx stars.

- Estimating relative contribution of dark matter to the local Dark Matter
So far, the majority is consistent with the thick disk. However, we have an interesting low metallicity tail to investigate.



Credit: Anna Frebel

Dark Matter in the Galactic Center

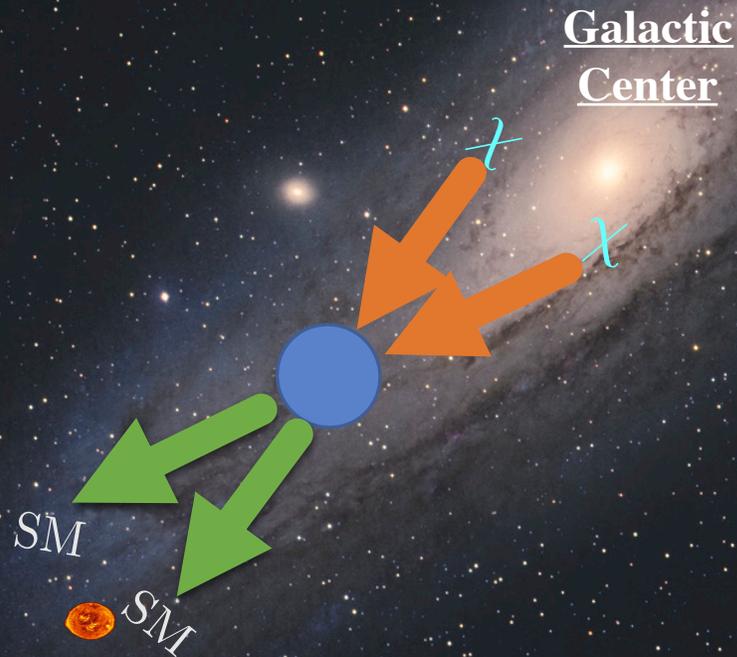


Dark Matter in the Galactic Center

J factor:

Integrated density squared of Dark Matter along the line of sight.

$$J = \int ds \int d\Omega \rho^2$$



Pros

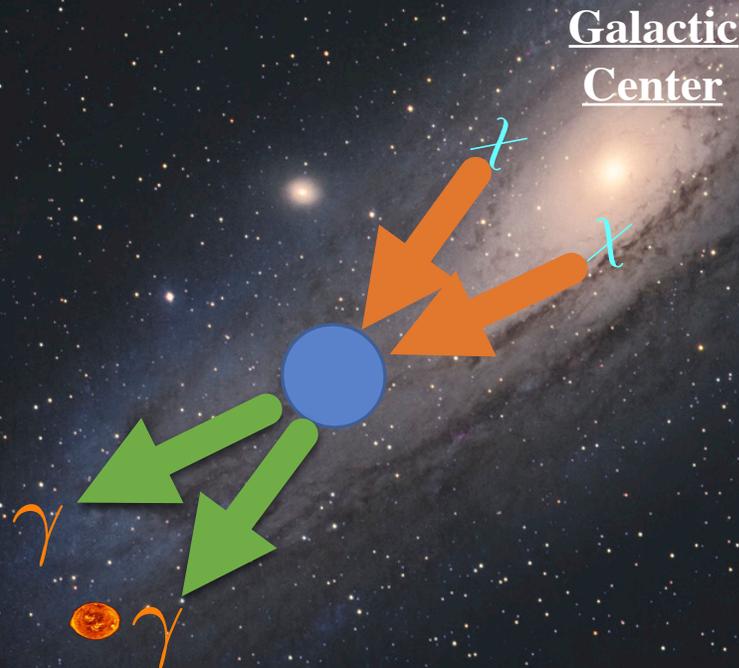
- Largest Density of Dark Matter

Cons

- Largest Astrophysical Backgrounds

Dark Matter in the Galactic Center

$$\mathcal{L} = \mathcal{L}_{\text{SM}} + \bar{\chi}(i\not{D} + M)\chi$$

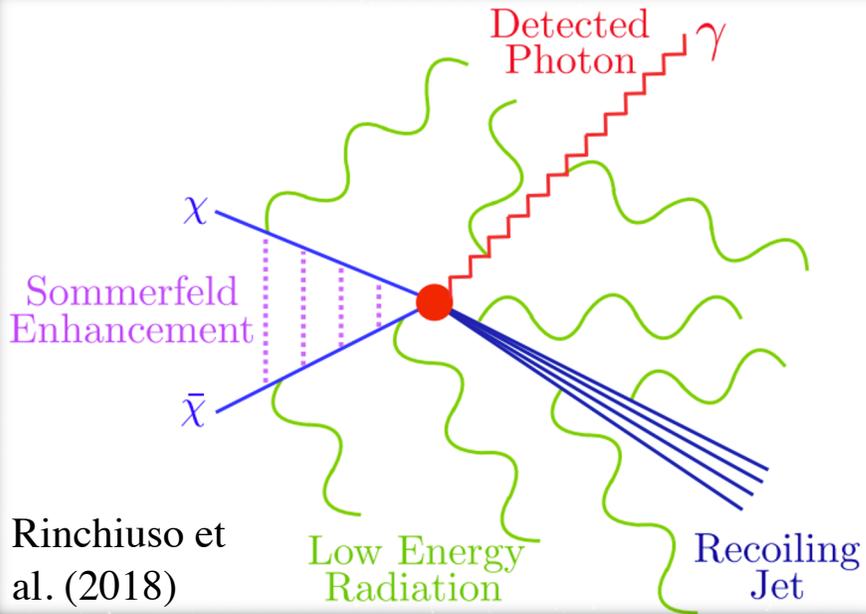


- χ A simple extension of the Standard Model.
Single New Electroweak Triplet Fermion.
Expected in TeV mass scale,
hard to probe in colliders.

Chardonnet, Salati, Fayet (1993)

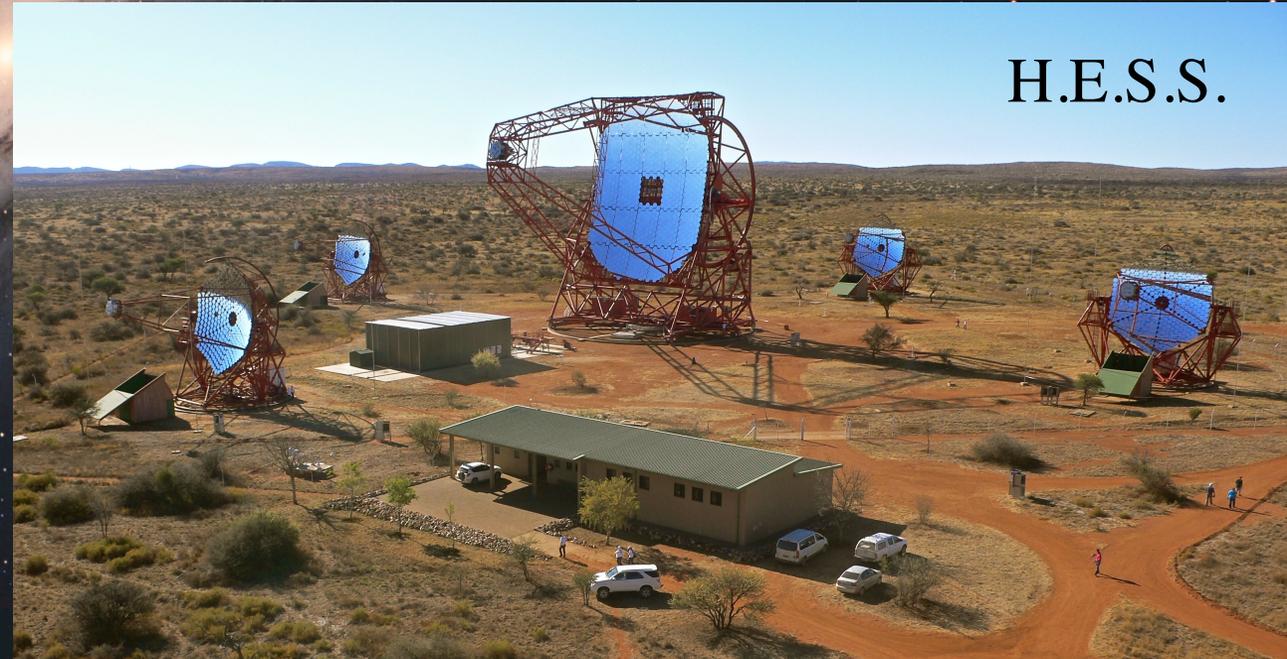
Hisano et al. (2003, 2004, 2005, 2006, 2007)

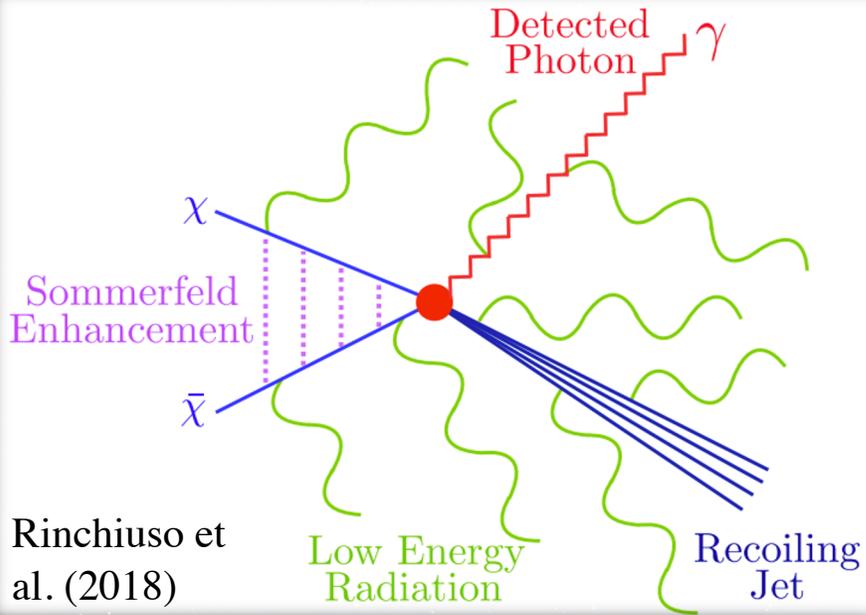
Cirelli et al. (2006, 2007)



Rinchiuso et al. (2018)

Dark Matter in the Galactic Center





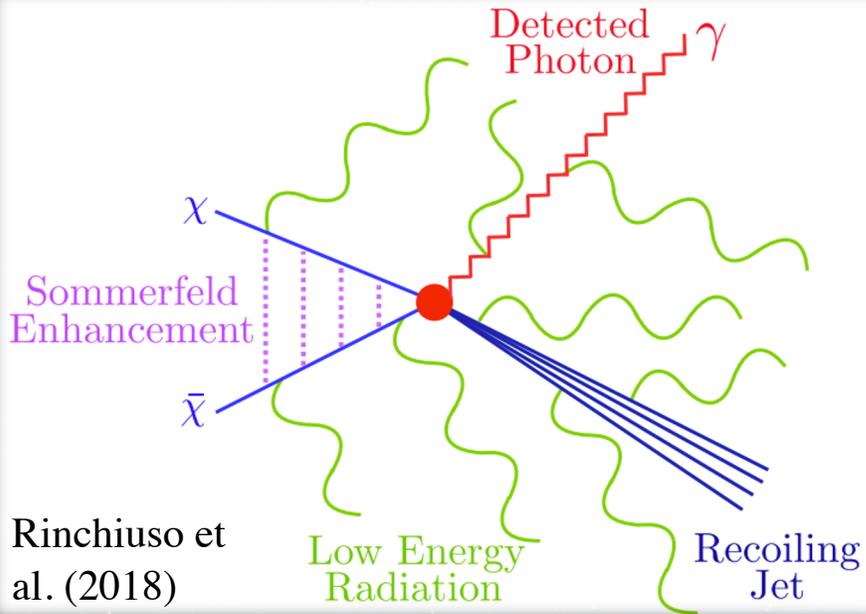
Rinchiuso et al. (2018)

Dark Matter in the Galactic Center

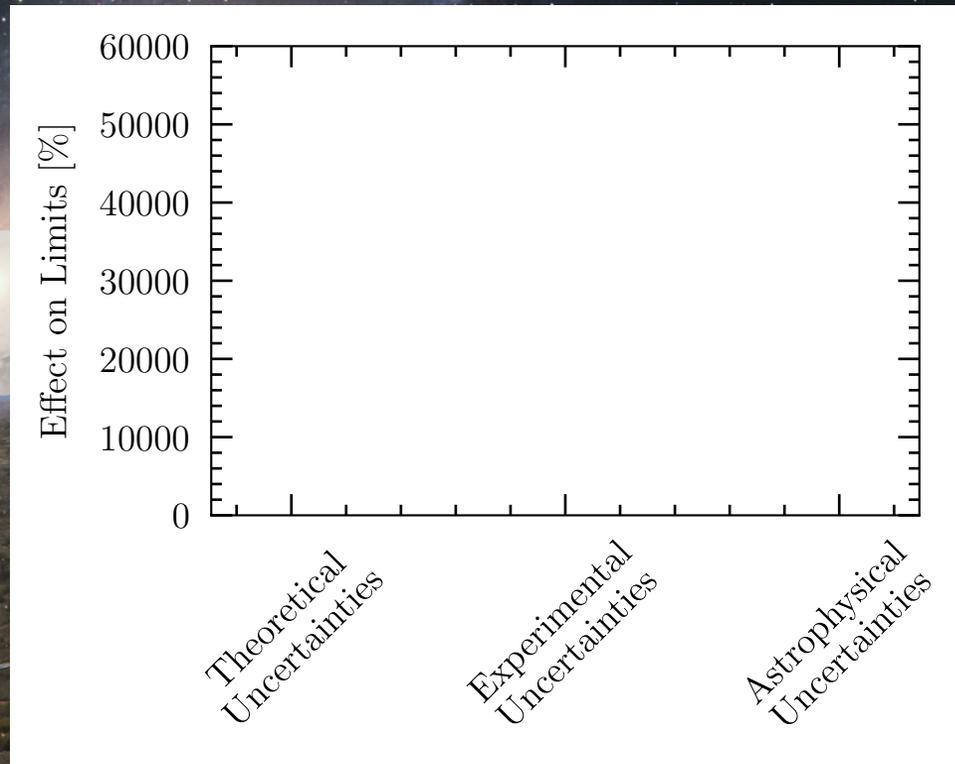


$$N_{\text{signal}} = \text{Flux}_{\gamma} \times \text{Experimental Efficiency} \times \text{Exposure}$$

$$\propto \langle \sigma v \rangle \int_{\text{line of sight}} \int_{\text{angles}} \rho_{\chi}^2 d\Omega ds \times A_{\text{eff}}^{\gamma}$$

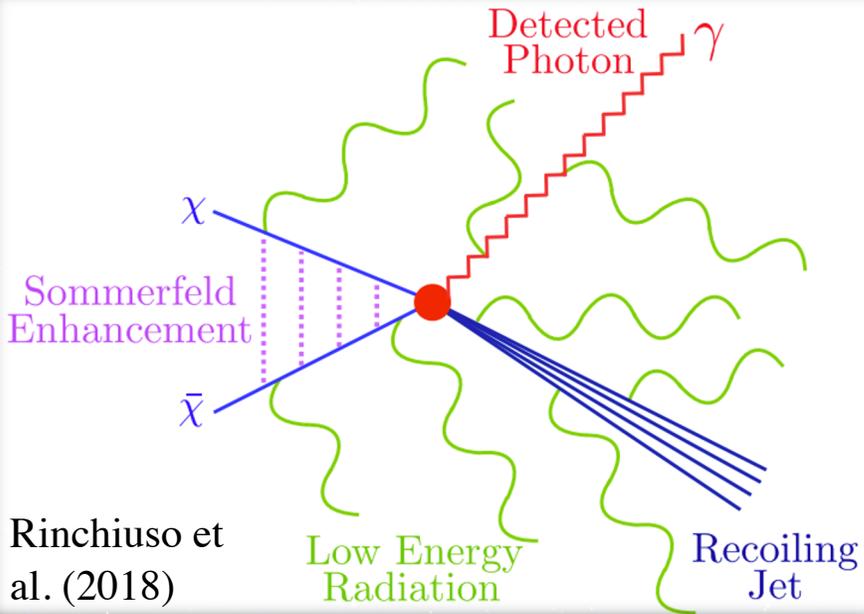


Dark Matter in the Galactic Center

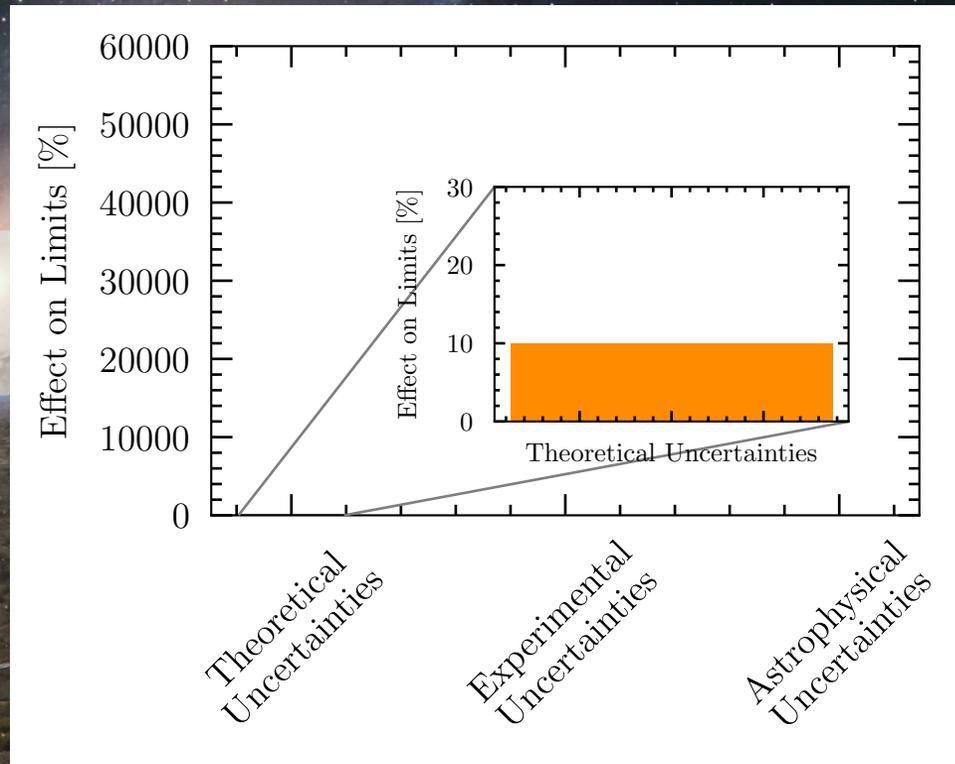


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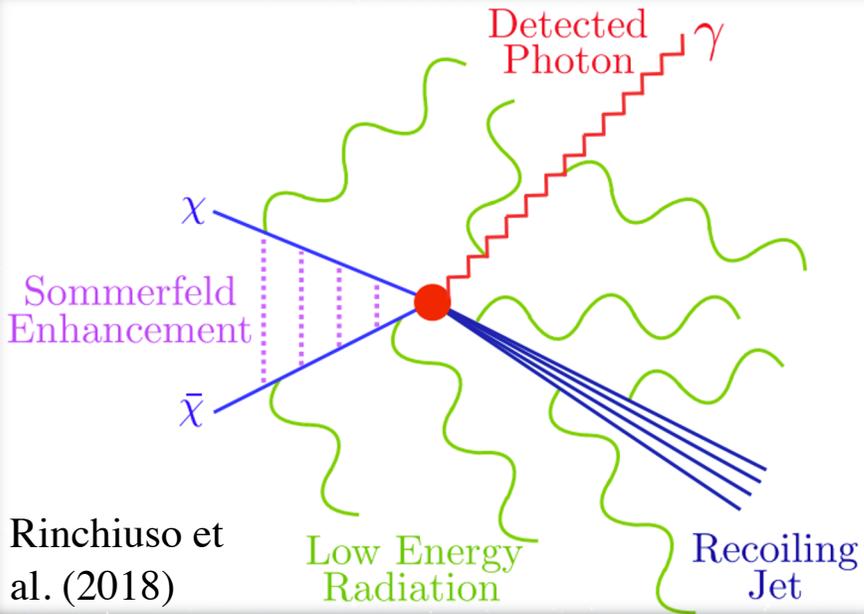
Dark Matter in the Galactic Center



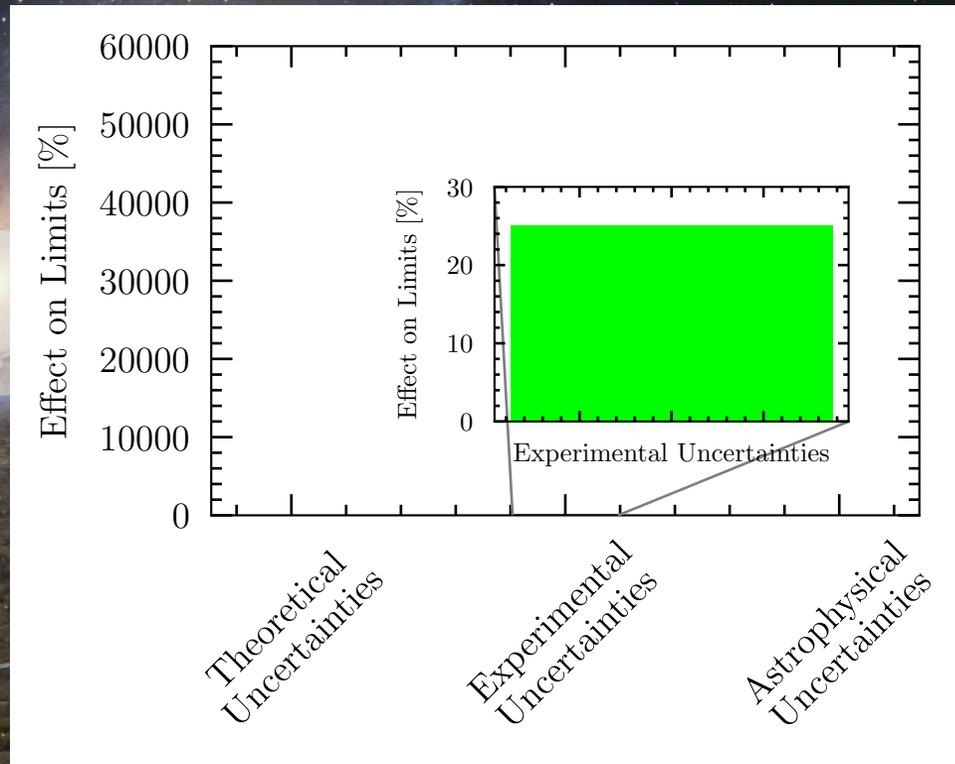
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- Bauer et al. (2015),
- Ovanesyan et al. (2015),
- Ovanesyan et al. (2017),
- Baumgart et al. (2017),
- Baumgart et al. (2018)



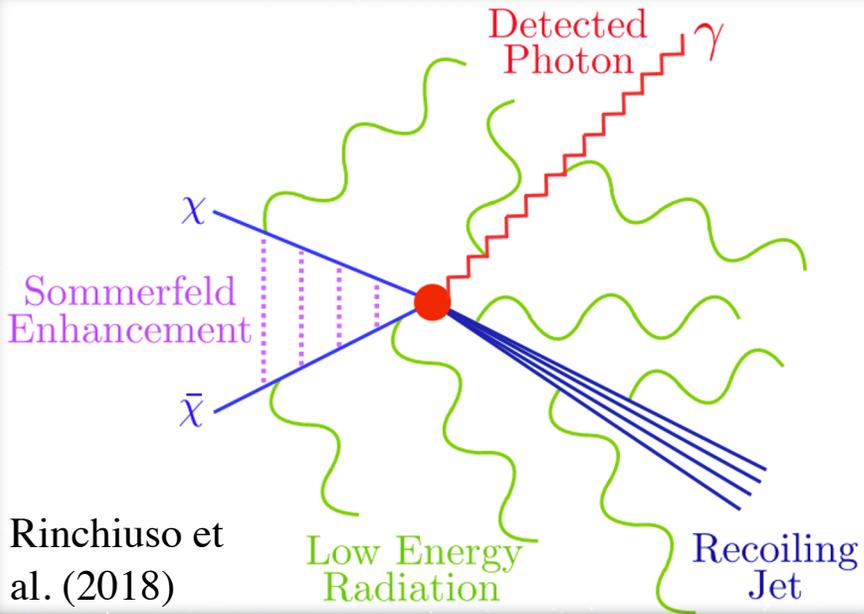
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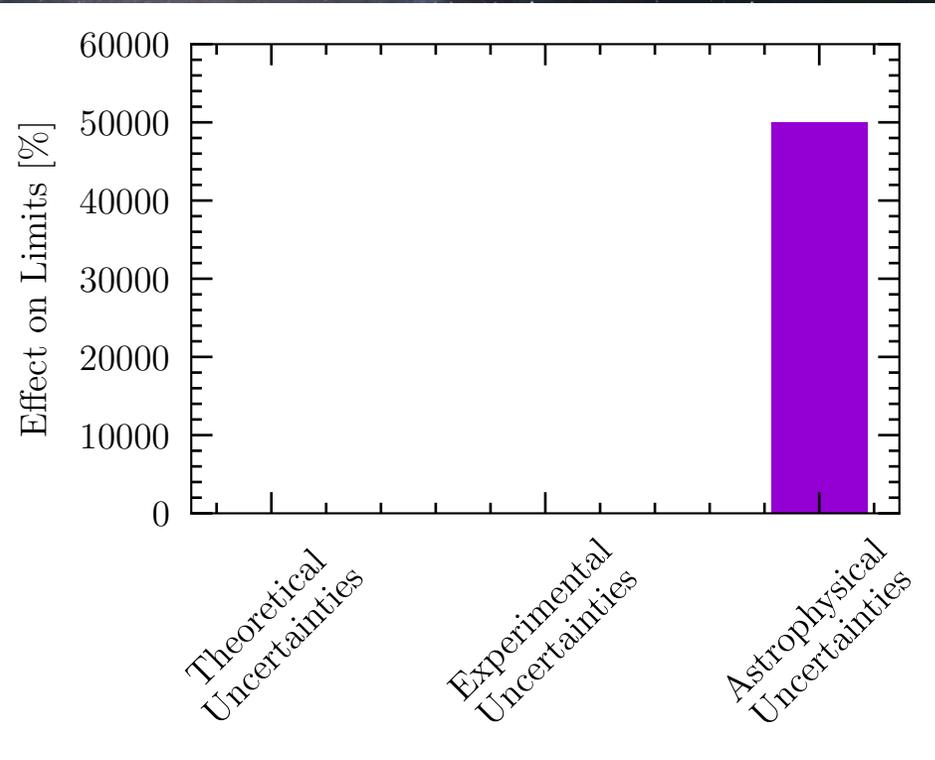
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H.E.S.S. Collaboration (2006)
 Holler et al. (2017)
 H.E.S.S. Collaboration (2018)



Rinchiuso et al. (2018)

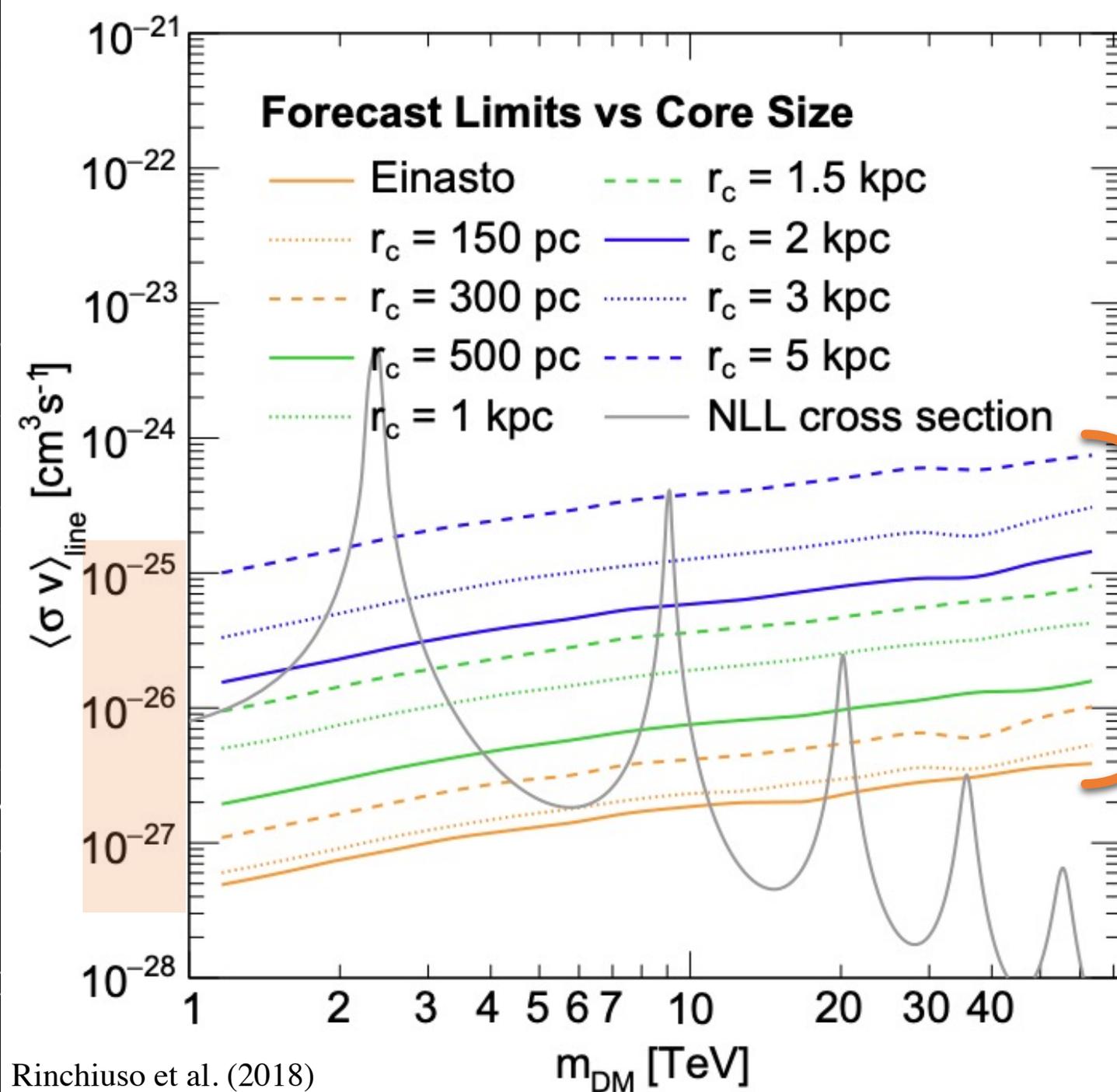
Dark Matter in the Galactic Center



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Dark Matter in the Galactic Center



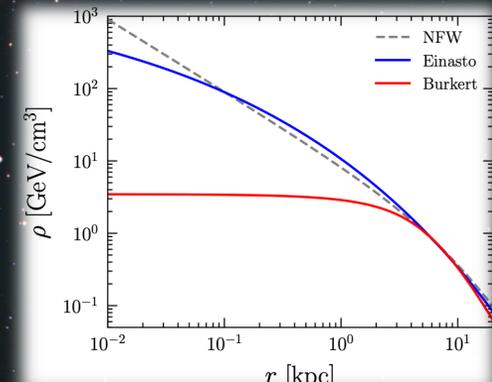
Different Profiles

$$J \propto \int \int_{\text{l.o.s}} \rho^2 ds d\Omega$$

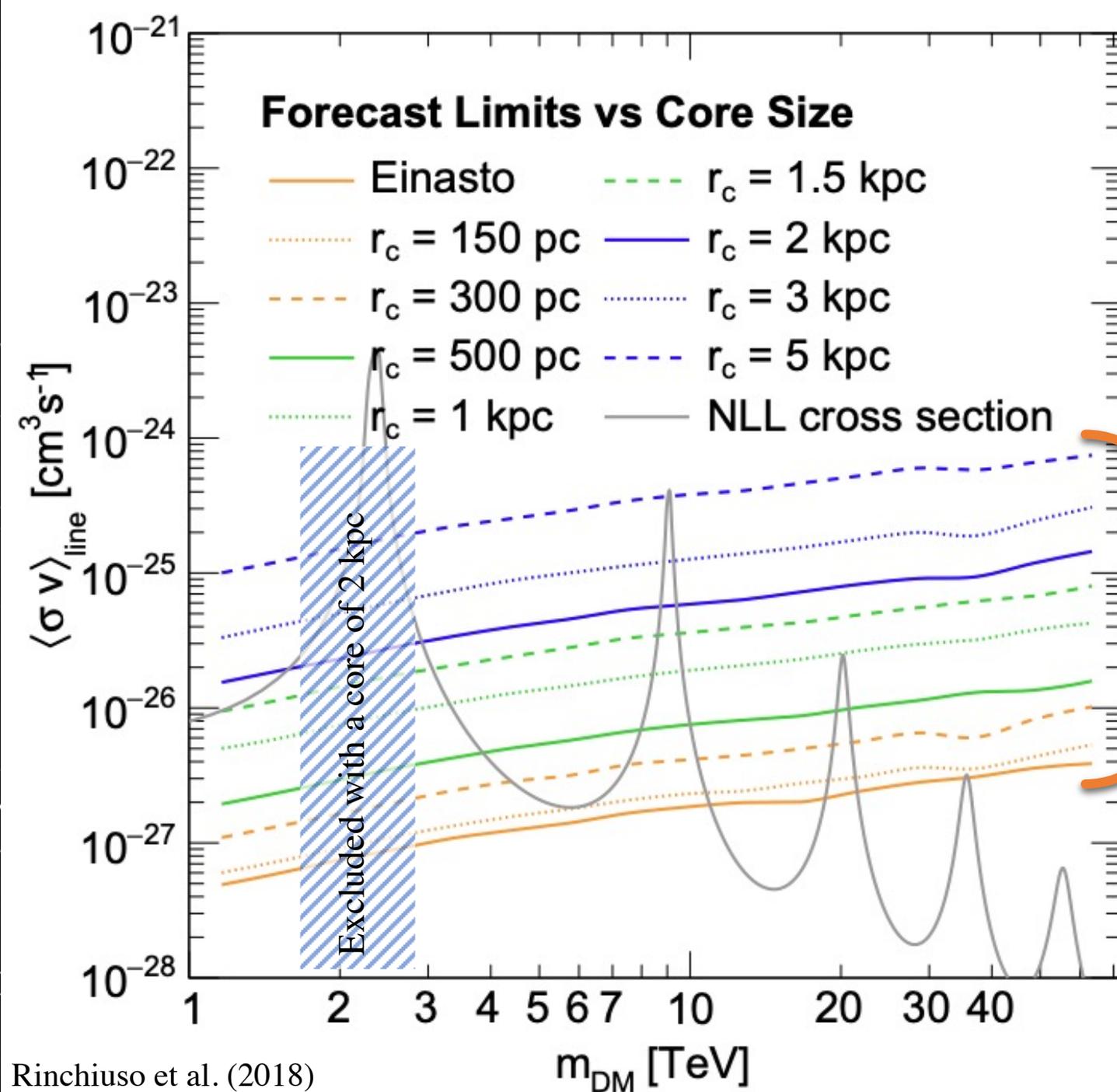
ρ : Dark Matter Density

ds : Line of Sight Integral

$d\Omega$: Solid Angle Integral



Dark Matter in the Galactic Center



Different Profiles

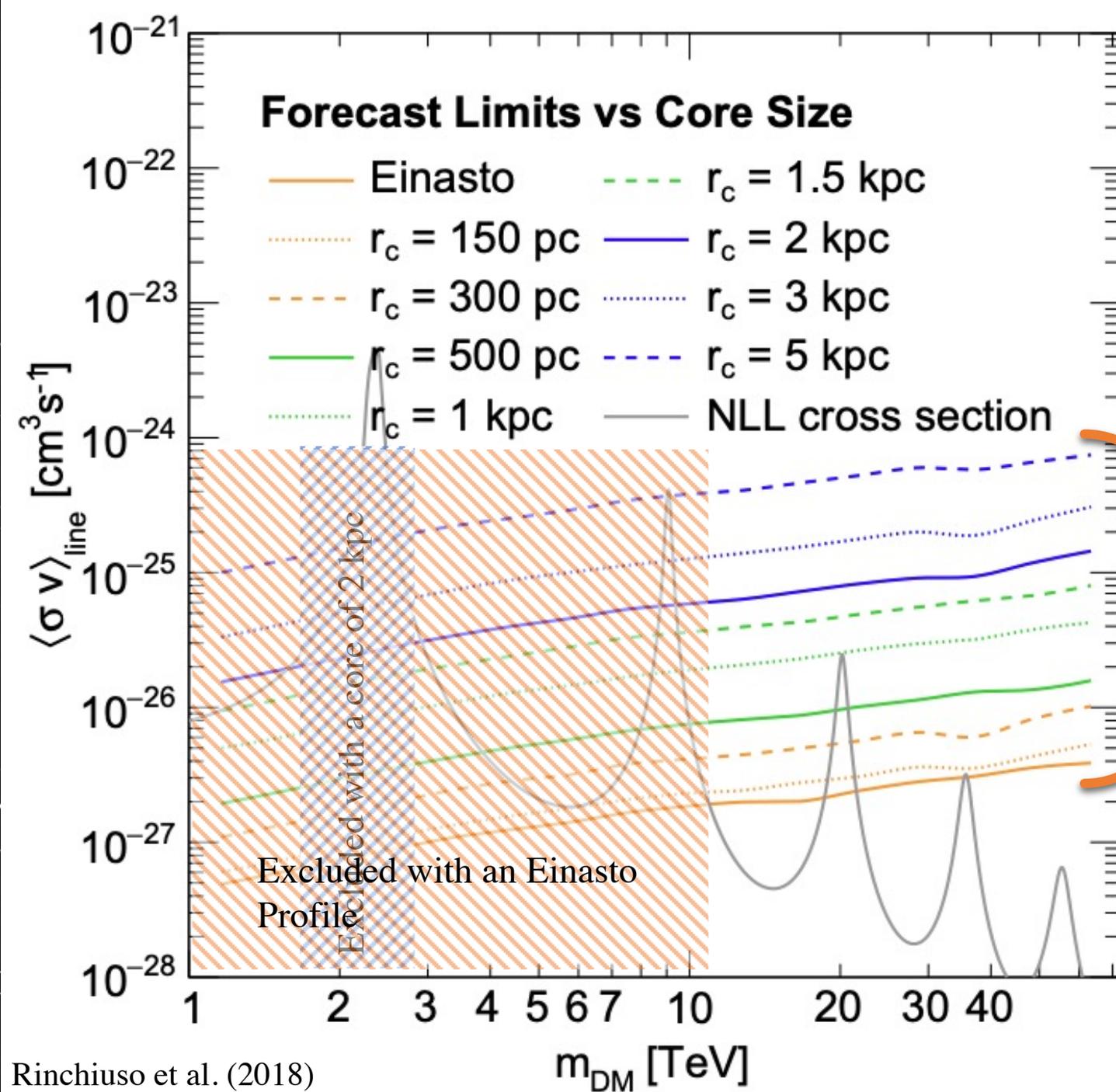
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Dark Matter in the Galactic Center



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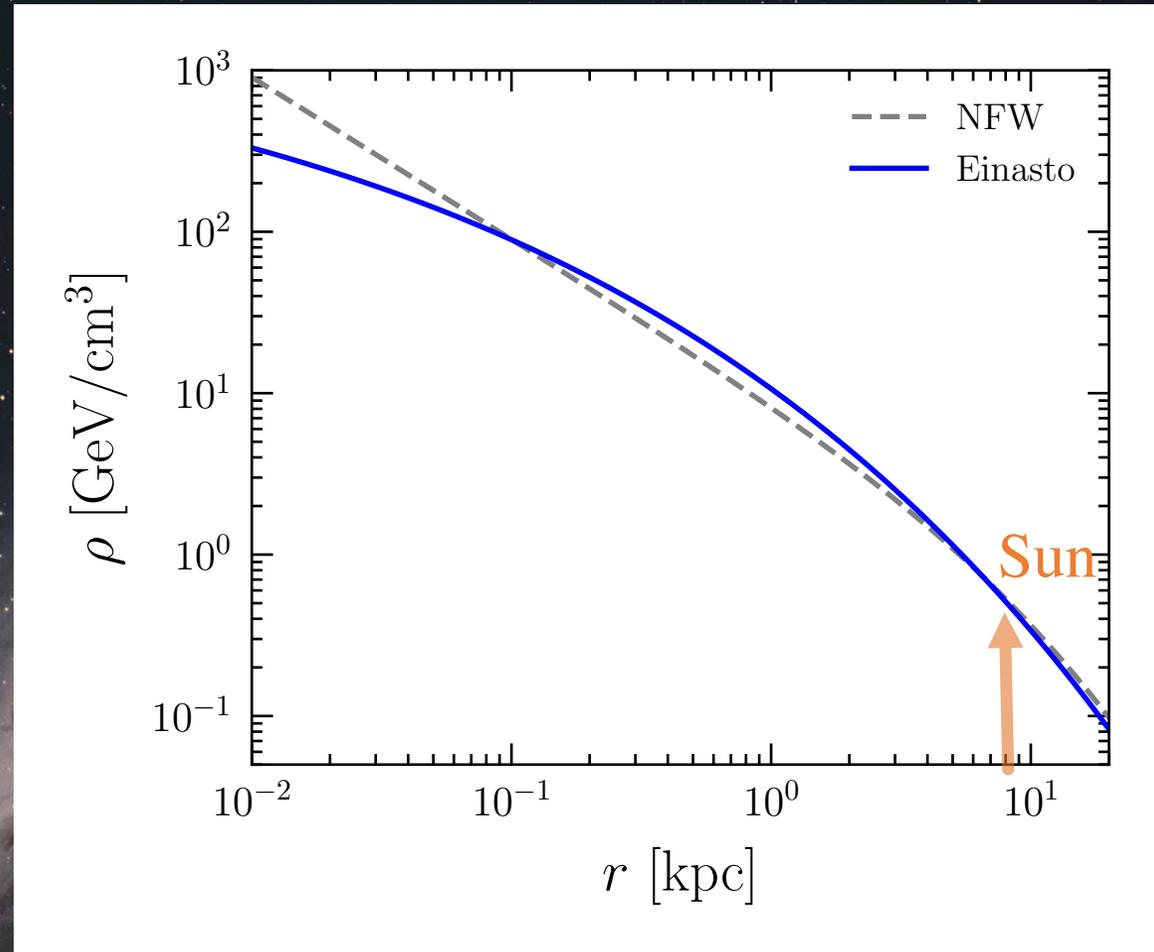
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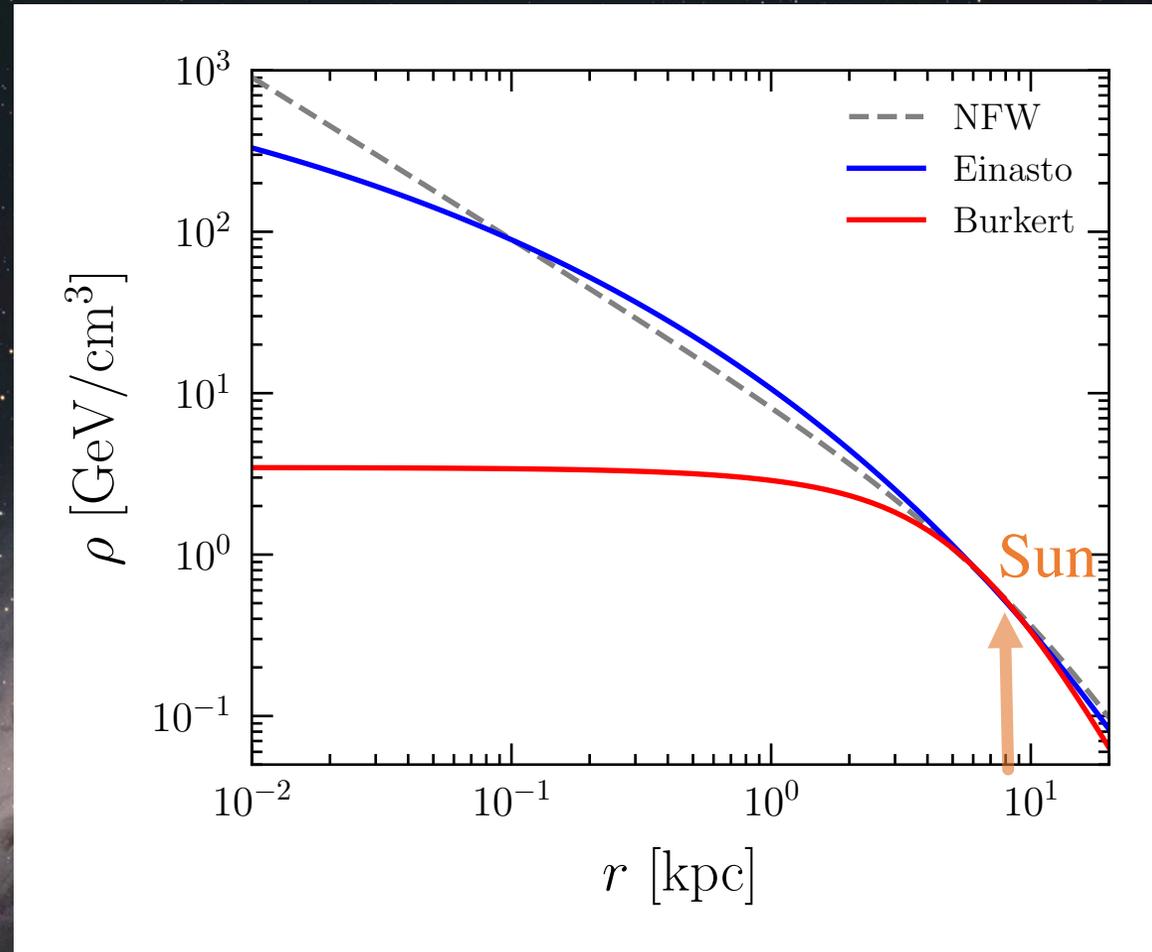
Dark Matter Density Distribution



Einasto (1965)
Navarro, Frenk, White (1996)
Fornasa & Green (2013)
Gaskins (2016)

Dark Matter Density Distribution

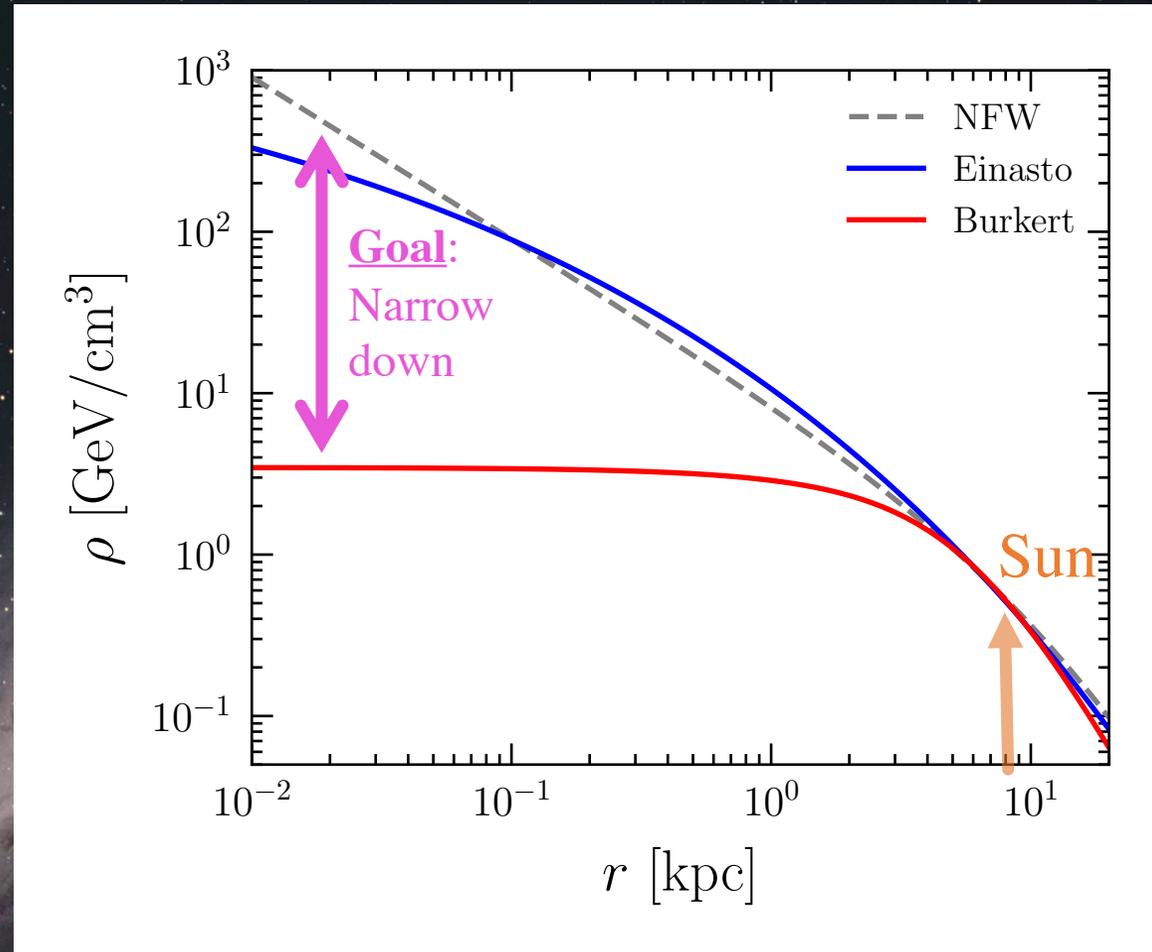
Density profiles can vary by orders of magnitude.



Einasto (1965)
Burkert (1995)
Navarro, Frenk, White (1996)
Fornasa & Green (2013)
Gaskins (2016)

Dark Matter Density Distribution

Density profiles can vary by orders of magnitude.



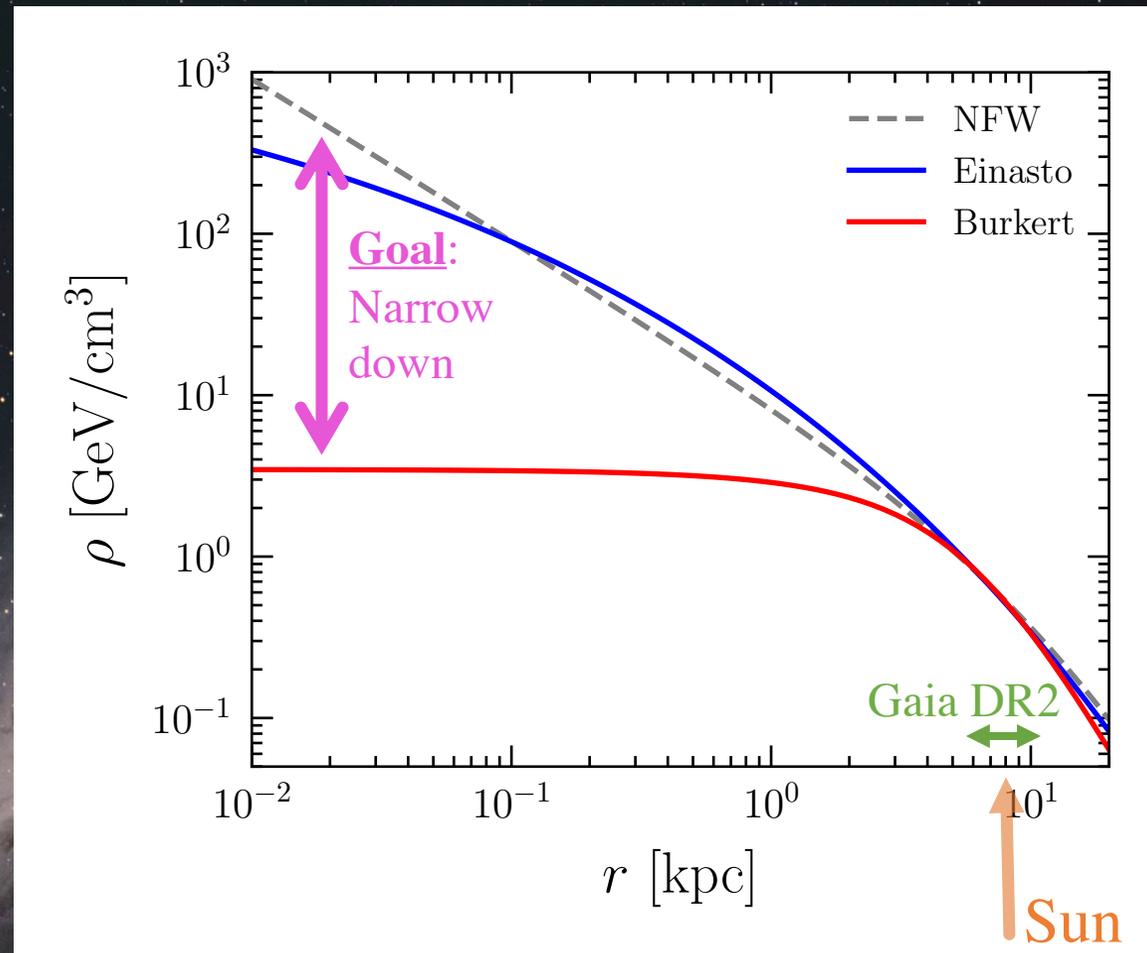
Einasto (1965)
Burkert (1995)
Navarro, Frenk, White (1996)
Fornasa & Green (2013)
Gaskins (2016)

Dark Matter Density Distribution

Measure the
Escape
Velocity

Obtain the
Milky Way
Potential

Deduce the
Dark Matter
Density
Profile



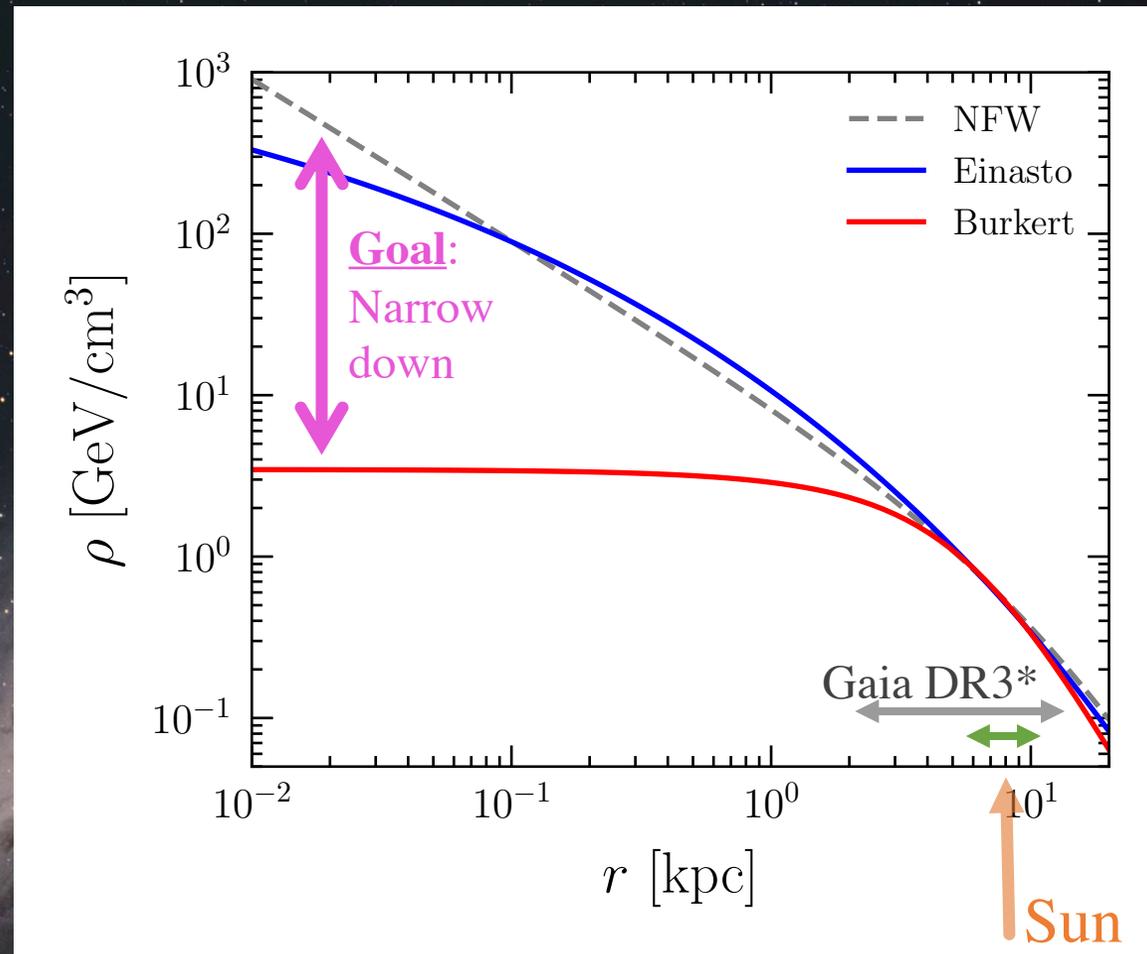
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(*Estimate)

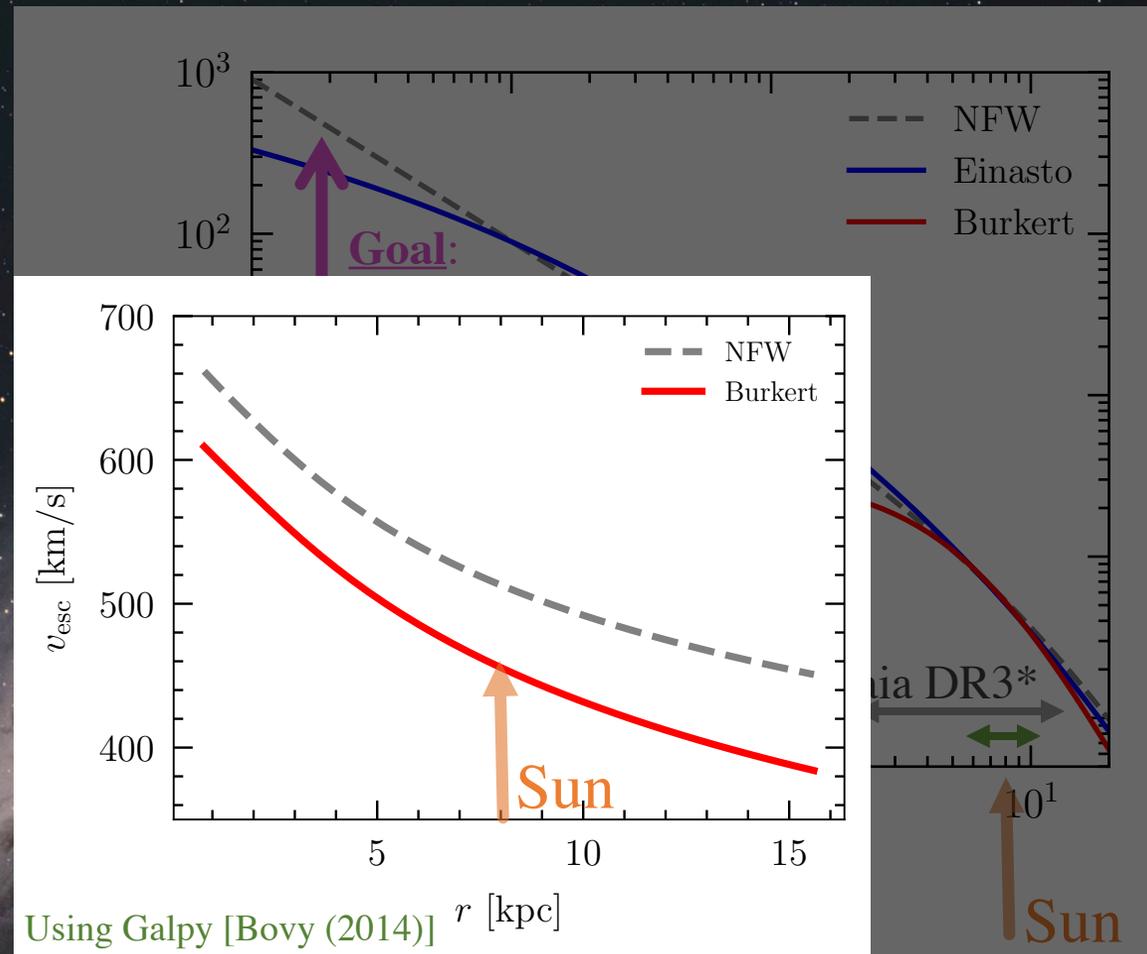
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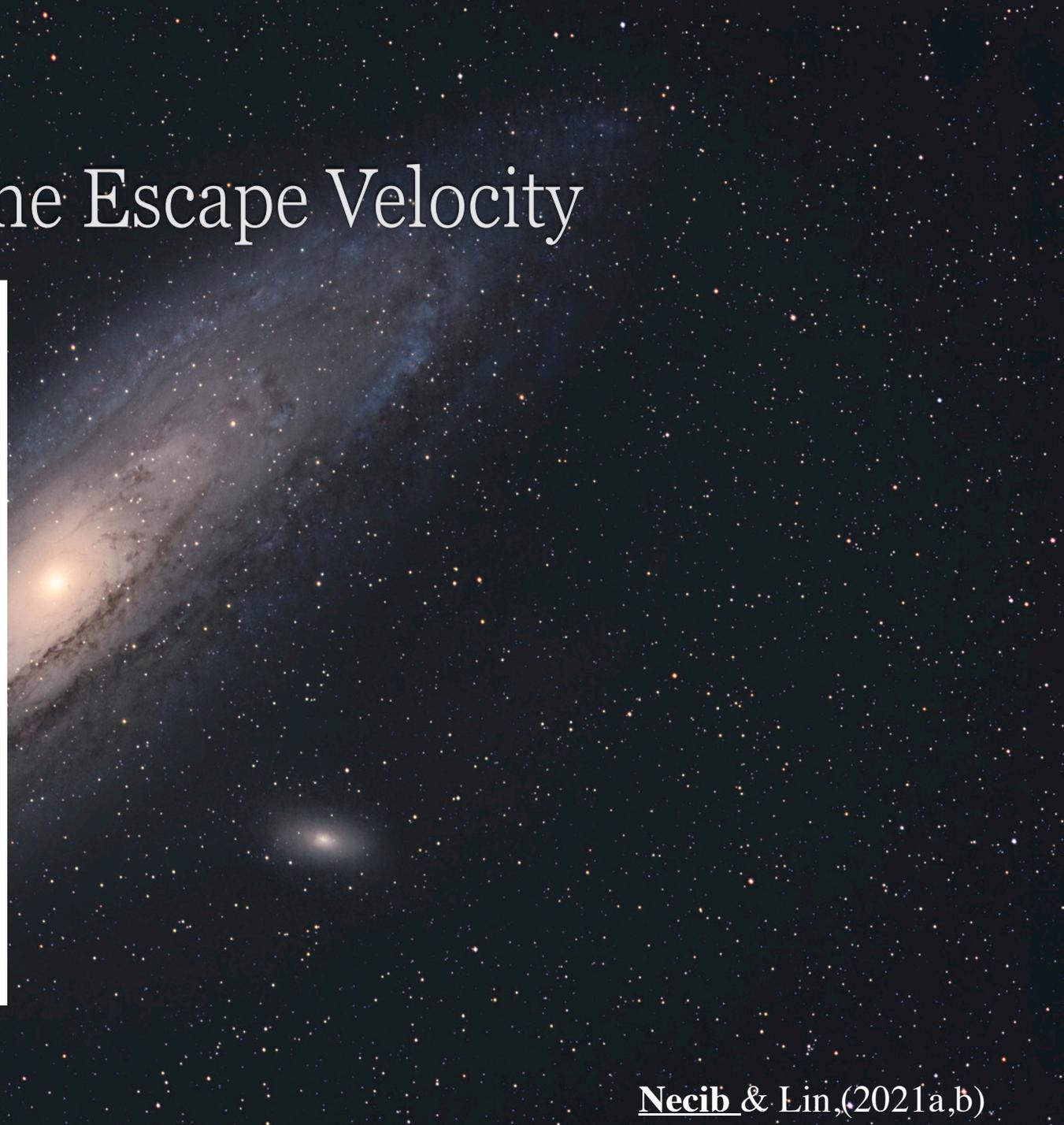
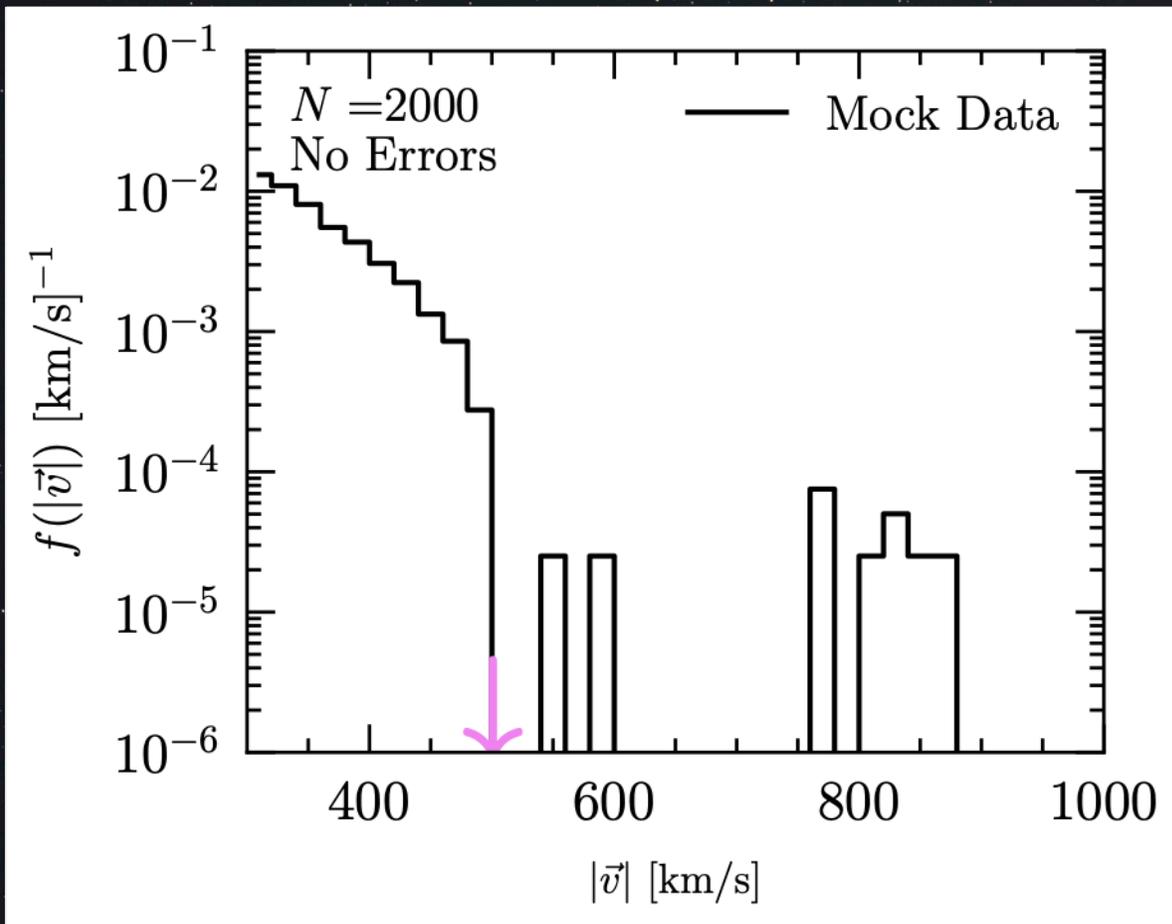
Deduce the
Dark Matter
Density
Profile



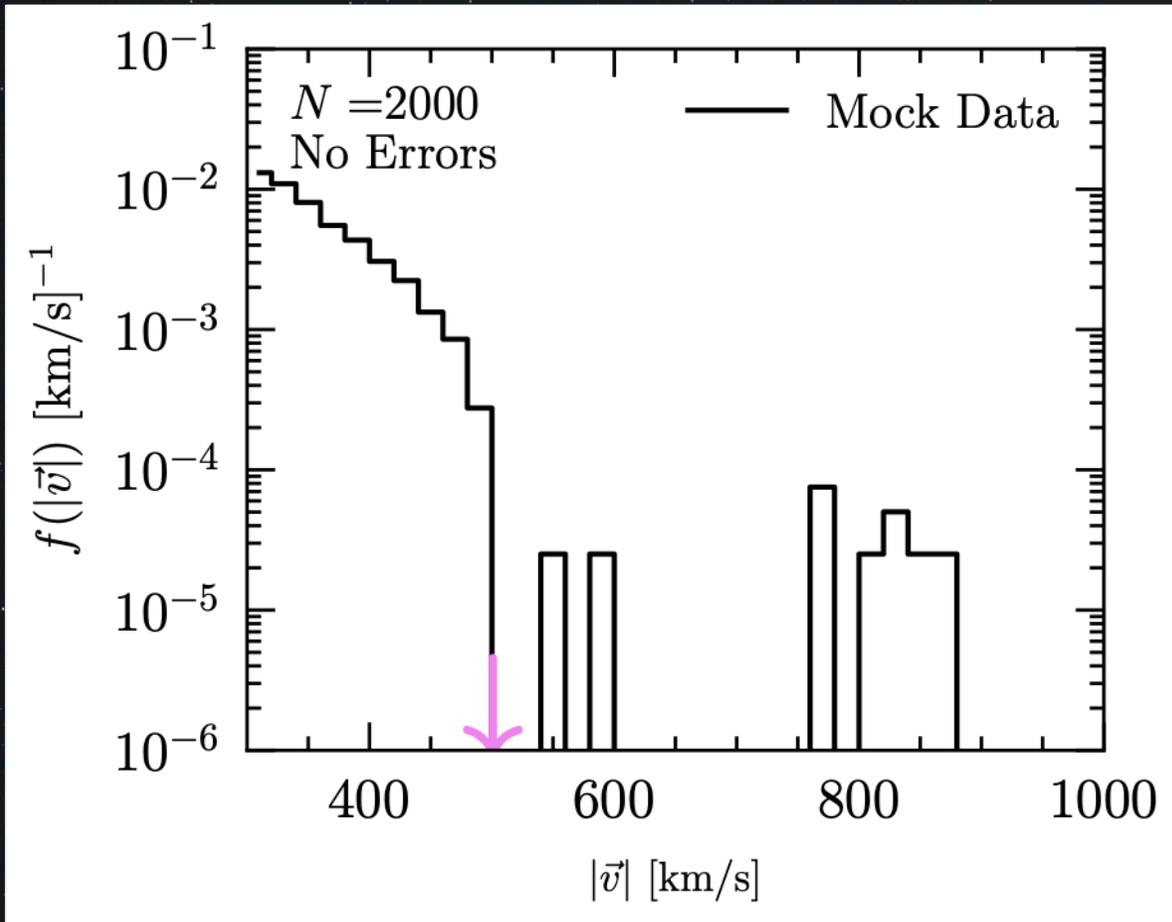
(*Estimate)

Einasto (1965)
Burkert (1995)
Navarro, Frenk, White (1996)
Fornasa & Green (2013)
Gaskins (2016)

Determining the Escape Velocity

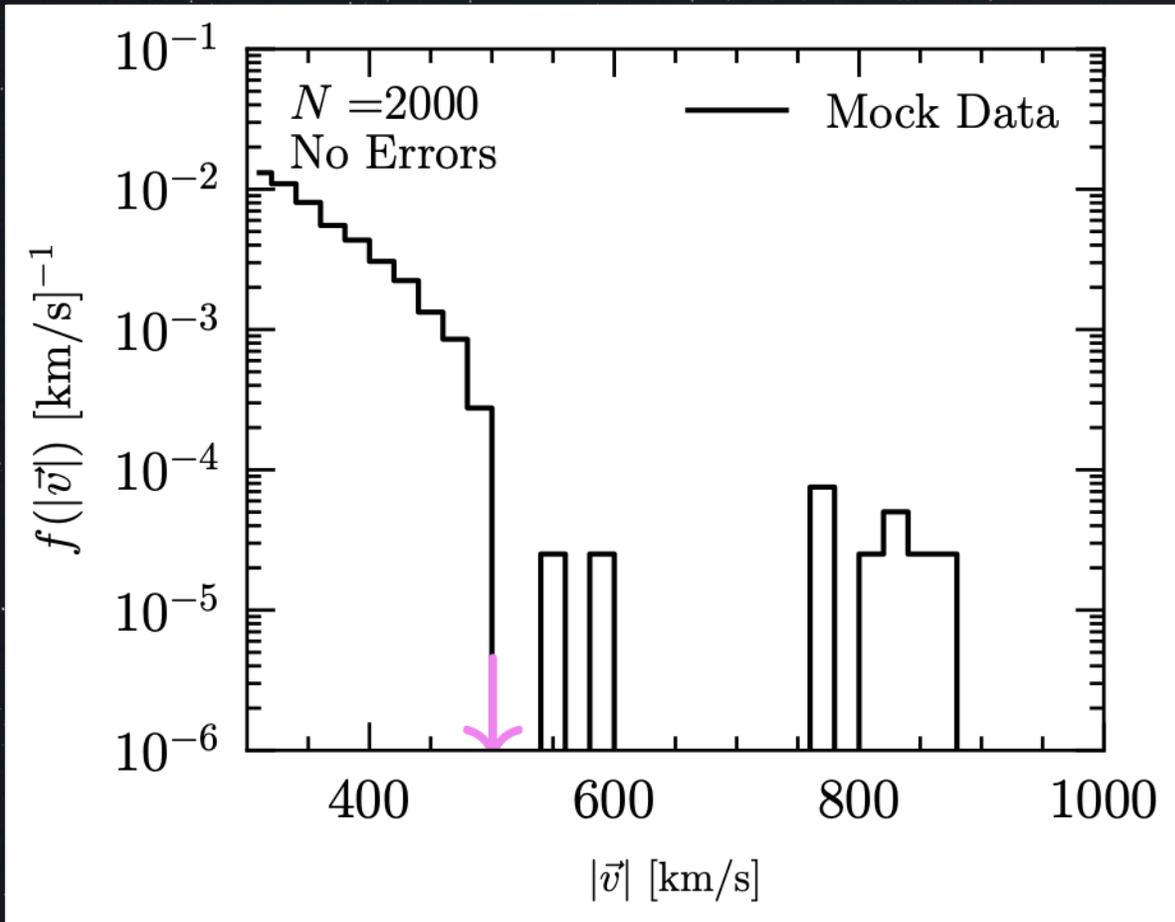


Determining the Escape Velocity



Goal:
Fit for the escape velocity

Determining the Escape Velocity

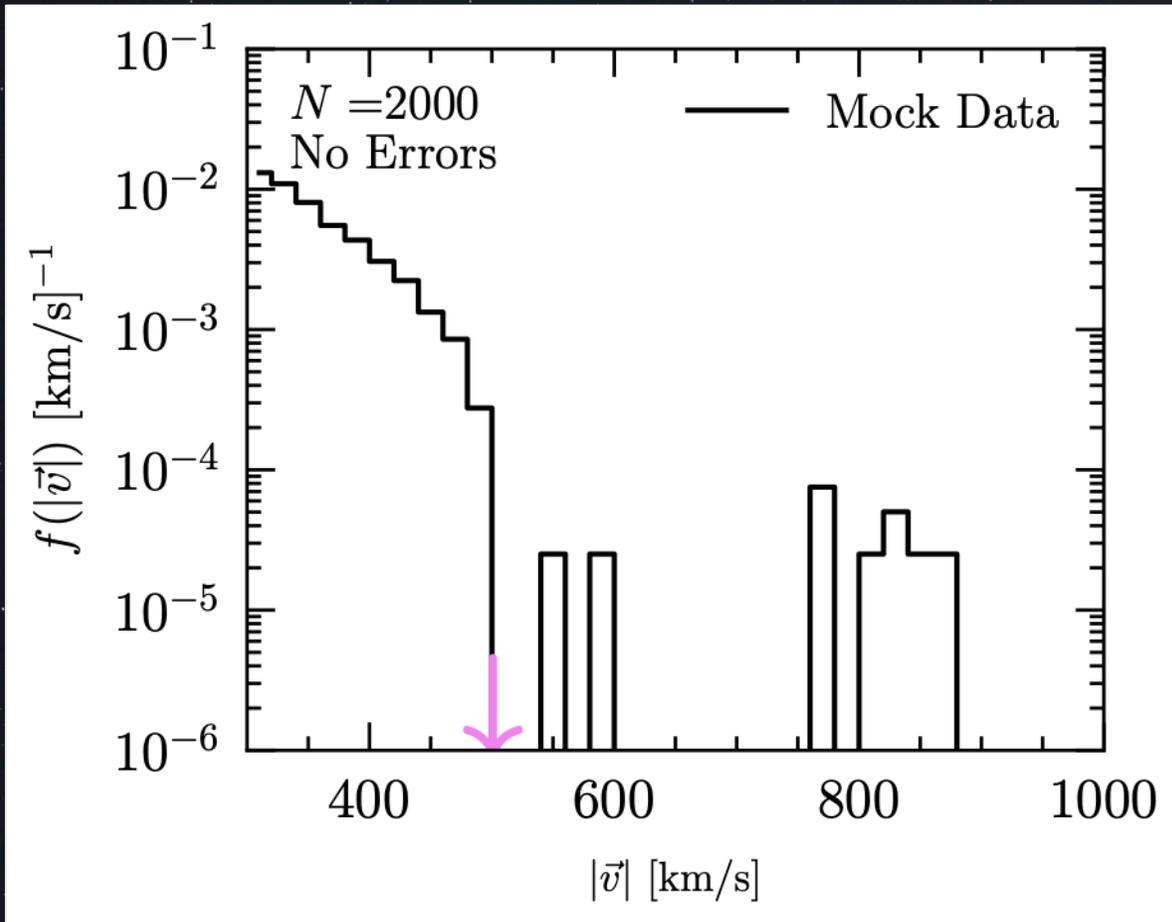


Goal:
Fit for the escape velocity

$$f(v) \propto (v_{\text{esc}} - v)^k$$

$$v > v_{\text{min}}$$

Determining the Escape Velocity



Goal:

Fit for the escape velocity

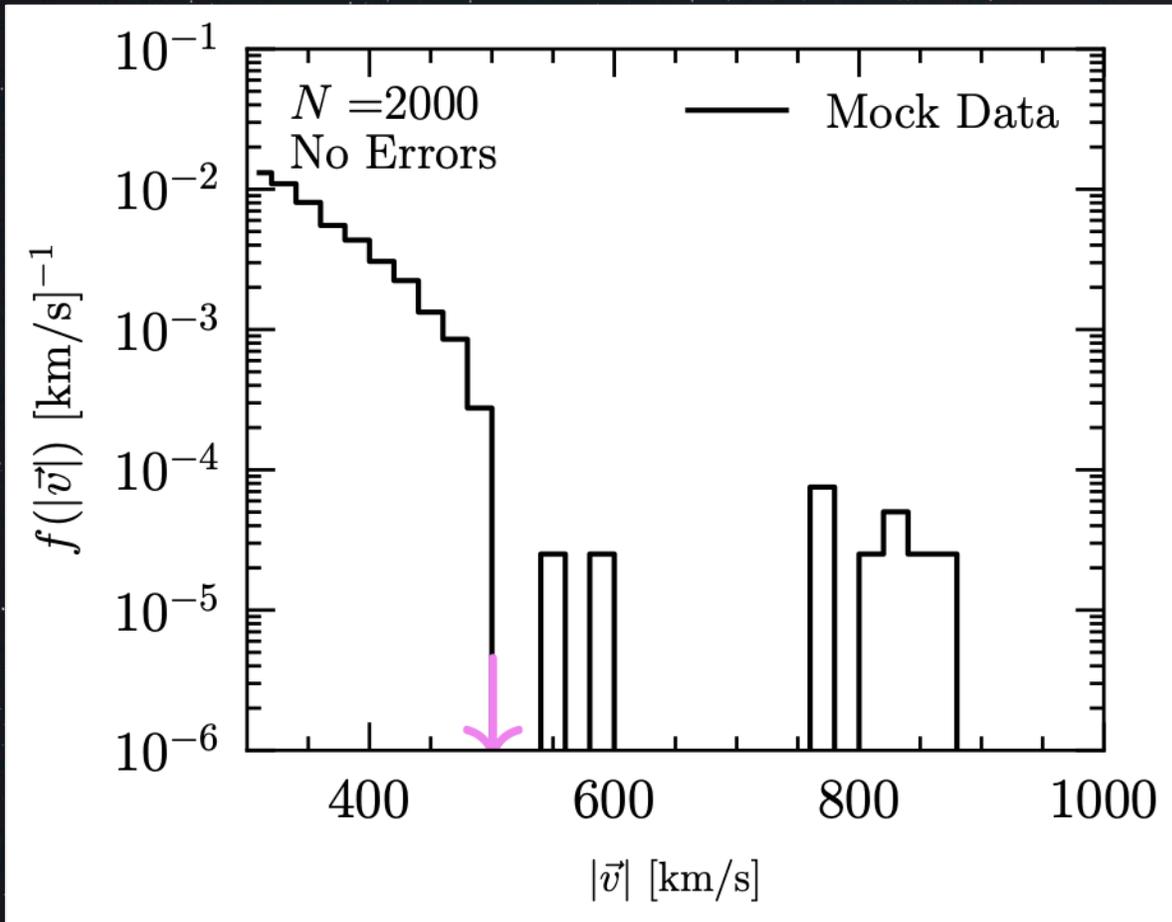
$$f(v) \propto (v_{\text{esc}} - v)^k$$

$$v > v_{\text{min}}$$

Complications:

1. Take into account errors.
2. Presence of outliers.
3. Presence of multiple components.

Determining the Escape Velocity



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Fit for the escape velocity

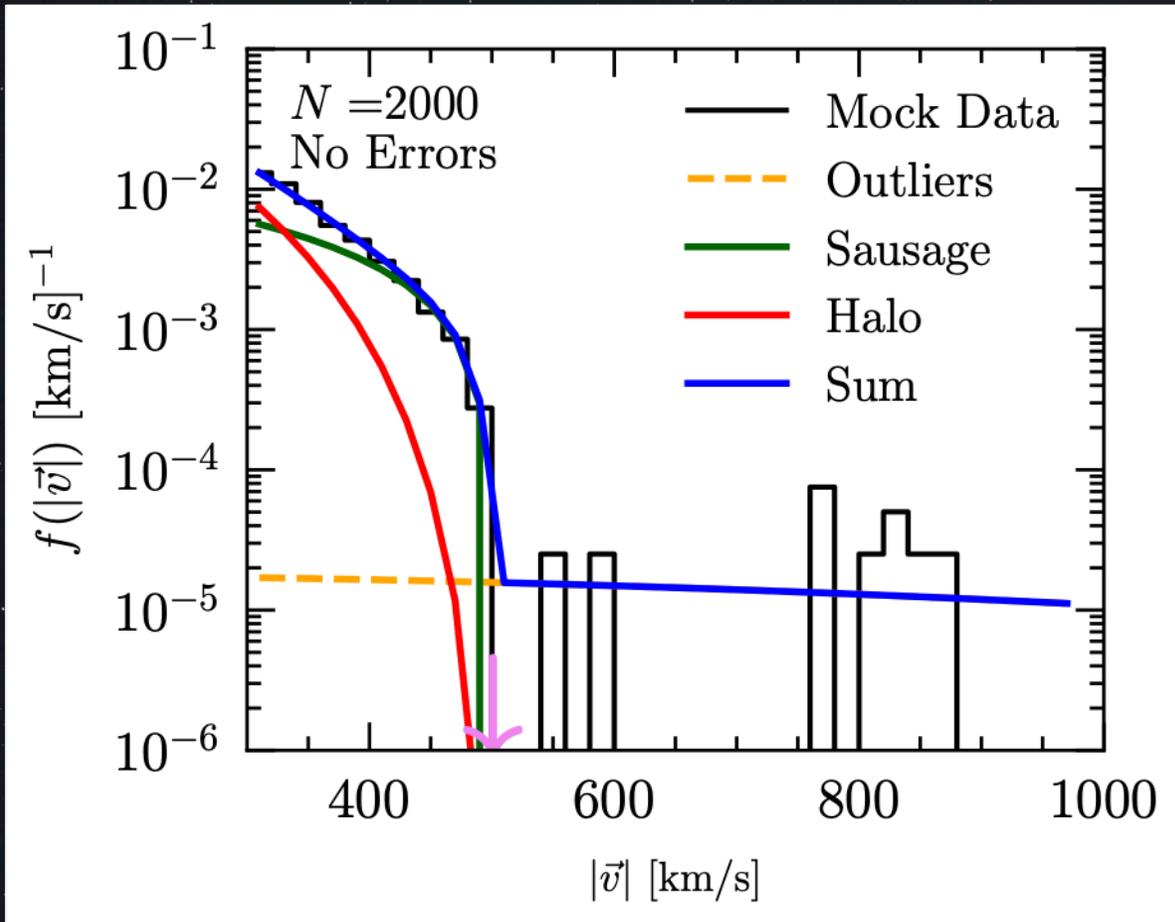
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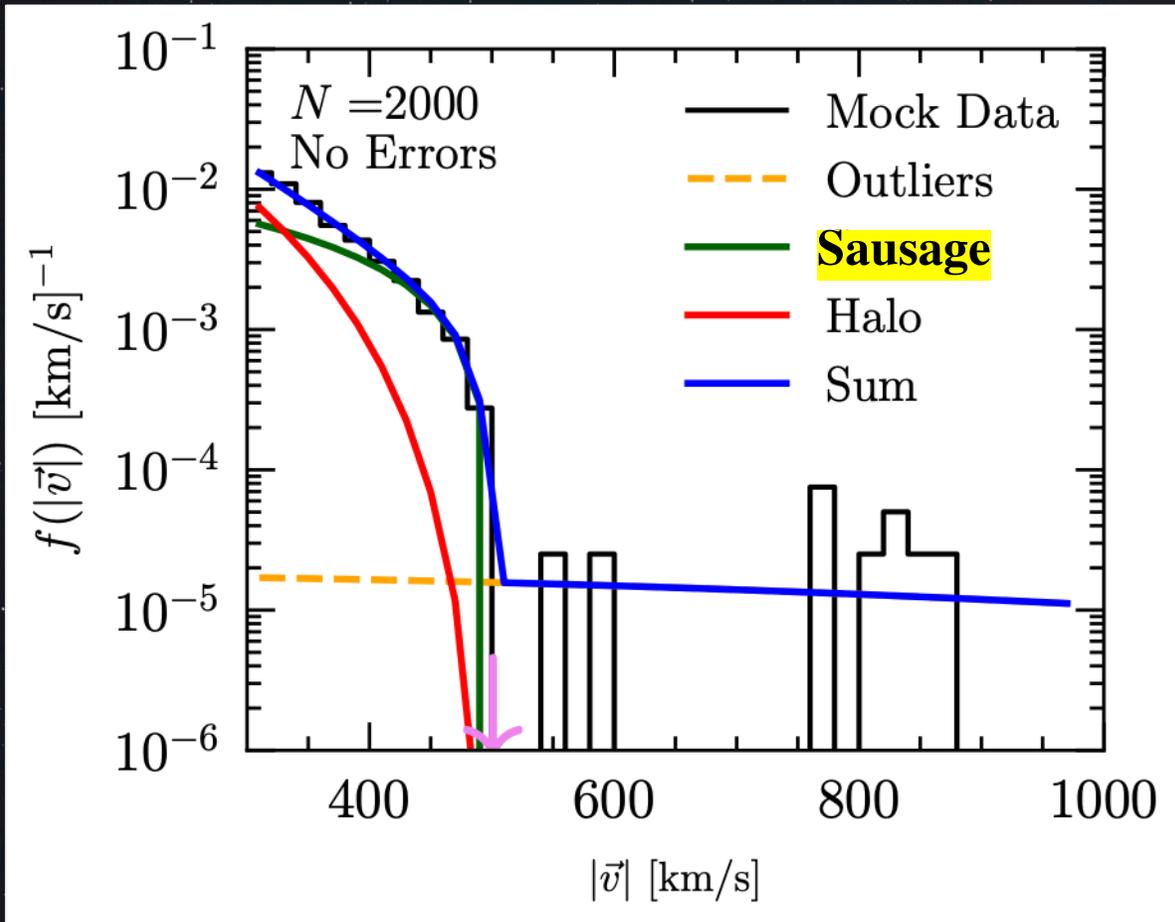
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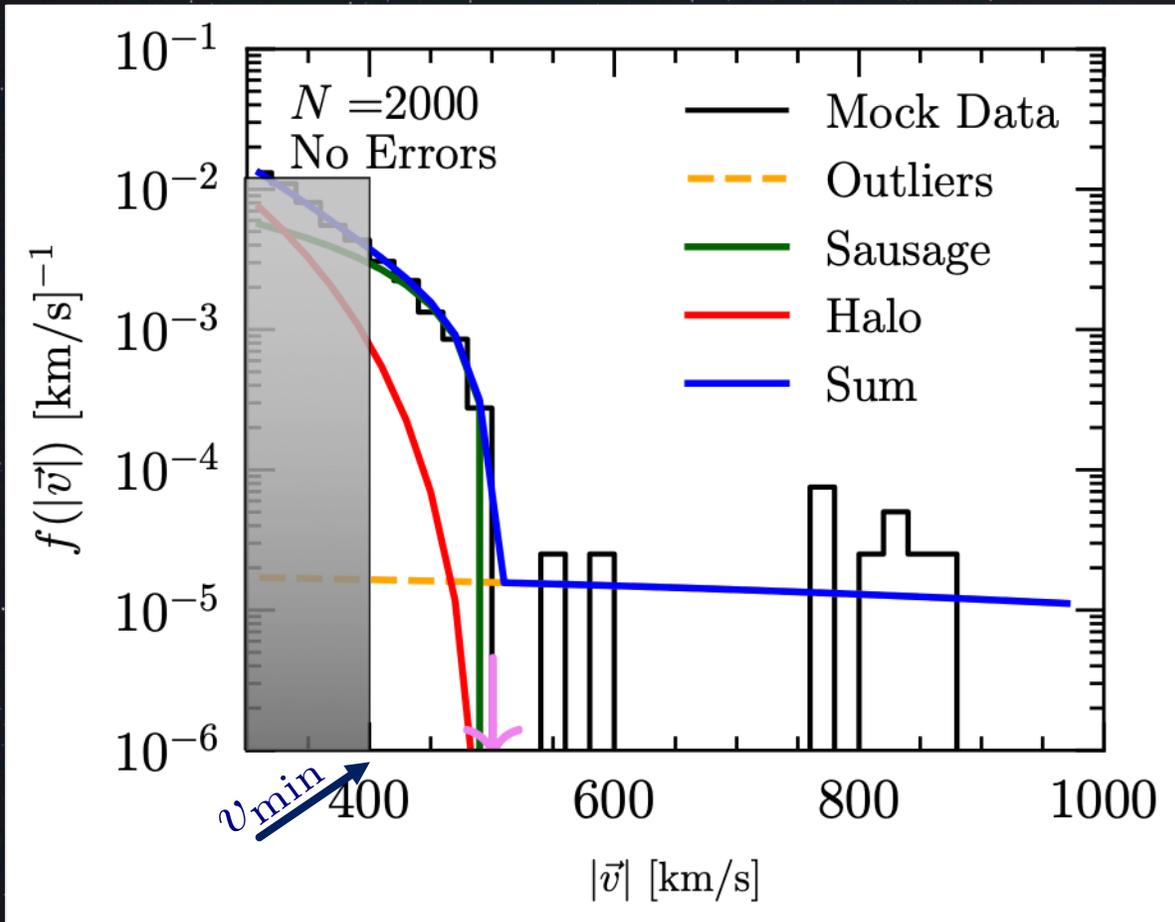
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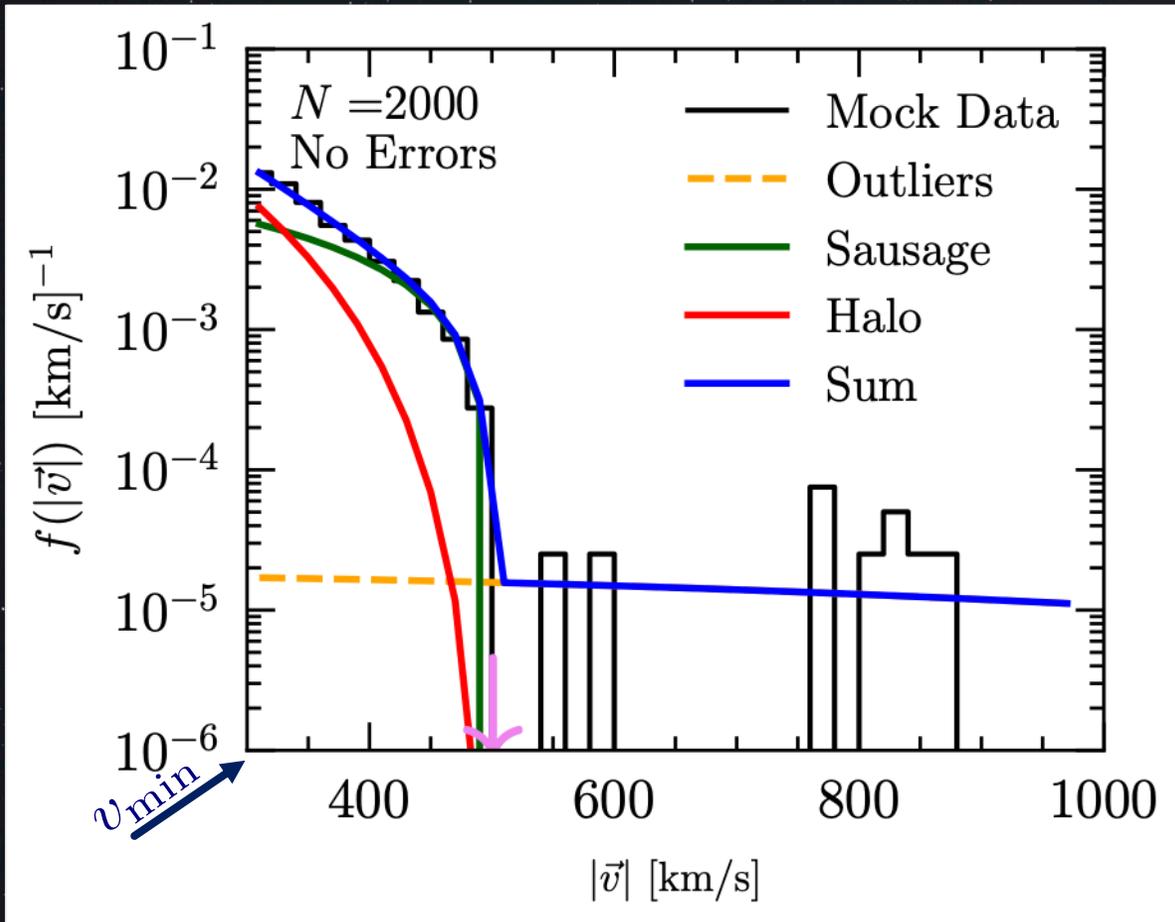
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$$v > v_{\min}$$

Complications:

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Goal:

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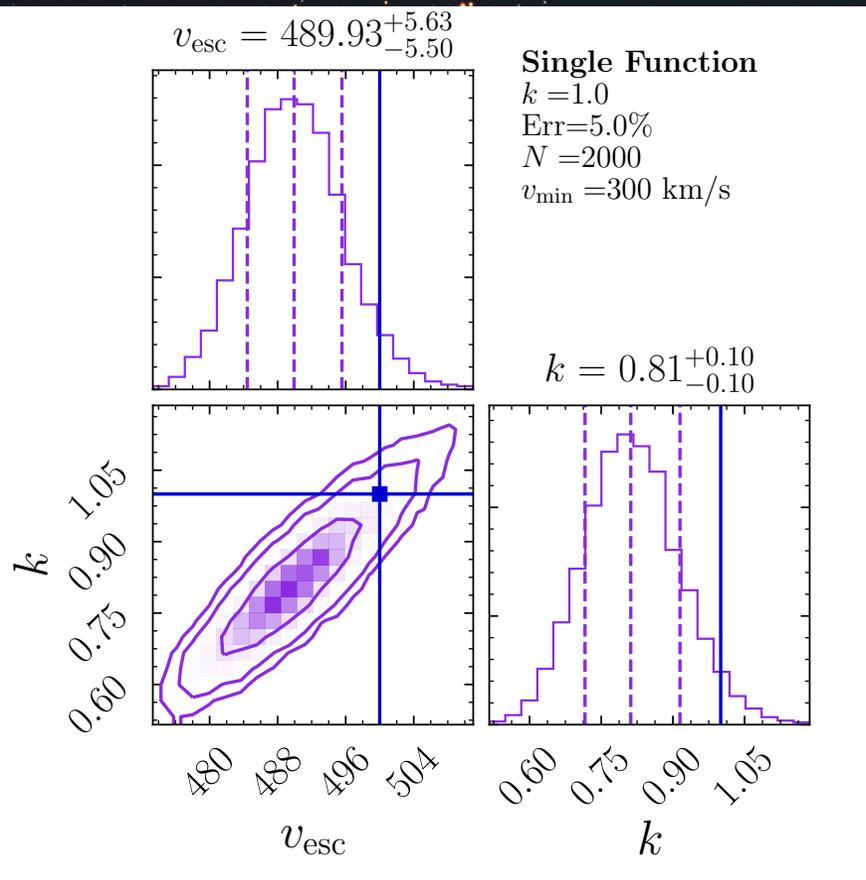
Leonard & Tremaine (1990)

Necib & Lin (2021a,b)

Determining the Escape Velocity

$$f(v) \propto (v_{\text{esc}} - v)^k$$

$$v > v_{\text{min}}$$



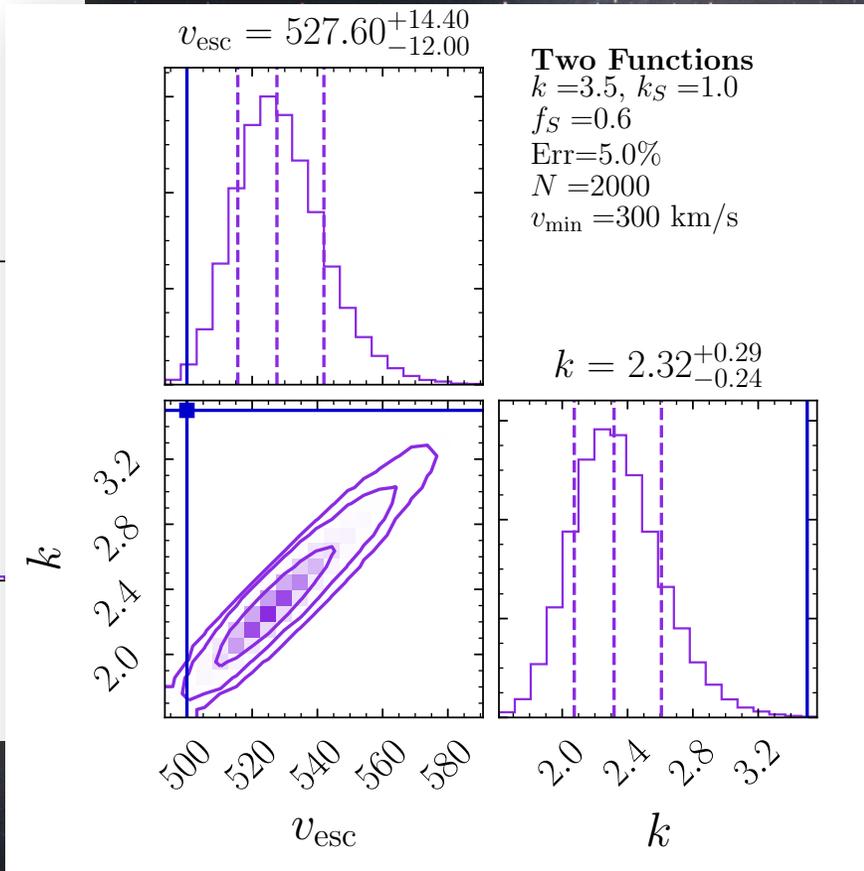
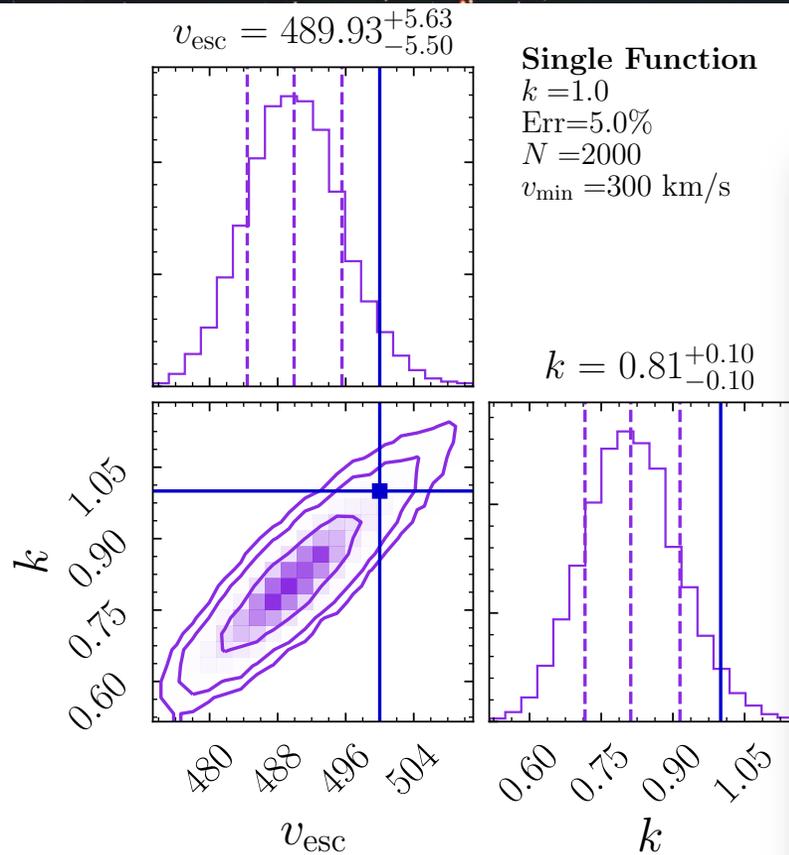
If the real function is dominated by a single distribution, the fit works.

Determining the Escape Velocity

$$f(v) \propto (v_{\text{esc}} - v)^k$$

$$v > v_{\text{min}}$$

If the real function is the sum of two functions, we do not get the right answer.



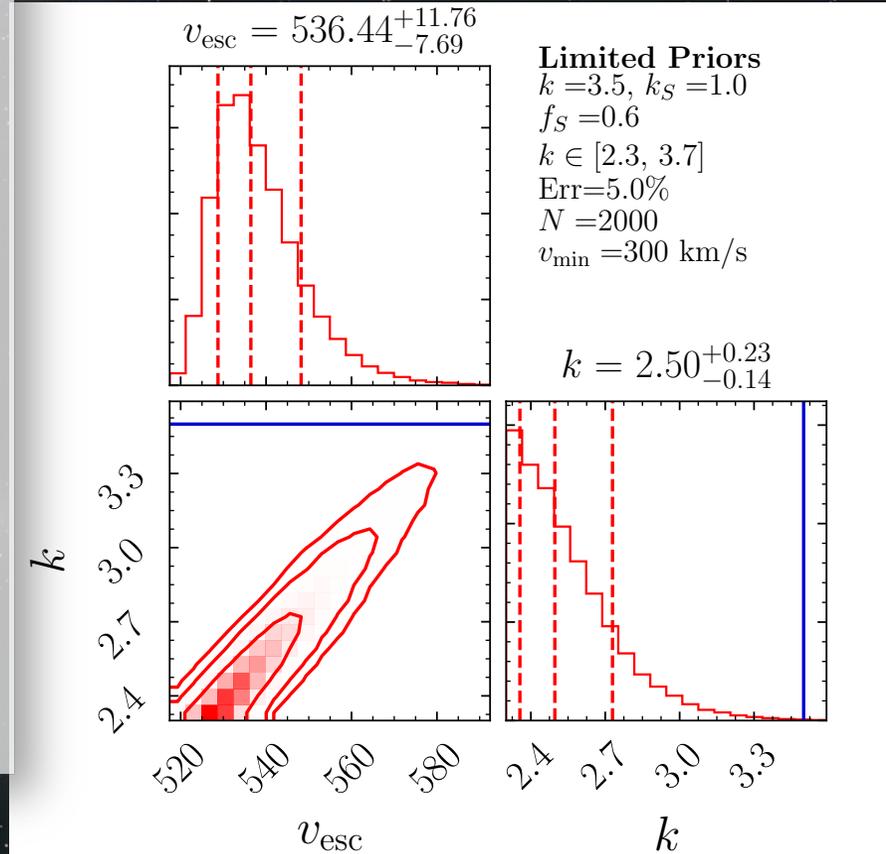
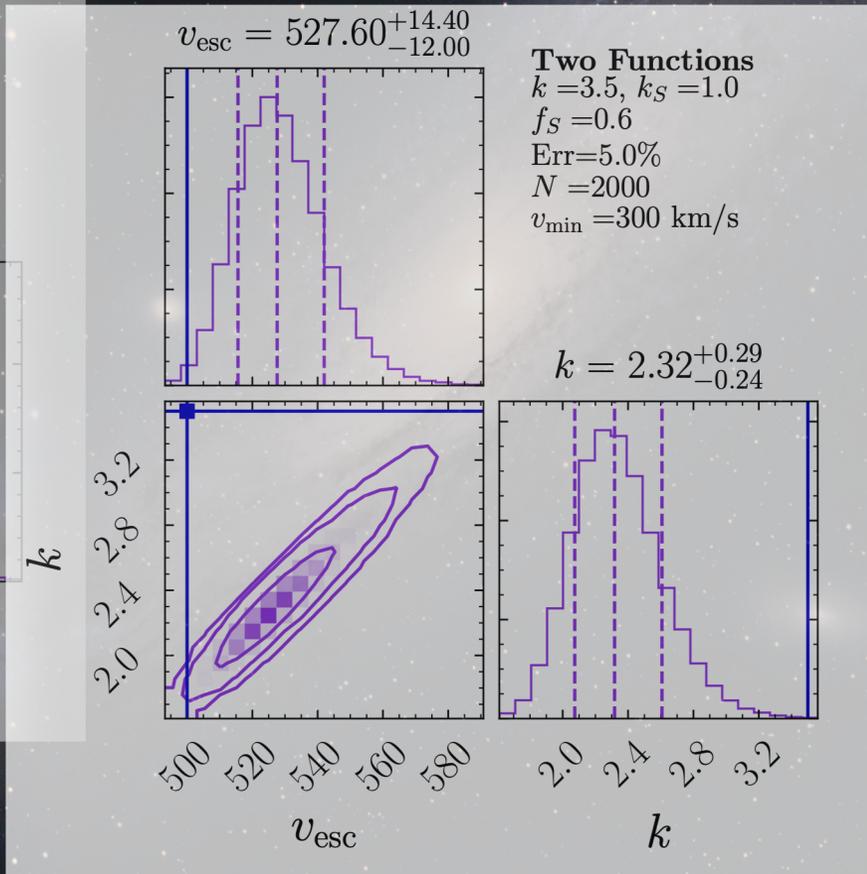
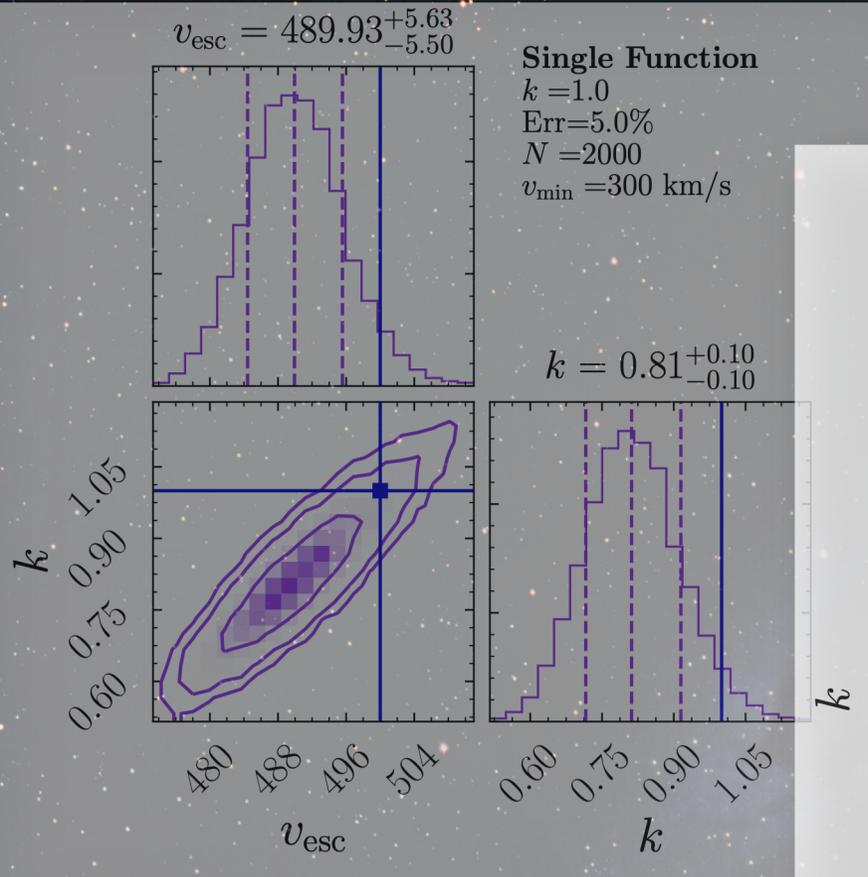
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Determining the Escape Velocity

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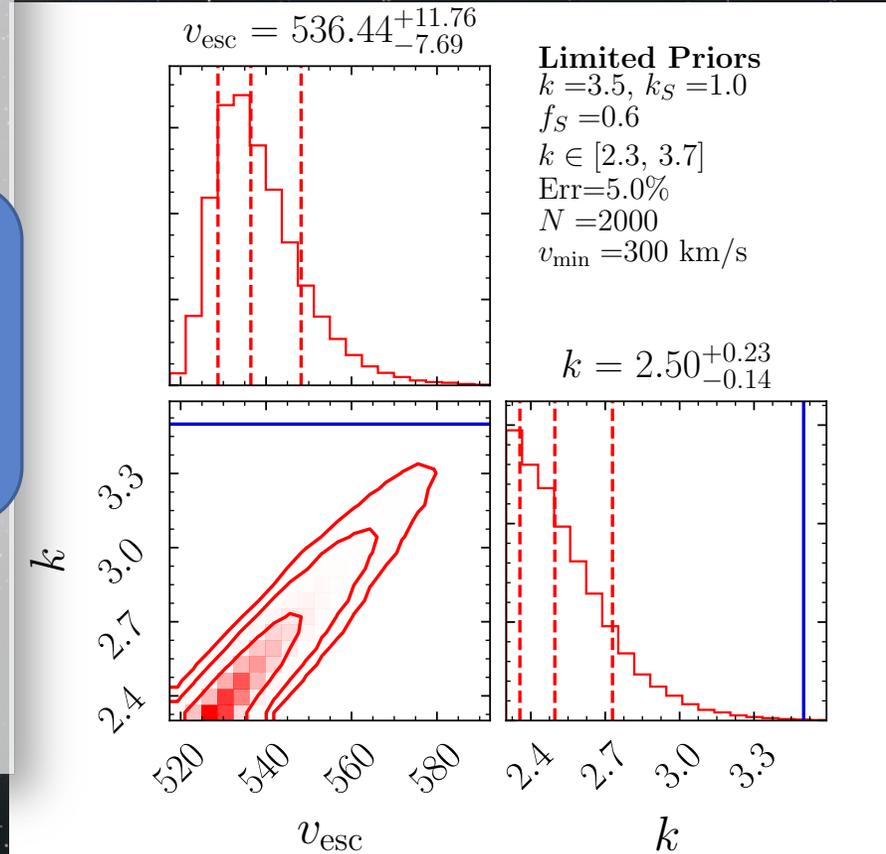
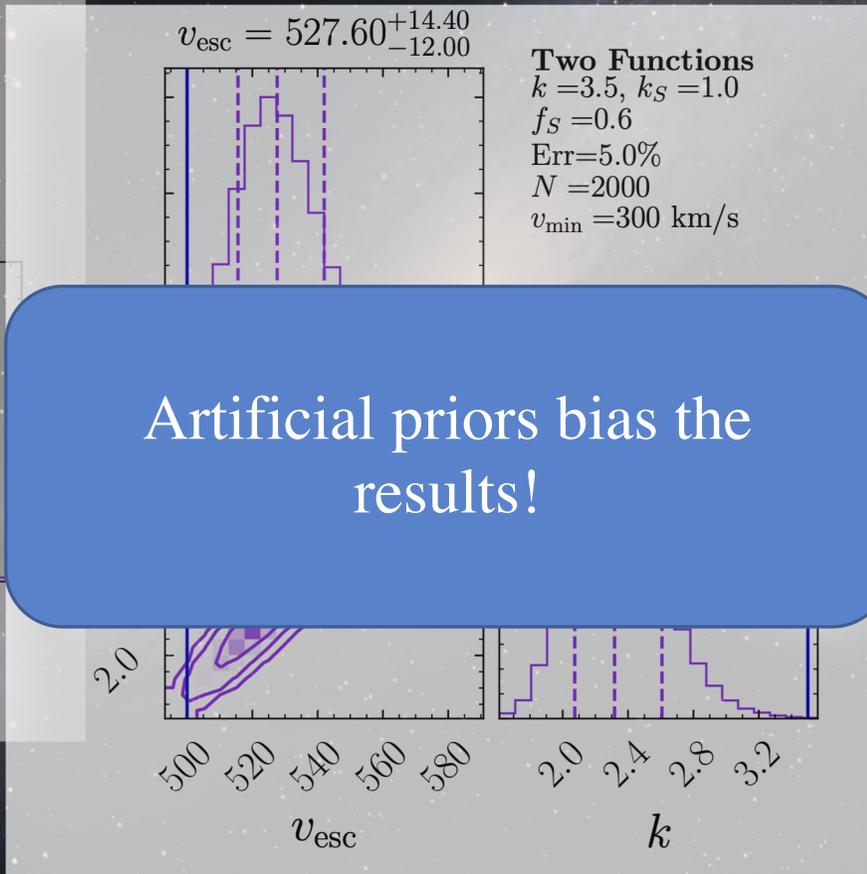
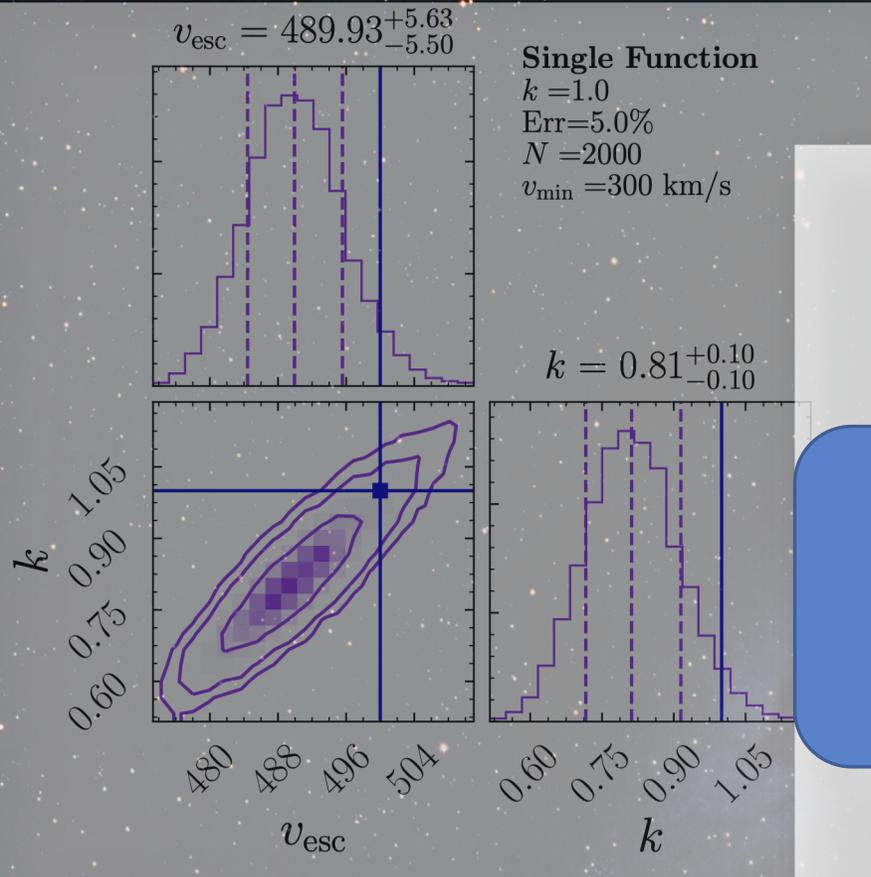
$$v > v_{\text{min}}$$



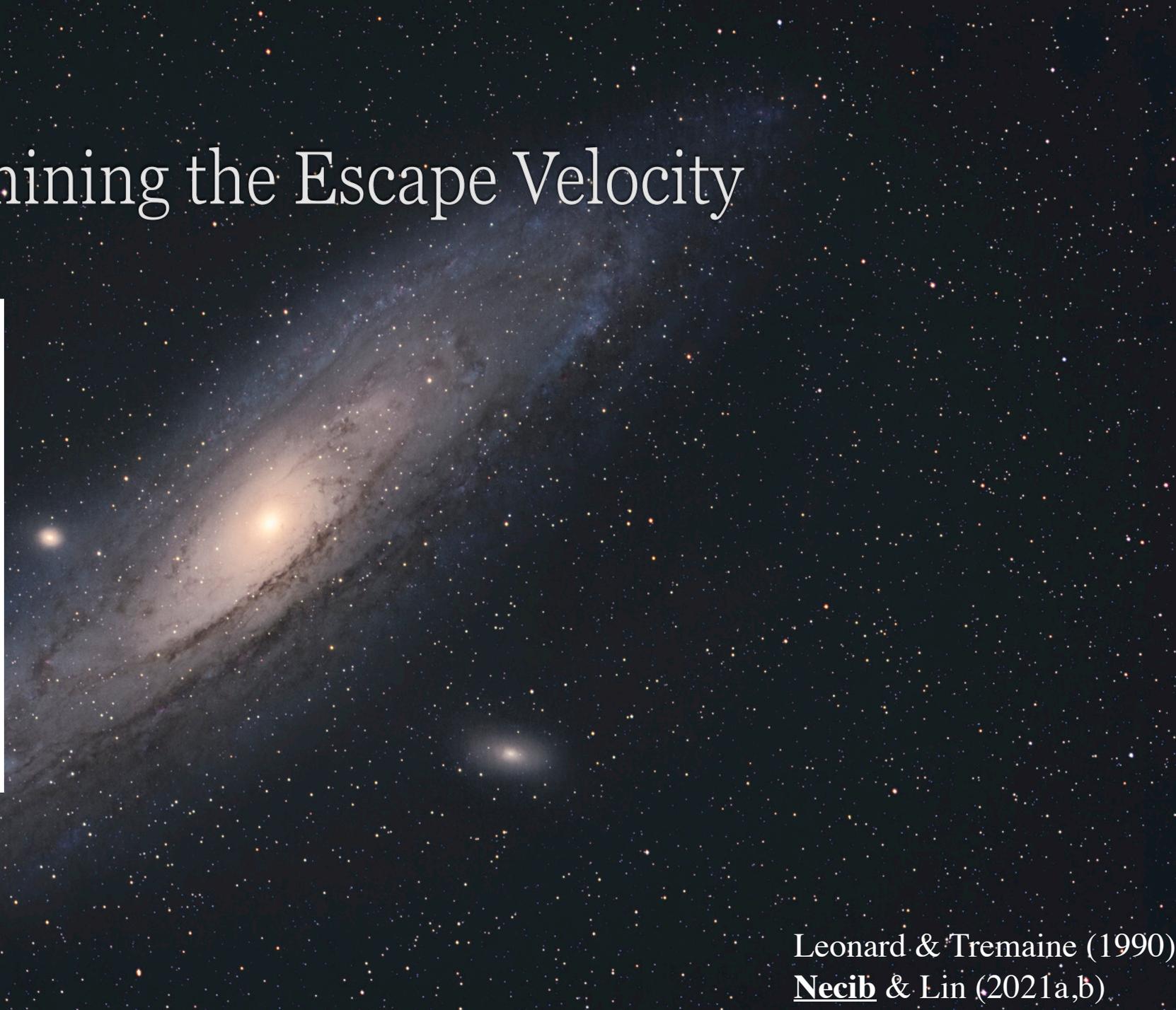
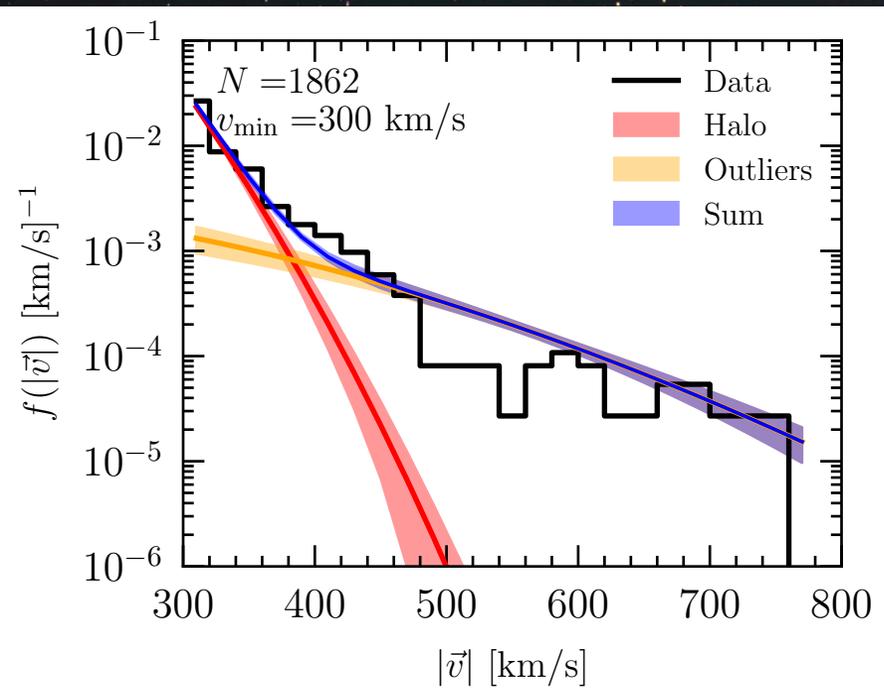
Determining the Escape Velocity

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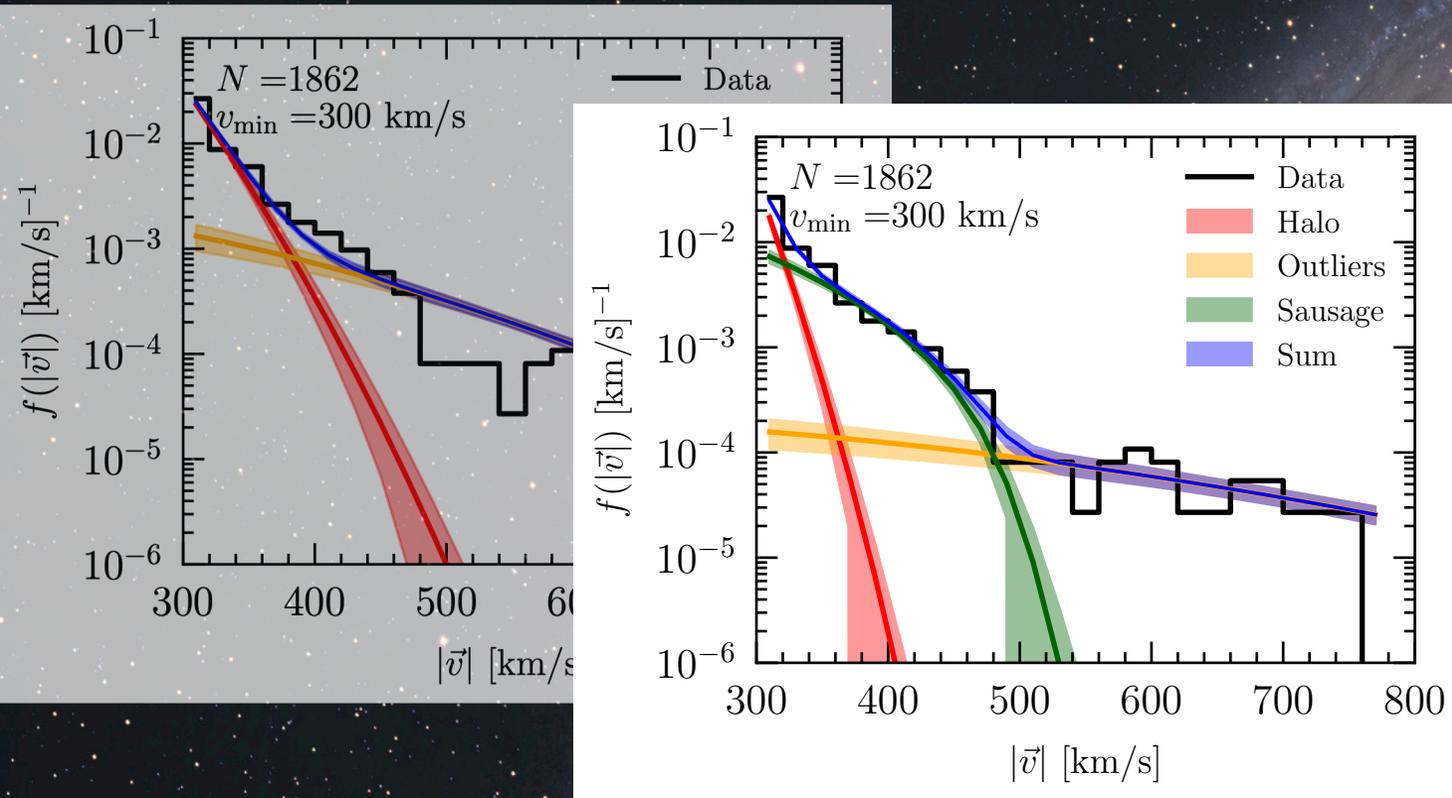
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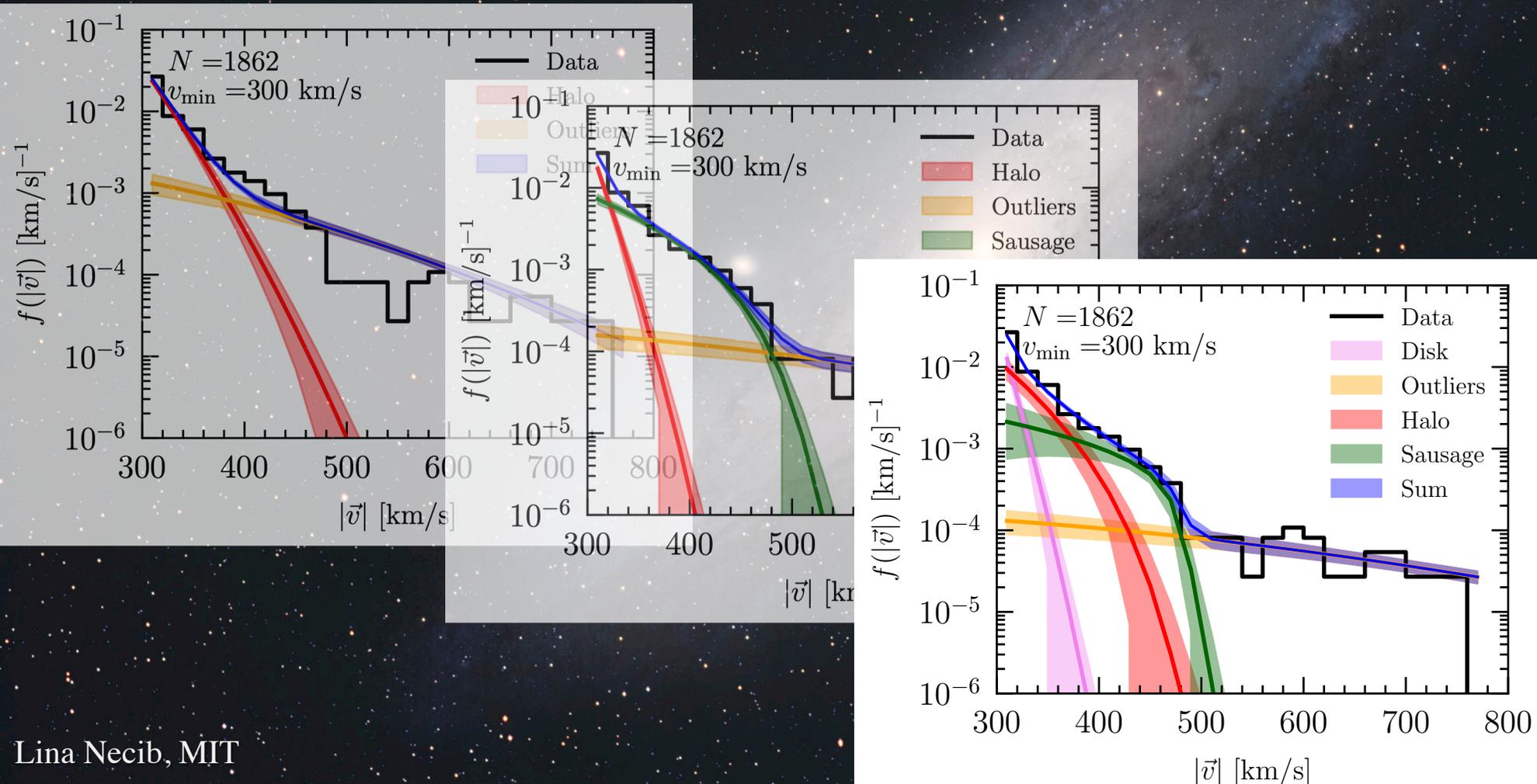
Determining the Escape Velocity



Determining the Escape Velocity

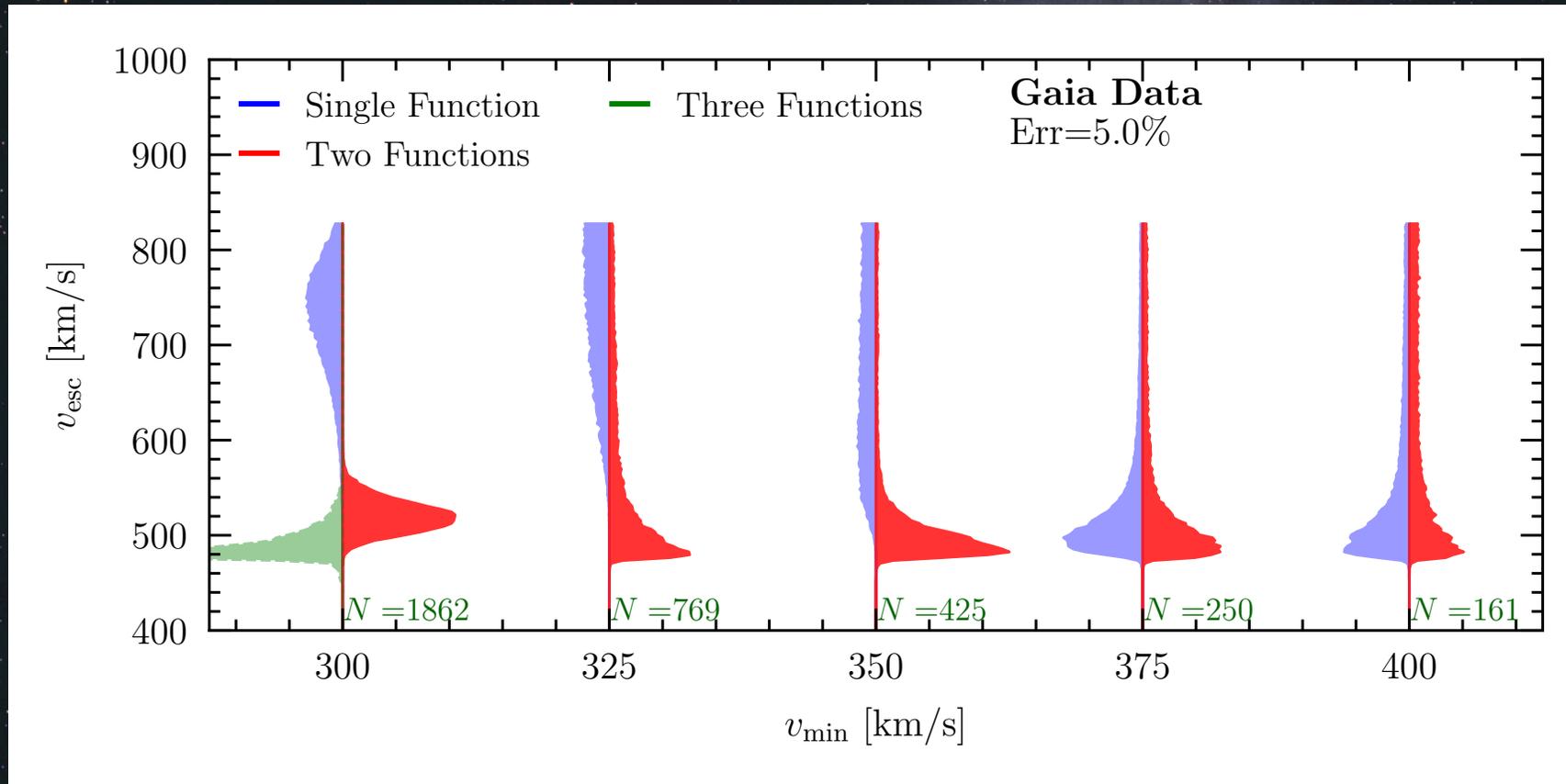


Determining the Escape Velocity

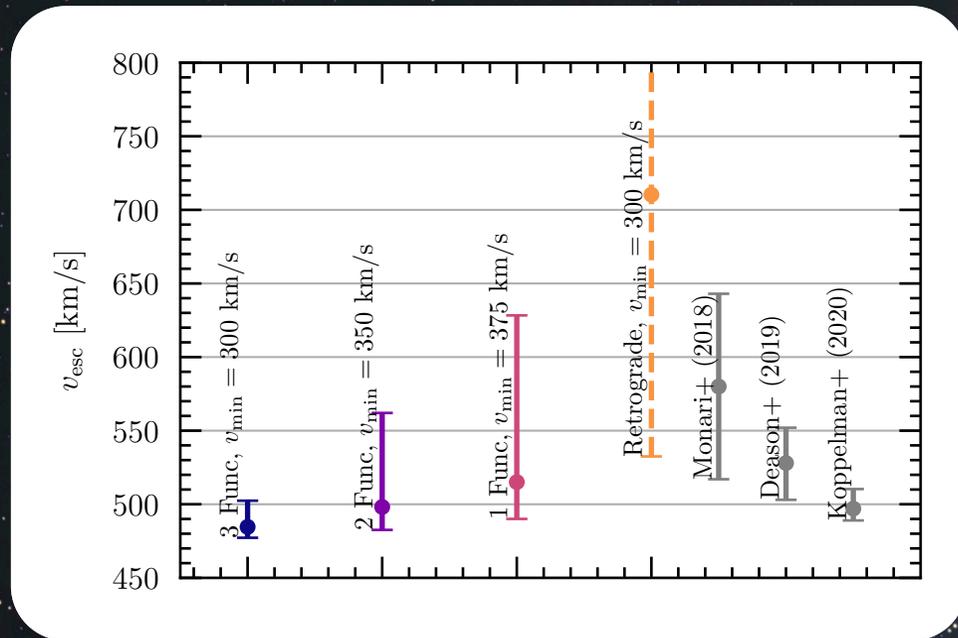


Leonard & Tremaine (1990)
Necib & Lin (2021a,b)

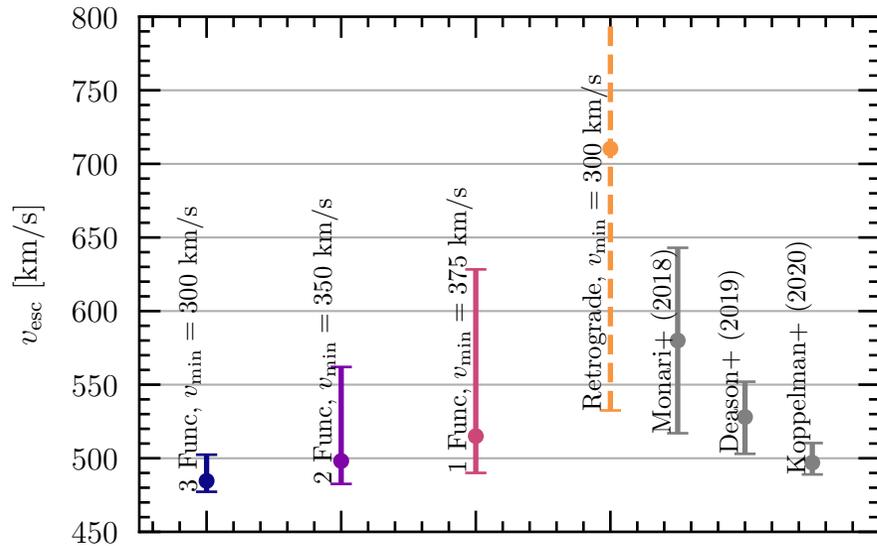
Determining the Escape Velocity



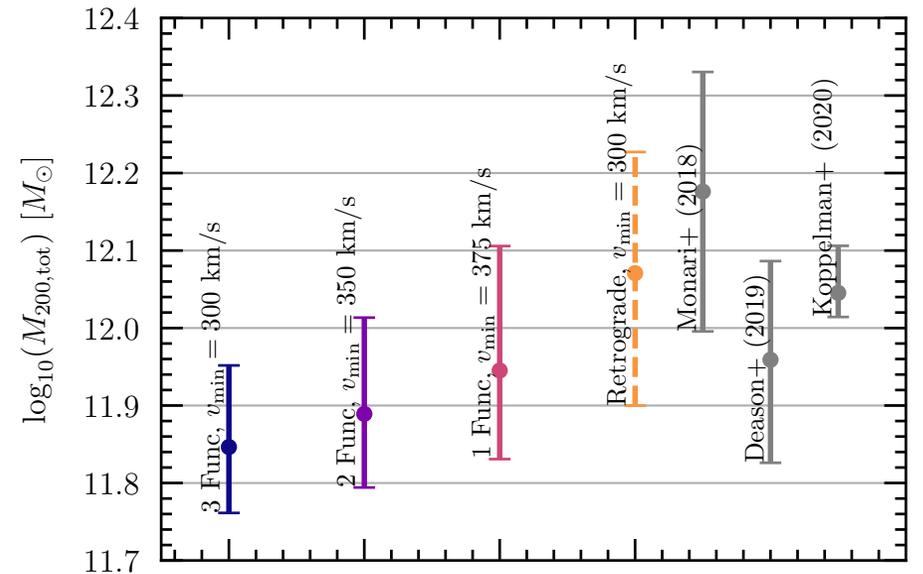
Determining the Escape Velocity



Determining the Escape Velocity



Assume
Milky Way
Potential



$$M_{200} = 7.0^{+1.9}_{-1.2} \times 10^{11} M_{\odot}$$

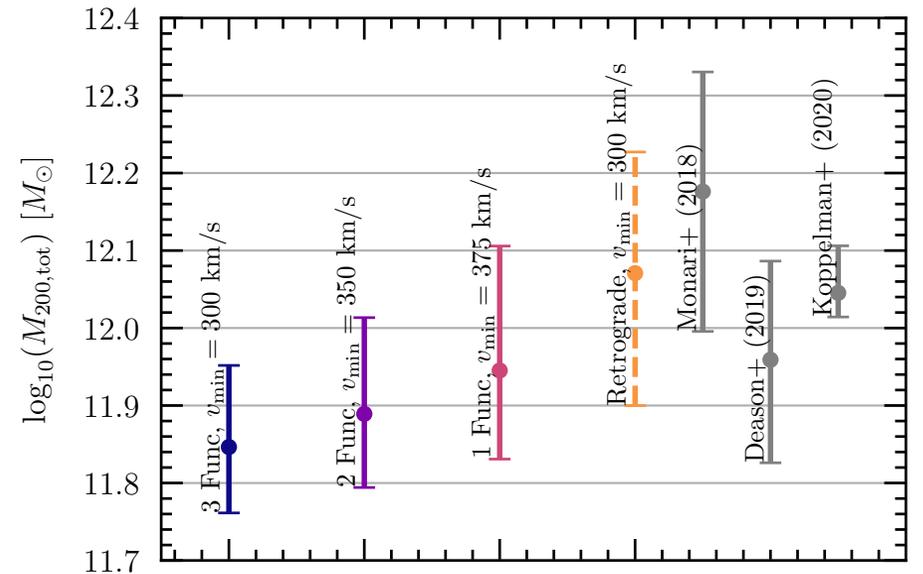
Leonard & Tremaine (1990)

Necib & Lin (2021a,b)

Determining the Escape Velocity

Stay Tuned for Full
Density Profile Results
and Updates with Gaia
DR3 in 2022!

Assume
Milky Way
Potential



$$M_{200} = 7.0^{+1.9}_{-1.2} \times 10^{11} M_{\odot}$$

Leonard & Tremaine (1990)

Necib & Lin (2021a,b)

Streams



Galactic
Center



Solar
Neighborhood



Pros

- Small Astrophysical Backgrounds

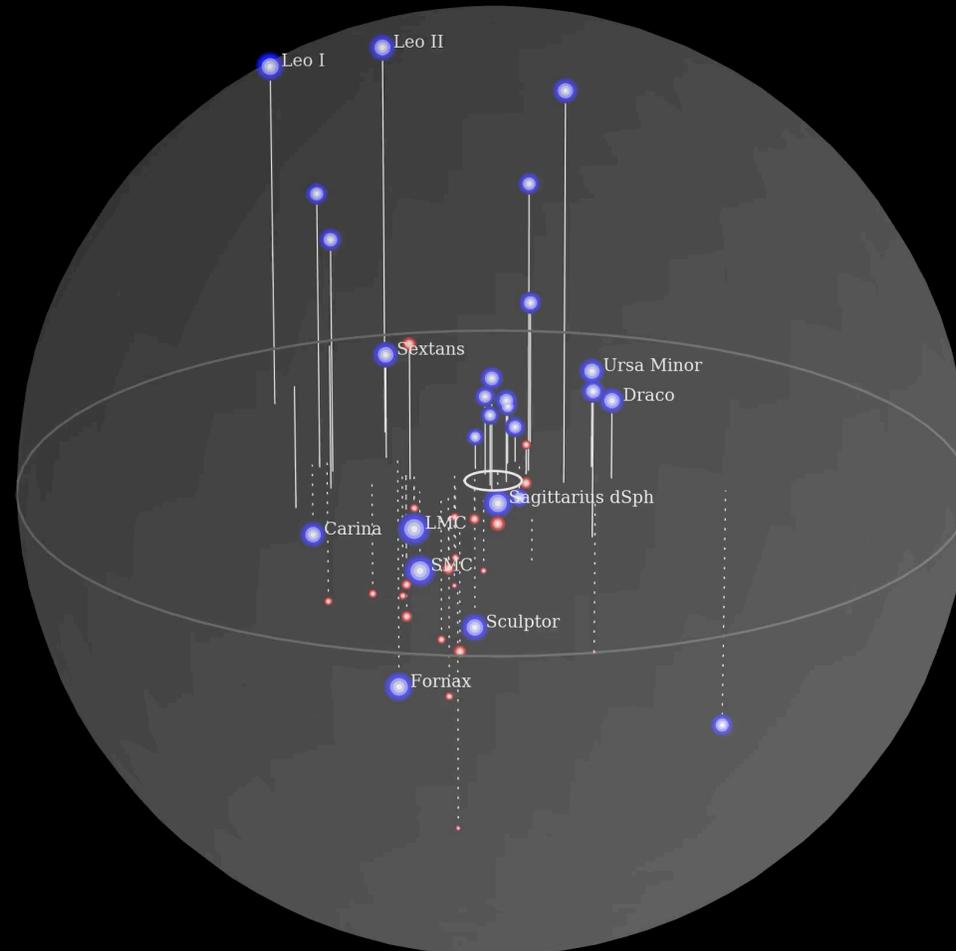
Cons

- Lower Dark Matter Density

Dwarf
Galaxies

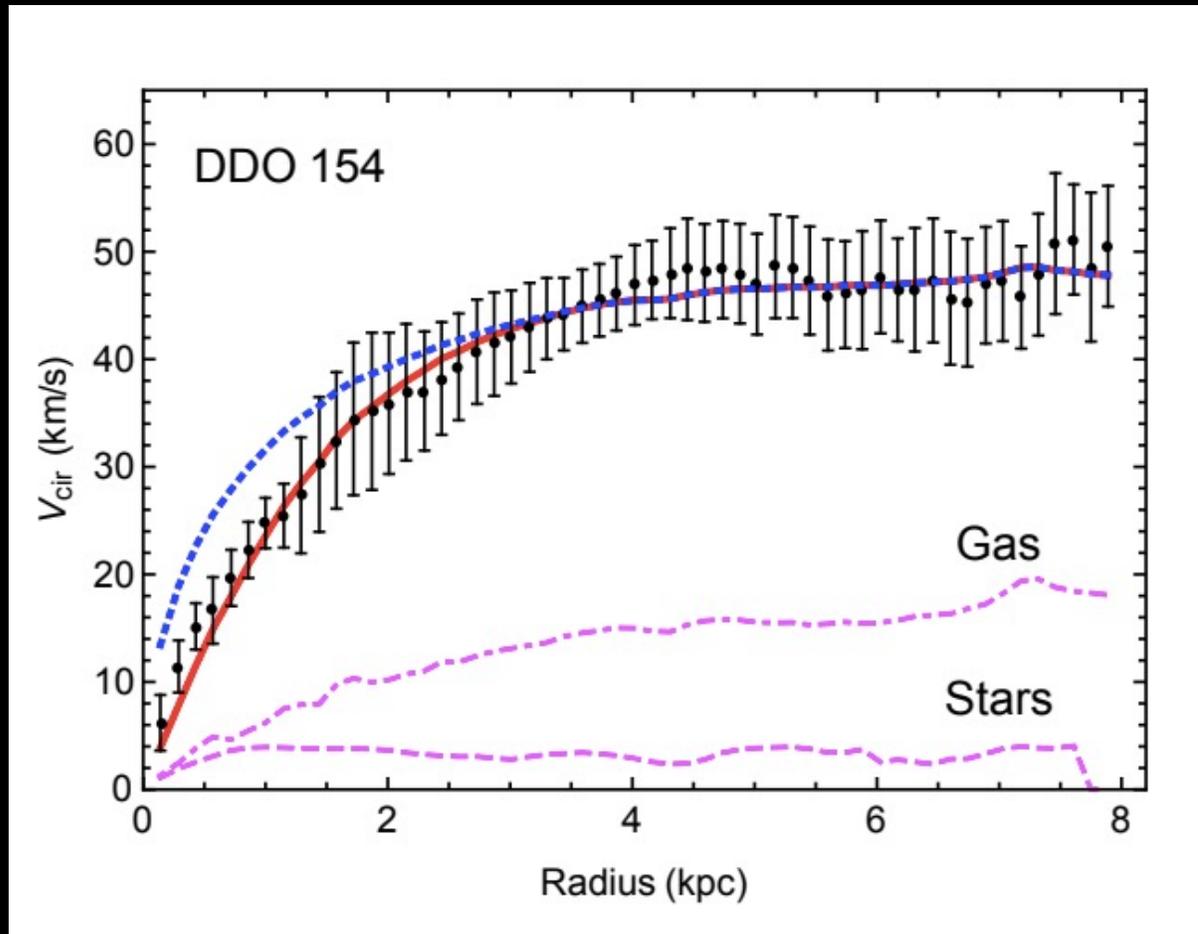


Dwarf Galaxies



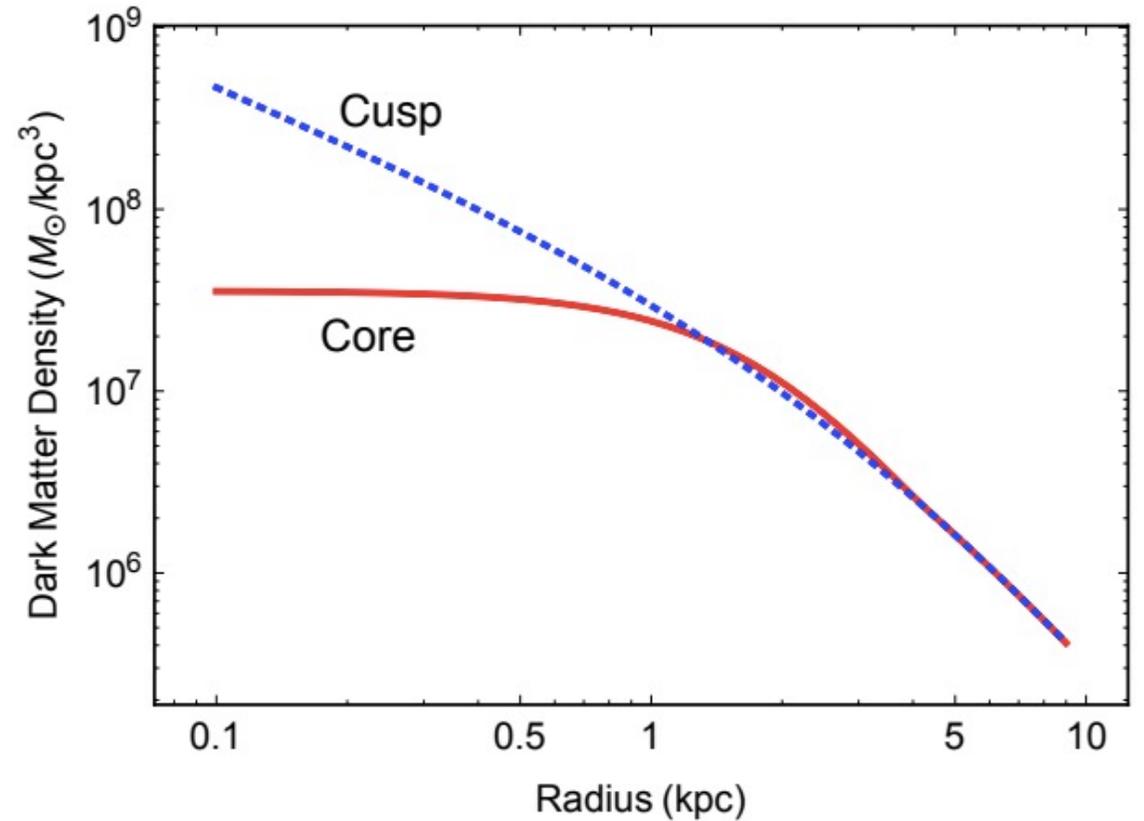
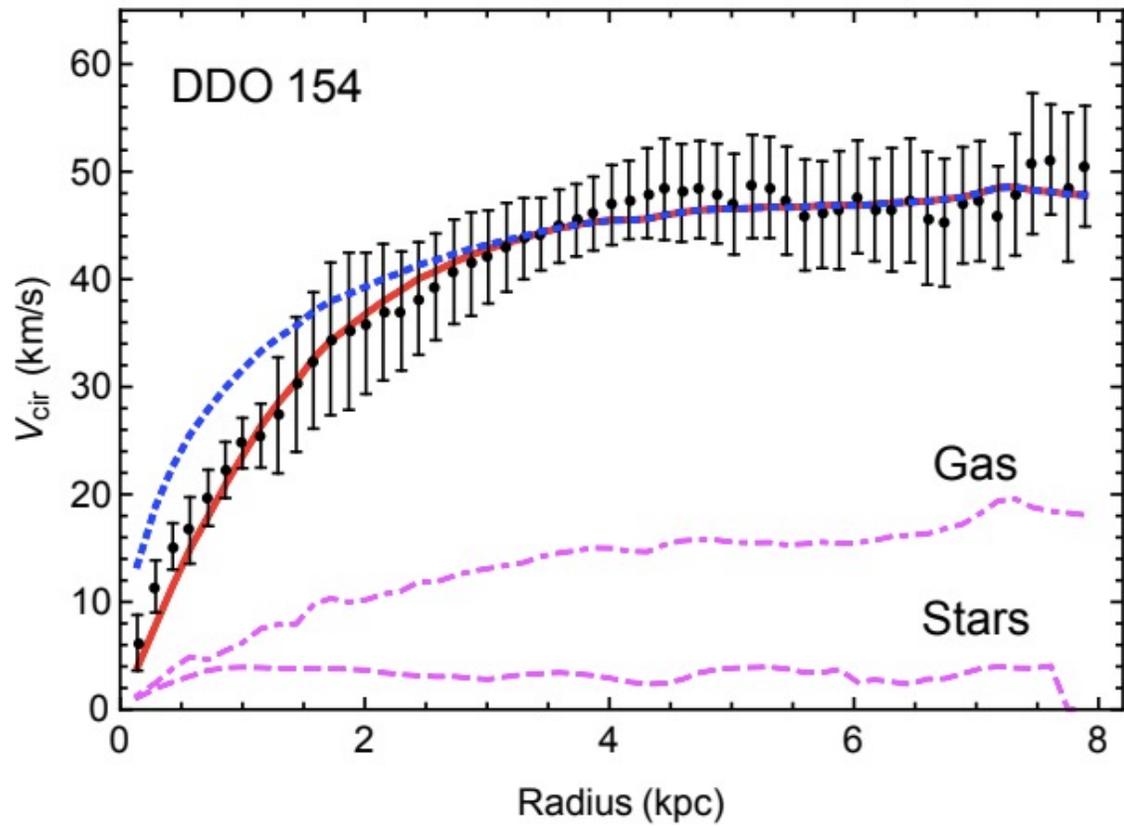
Pawłowski/Bullock/Boylan-Kolchin

Density Profiles of Dwarf Galaxies



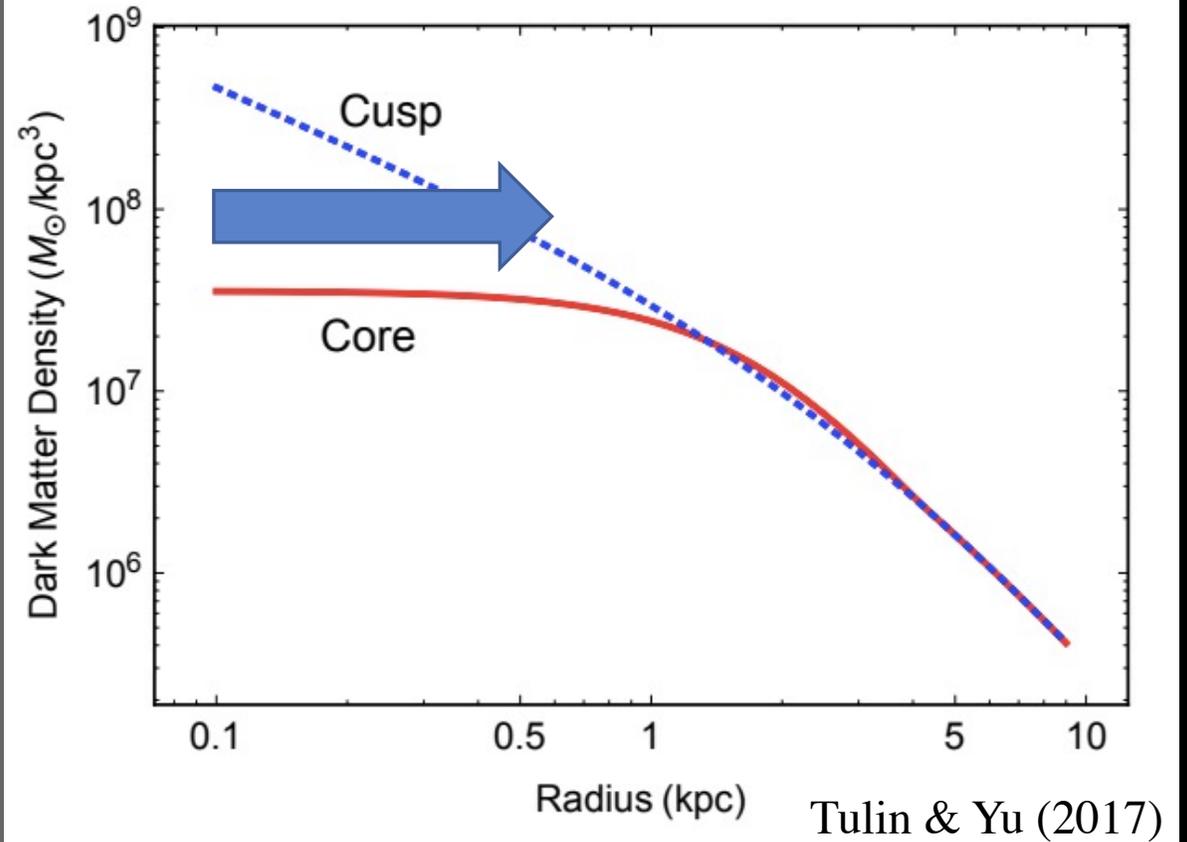
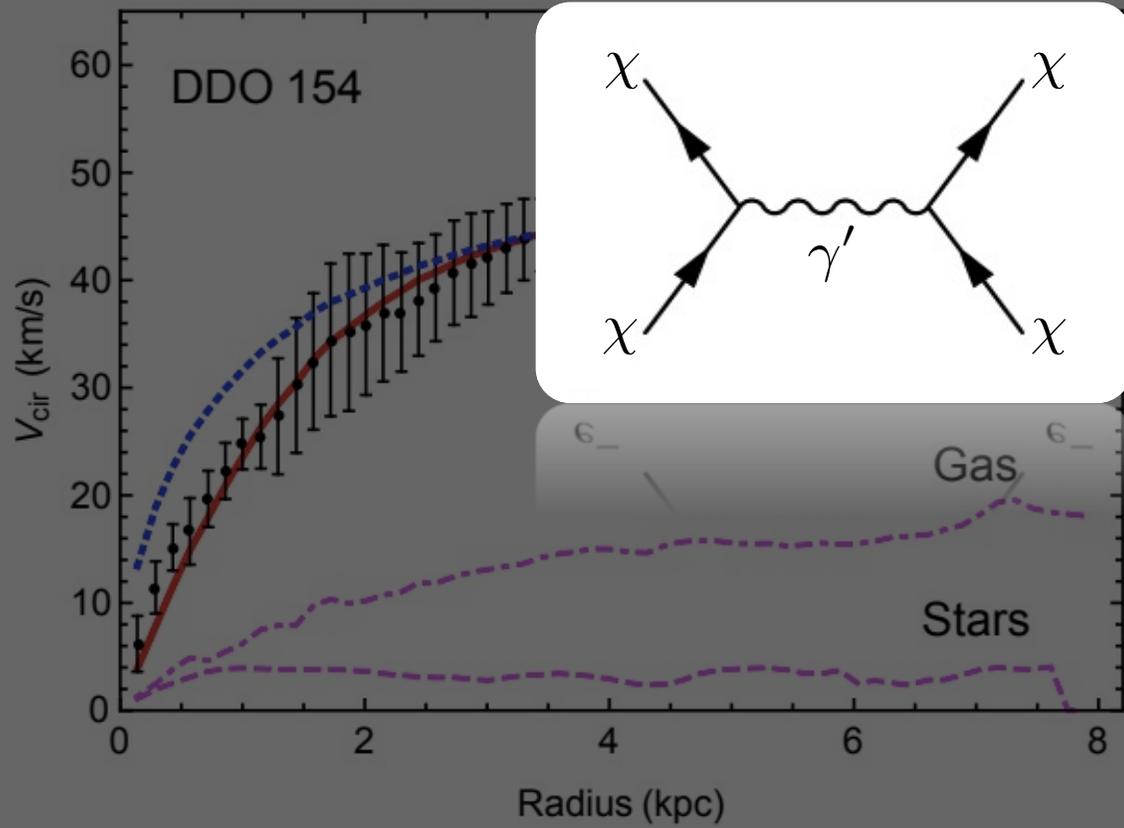
Oh et al. (2015)
Tulin & Yu (2017)

Density Profiles of Dwarf Galaxies



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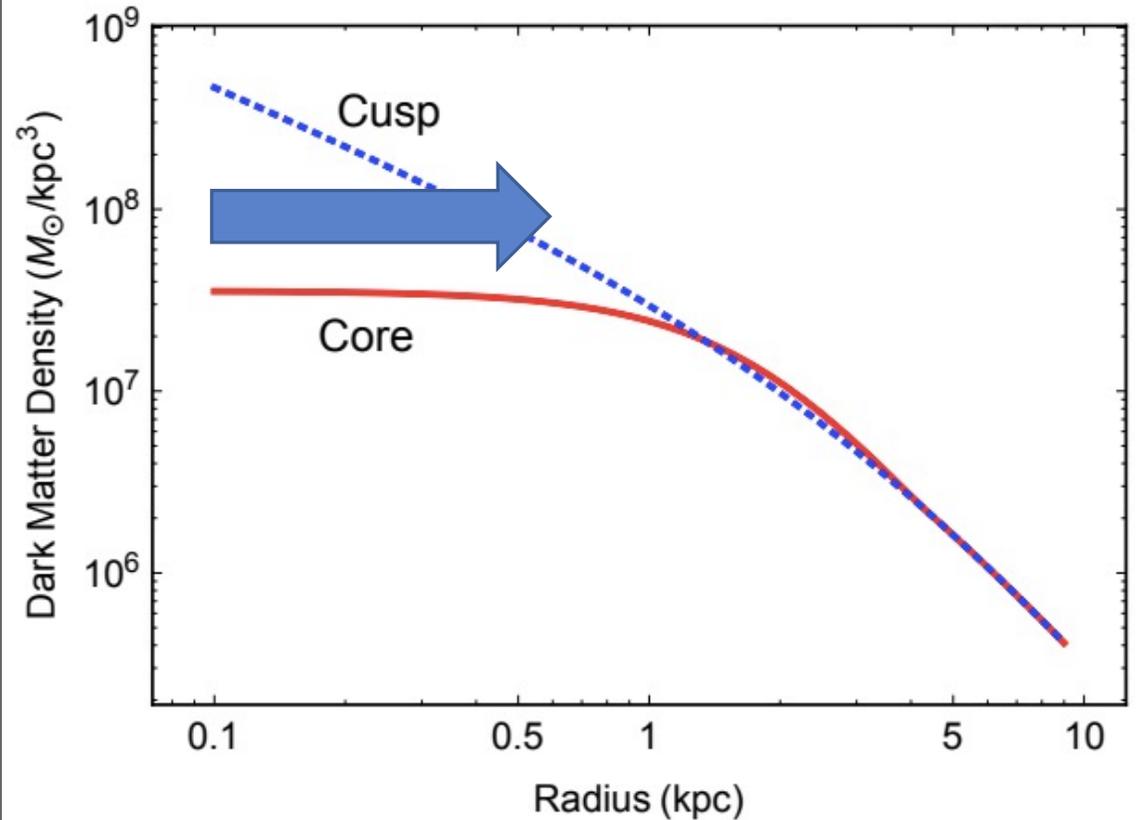
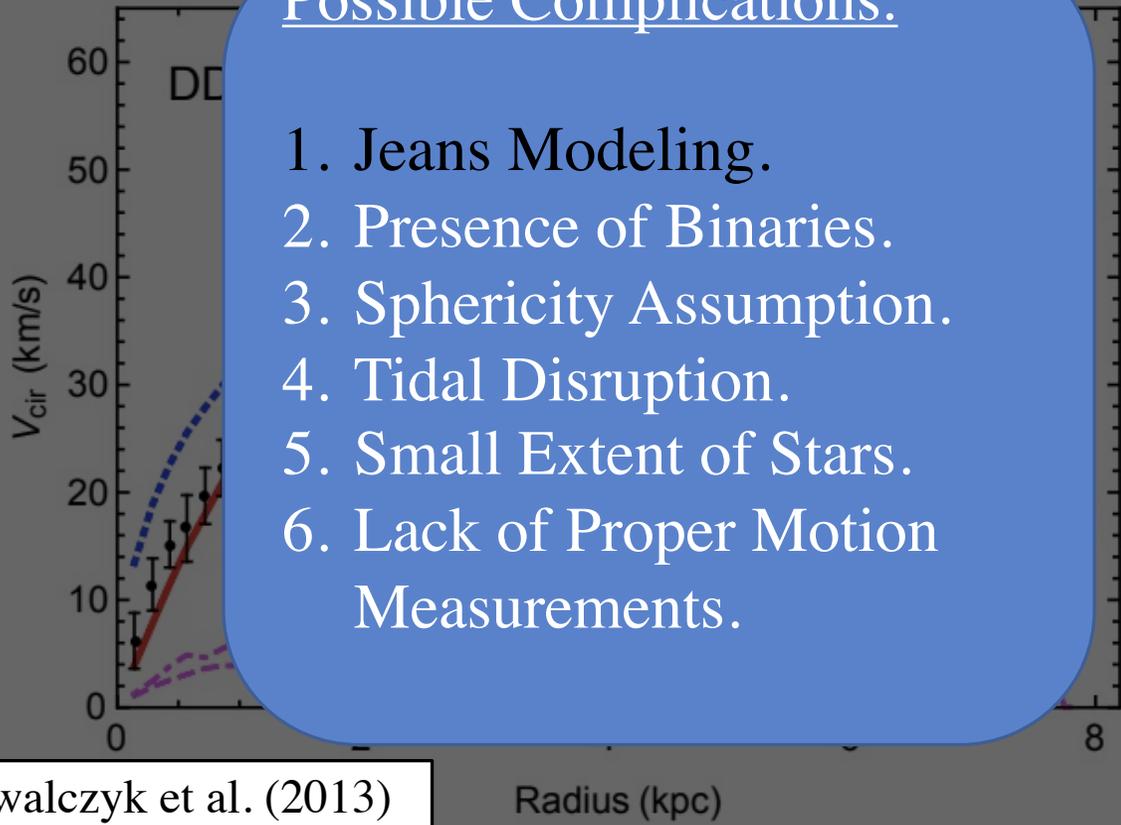


Spergel & Steinhardt (2000)
Oh et al. (2015)

Density Profiles of Dwarf Galaxies

Possible Complications:

1. Jeans Modeling.
2. Presence of Binaries.
3. Sphericity Assumption.
4. Tidal Disruption.
5. Small Extent of Stars.
6. Lack of Proper Motion Measurements.

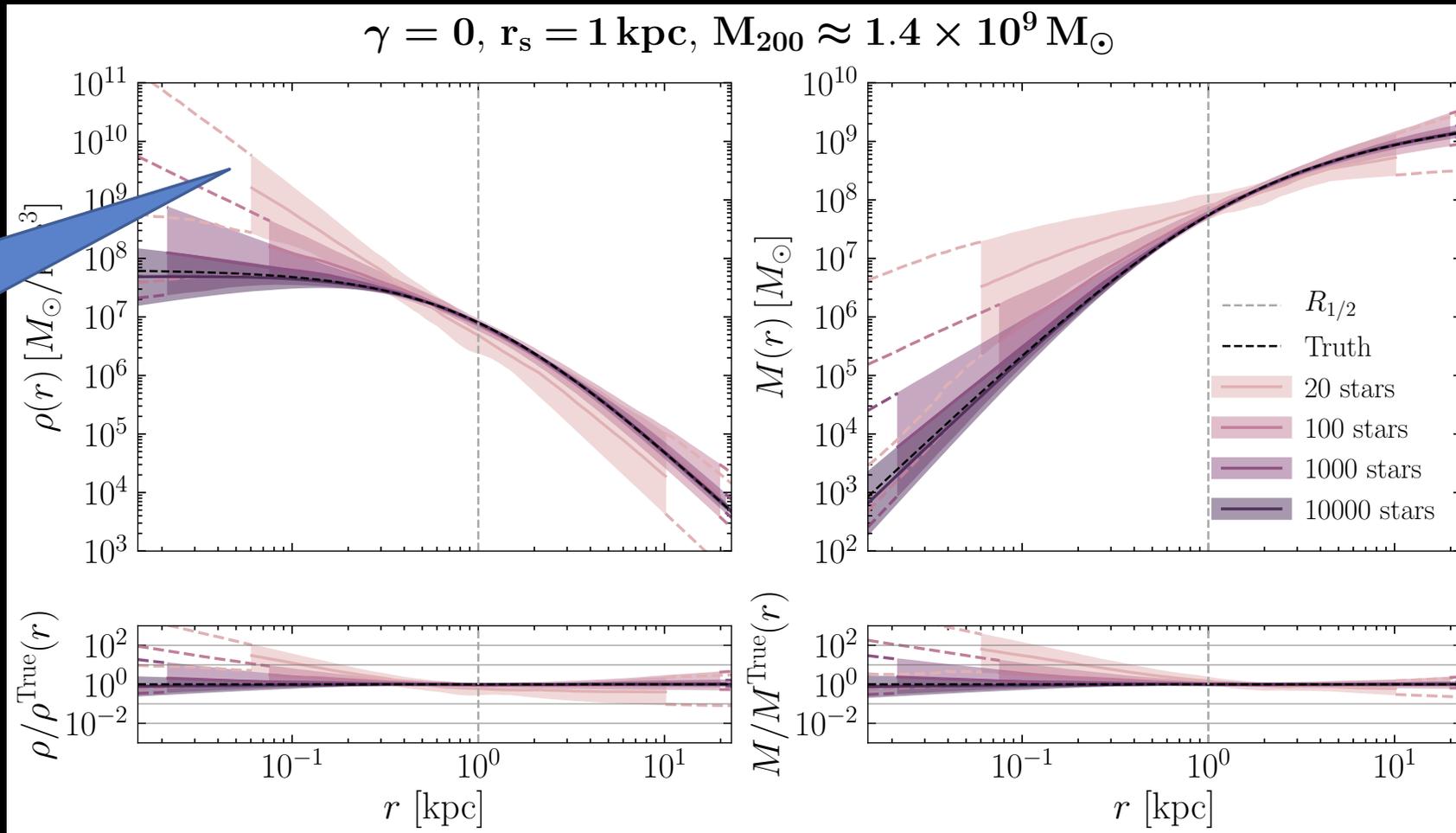


Kowalczyk et al. (2013)
Ackermann et al. (2015)
Bonnivard et al. (2015)
El-Badry et al. (2017)

Oh et al. (2015)
Tulin & Yu (2017)

Density Profiles of Dwarf Galaxies

$\gamma = 0$, $r_s = 1$ kpc, $M_{200} \approx 1.4 \times 10^9 M_\odot$

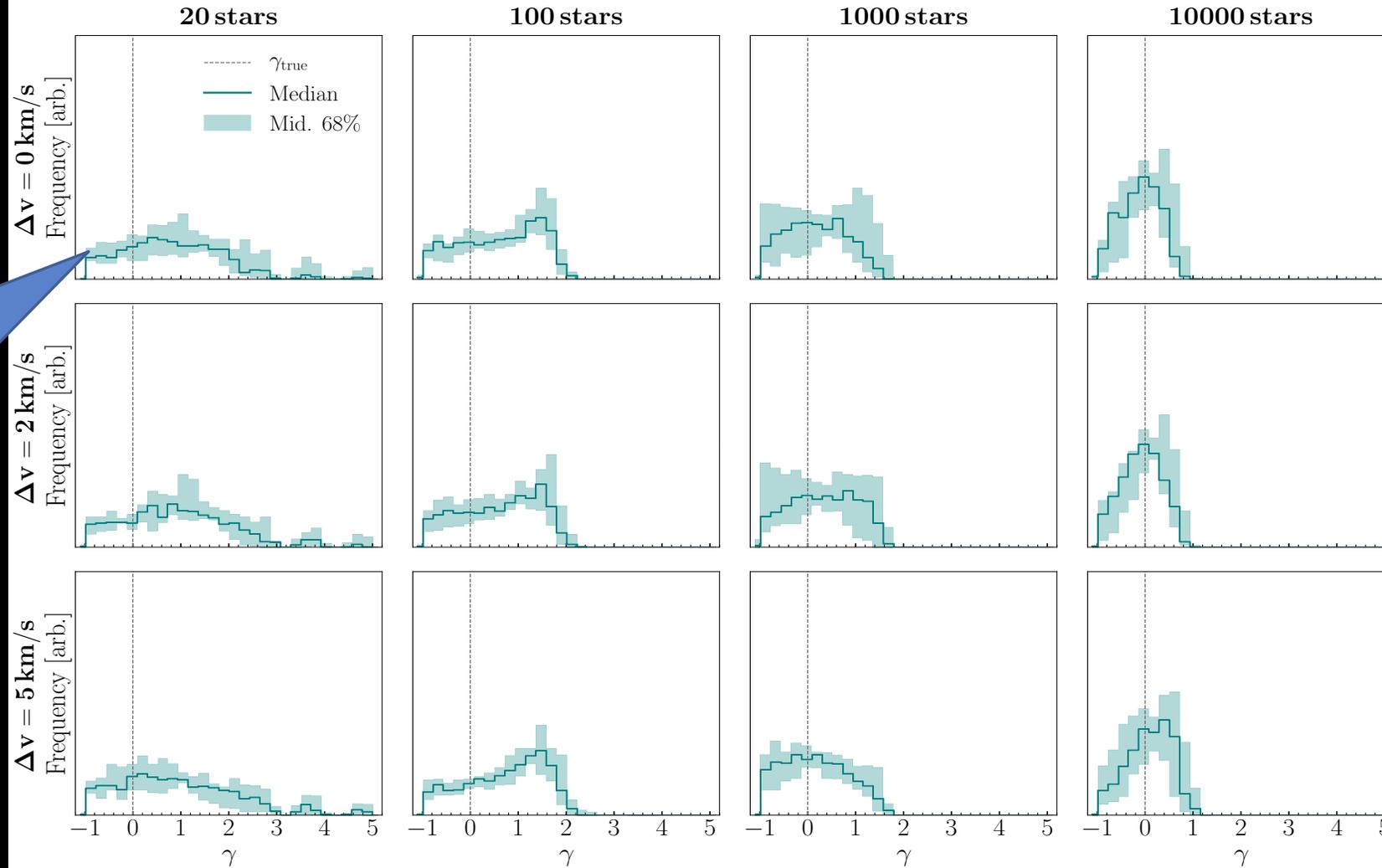


Cores are harder to resolve with fewer statistics

Chang & Necib (2020)

Density Profiles of Dwarf Galaxies

$\gamma = 0$, $r_s = 1$ kpc, $M_{200} \approx 1.4 \times 10^9 M_\odot$



Cores are harder to resolve with fewer statistics

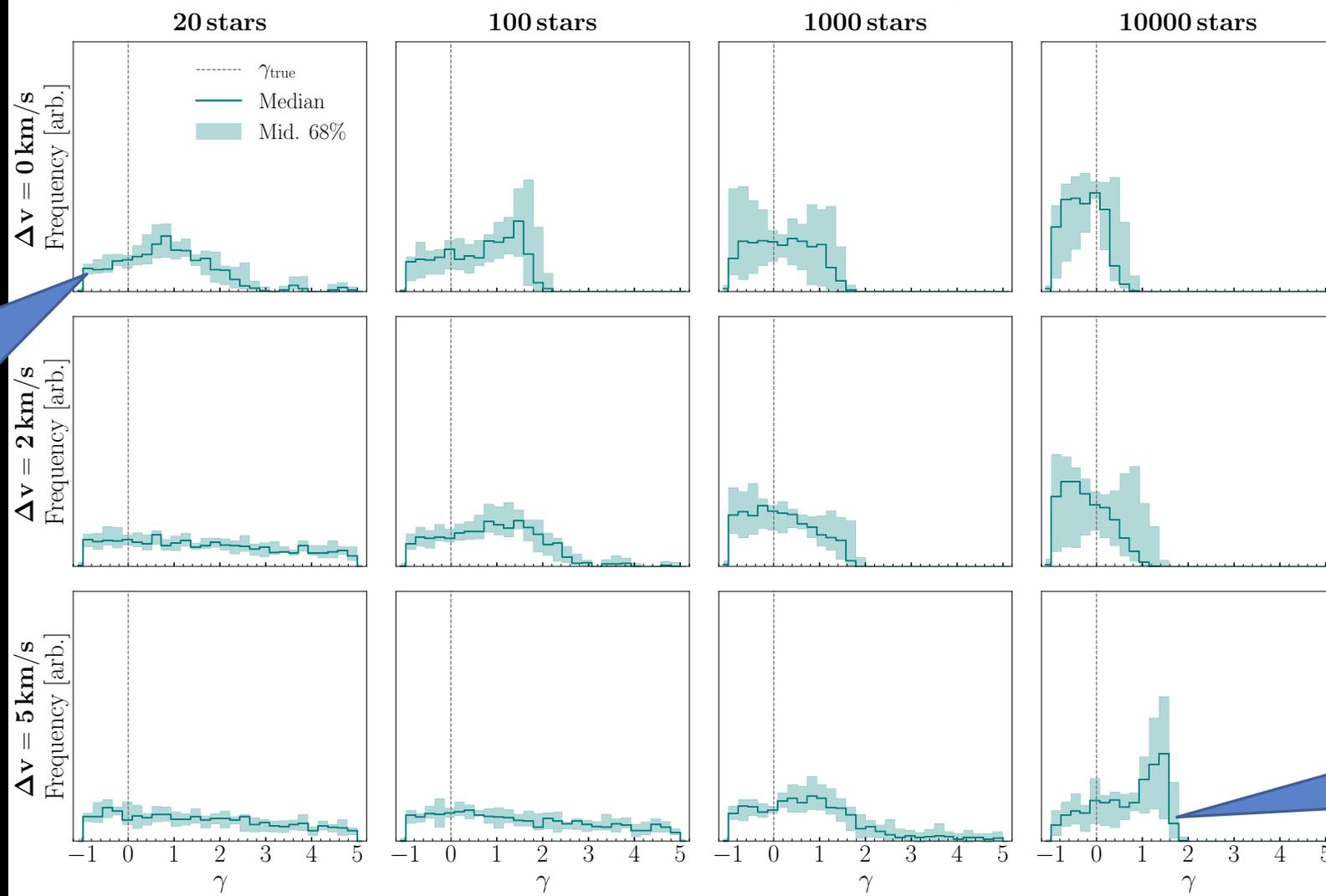
Cusp: $\gamma = 1$
Core: $\gamma = 0$

$$\rho = \rho_0 \left(\frac{r}{r_s} \right)^{-\gamma} \left(1 + \frac{r}{r_s} \right)^{-(3-\gamma)}$$

Chang & Necib (2020)

Density Profiles of Dwarf Galaxies

$\gamma = 0$, $r_s = 0.2 \text{ kpc}$, $M_{200} \approx 1.1 \times 10^7 M_\odot$



Cores are harder to resolve with fewer statistics

Cusp: $\gamma = 1$
Core: $\gamma = 0$

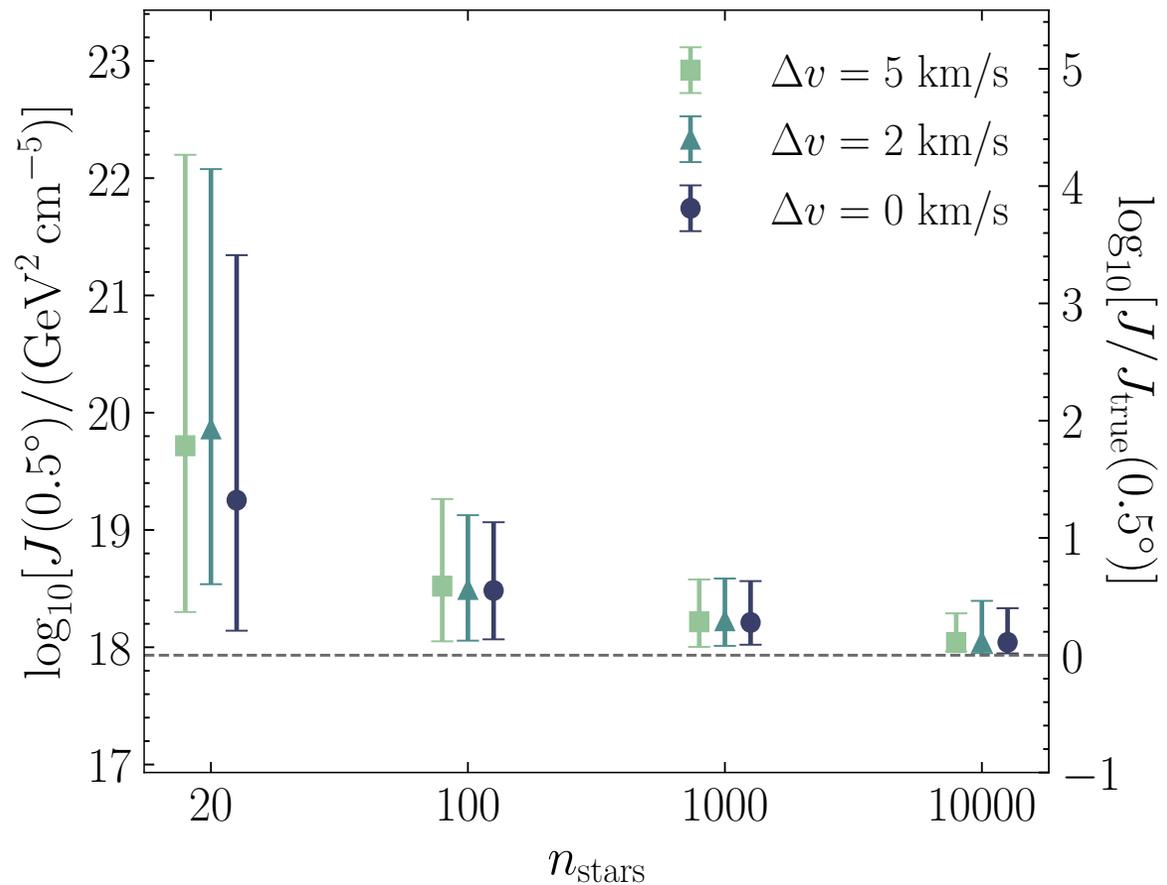
The errors are worse for the ultrafaints

$$\rho = \rho_0 \left(\frac{r}{r_s} \right)^{-\gamma} \left(1 + \frac{r}{r_s} \right)^{-(3-\gamma)}$$

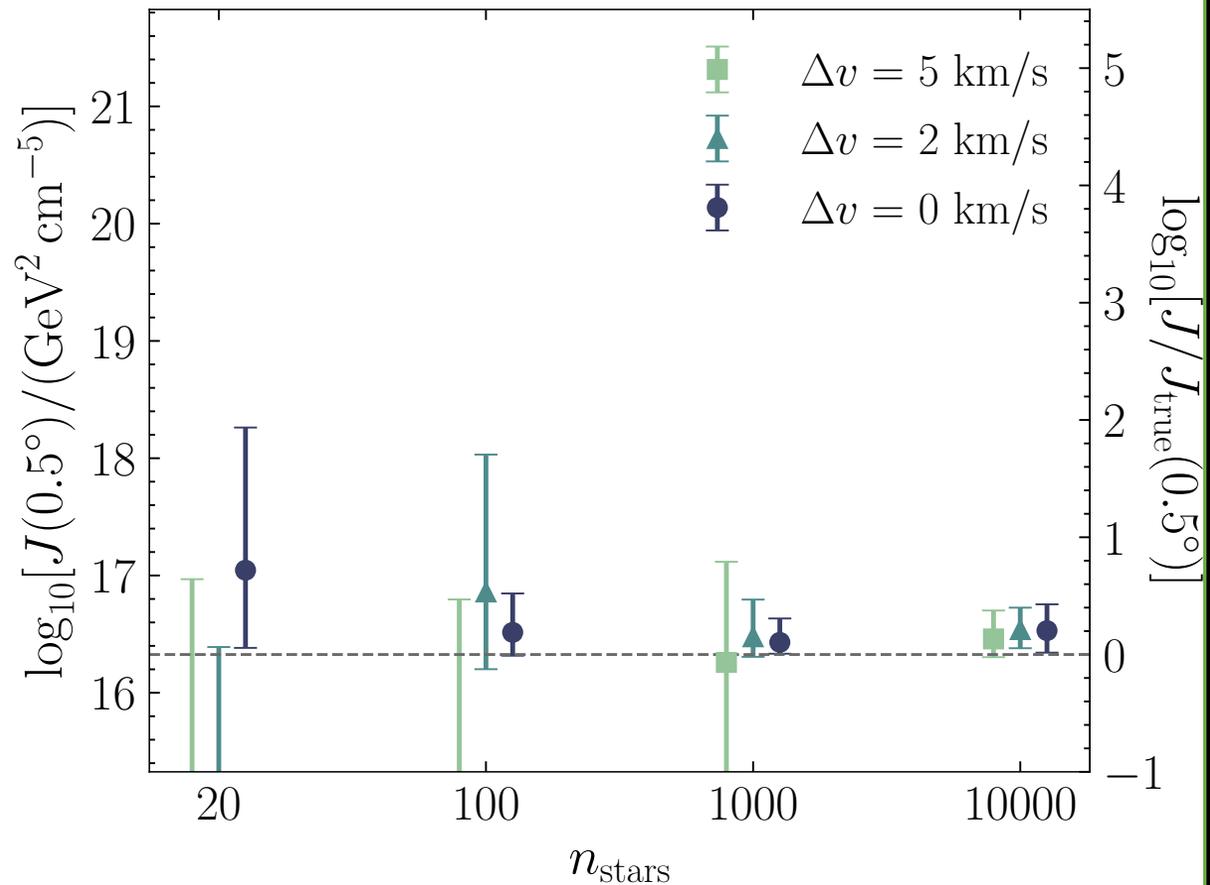
Chang & **Necib** (2020)

J factors of Dwarf Galaxies

$\gamma = 0, r_s = 1 \text{ kpc}, M_{200} \approx 1.4 \times 10^9 M_\odot$



$\gamma = 0, r_s = 0.2 \text{ kpc}, M_{200} \approx 1.1 \times 10^7 M_\odot$



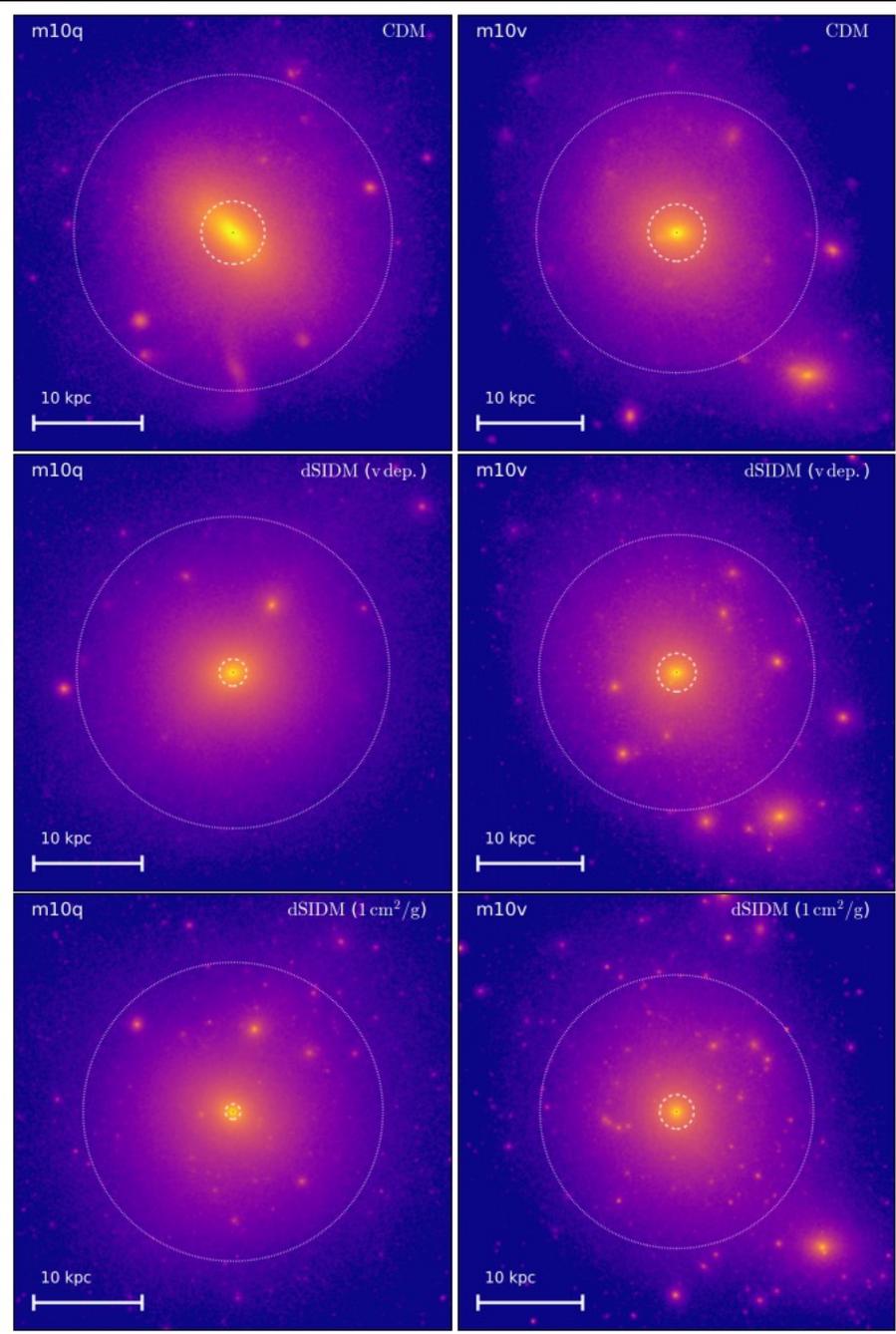
$$J = \int ds \int d\Omega \rho^2$$

Dwarf	N Stars	$\log_{10} J(0.5^\circ)$ [GeV ² cm ⁻⁵]	Dispersion [km/s]	References
Ursa Major II	20	19.42 ^{+0.44} _{-0.42}	5.6 ^{+1.4} _{-1.4}	Simon (2019)
Segue 1	70	19.36 ^{+0.32} _{-0.35}	3.7 ^{+1.4} _{-1.1}	Simon & Geha (2007)
Coma Berenices	59	19.02 ^{+0.37} _{-0.41}	4.6 ^{+0.8} _{-0.8}	Simon & Geha (2007)
Ursa Minor	313	18.93 ^{+0.27} _{-0.19}	9.5 ^{+1.2} _{1.2}	Walker et al. (2009b)
Draco	292	18.84 ^{+0.12} _{-0.13}	9.1 ^{+1.2} _{-1.2}	Walker et al. (2009b)
Sculptor	1365	18.54 ^{+0.06} _{-0.05}	9.2 ^{+1.1} _{-1.1}	Walker et al. (2009a)
Bootes I	37	18.24 ^{+0.40} _{-0.37}	4.6 ^{+0.8} _{-0.6}	Koposov et al. (2011)
Leo II	126	17.97 ^{+0.20} _{-0.18}	7.4 ^{+0.4} _{-0.4}	Spencer et al. (2017)
Carina	774	17.87 ^{+0.10} _{-0.09}	6.6 ^{+1.2} _{-1.2}	Walker et al. (2009a)
Ursa Major I	39	17.87 ^{+0.56} _{-0.33}	7.0 ^{+1.0} _{-1.0}	Simon (2019)
Leo I	267	17.84 ^{+0.20} _{-0.16}	9.2 ^{+0.4} _{-0.4}	Mateo et al. (2008)
Fornax	2483	17.83 ^{+0.12} _{-0.06}	11.7 ^{+0.9} _{-0.9}	Walker et al. (2009a)
Canes Venatici II	25	17.65 ^{+0.45} _{-0.43}	4.6 ^{+1.0} _{-1.0}	Simon & Geha (2007)
Sextans	441	17.52 ^{+0.28} _{-0.18}	7.9 ^{+1.3} _{-1.3}	Walker et al. (2009a)
Canes Venatici I	214	17.43 ^{+0.37} _{-0.28}	7.6 ^{+0.4} _{-0.4}	Simon & Geha (2007)
Leo T	19	17.11 ^{+0.44} _{-0.39}	7.5 ^{+1.6} _{-1.6}	Simon & Geha (2007)
Hercules	30	16.86 ^{+0.74} _{-0.68}	5.1 ^{+0.2} _{-0.2}	Simon & Geha (2007)
Leo V	5	16.37 ^{+0.94} _{-0.87}	2.3 ^{+3.2} _{-1.6}	Collins et al. (2017)
Leo IV	18	16.32 ^{+1.06} _{-1.69}	3.3 ^{+1.7} _{-1.7}	Simon & Geha (2007)
Segue 2	25	16.21 ^{+1.06} _{-0.98}	< 2.2	Kirby et al. (2013)

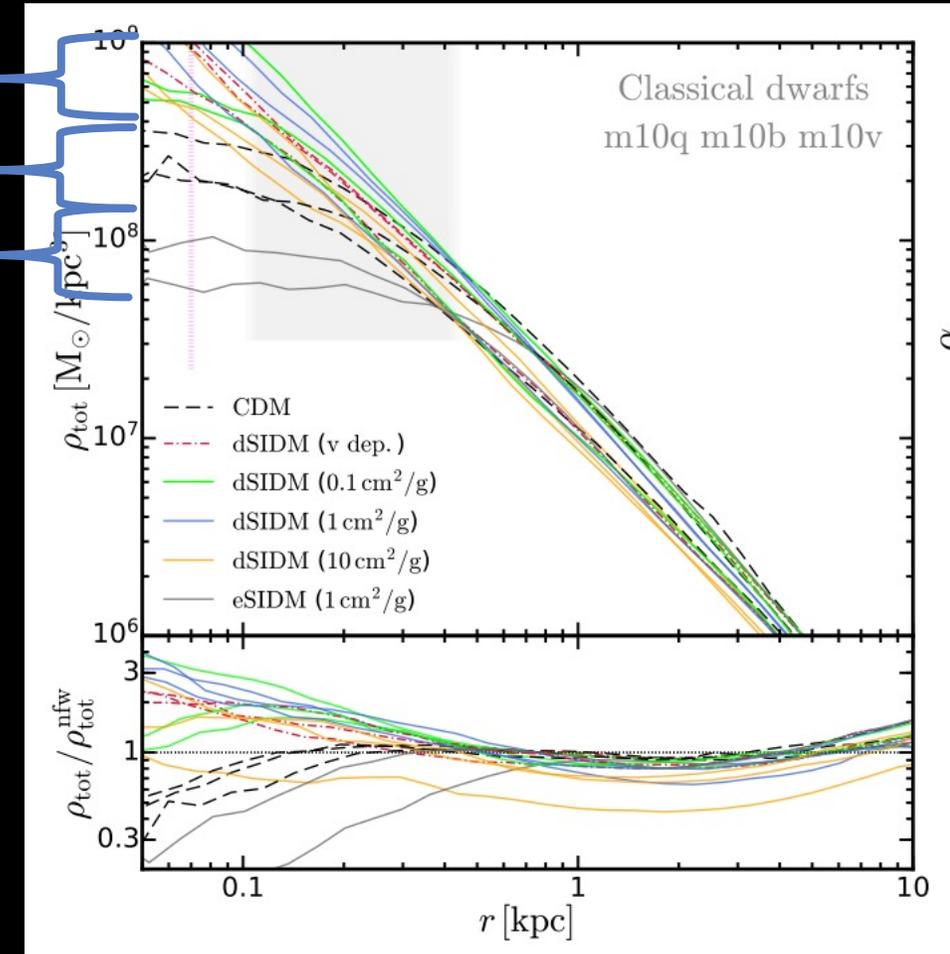
Dwarf	N Stars	$\log_{10} J(0.5^\circ)$ [GeV ² cm ⁻⁵]	Dispersion [km/s]	References
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Once we have the Density Profiles of Galaxies under
Control

Beyond Cold Dark Matter



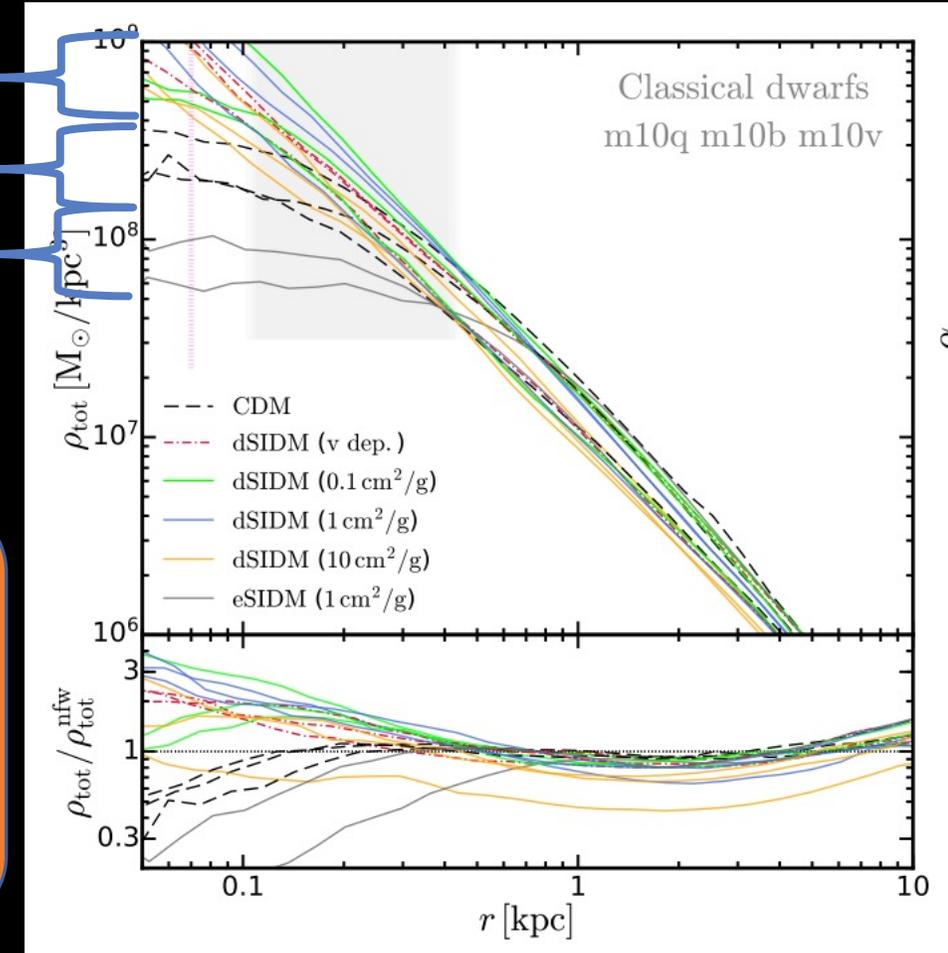
Dissipative Dark Matter
Cold Dark Matter
Self Interacting Dark Matter



Beyond Cold Dark Matter

Dissipative Dark Matter
 Cold Dark Matter
 Self Interacting Dark Matter

We need to explore transferring Particle Physics Dark Matter Models into Hydrodynamics large scale simulations



Streams



Galactic
Center

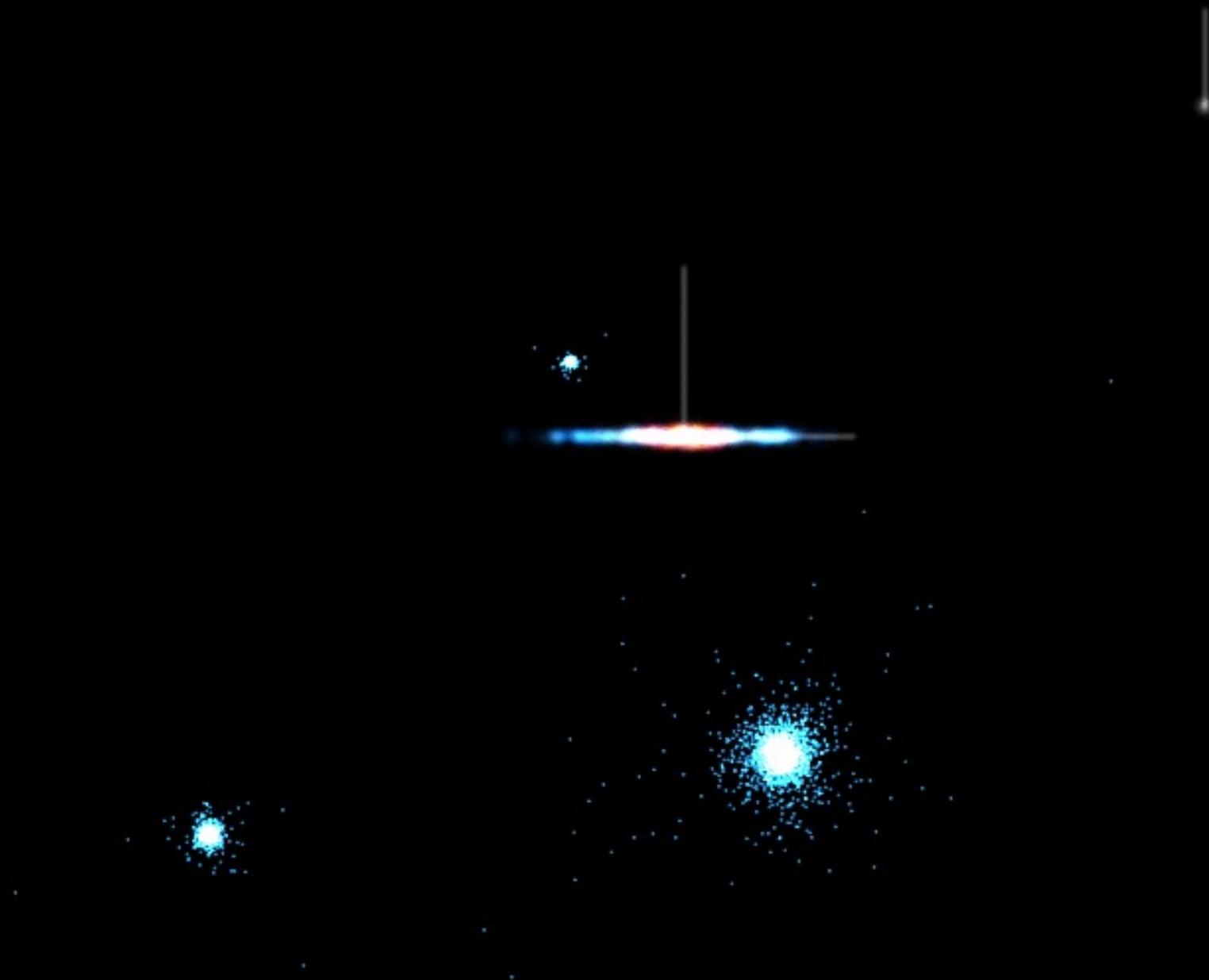


Solar
Neighborhood

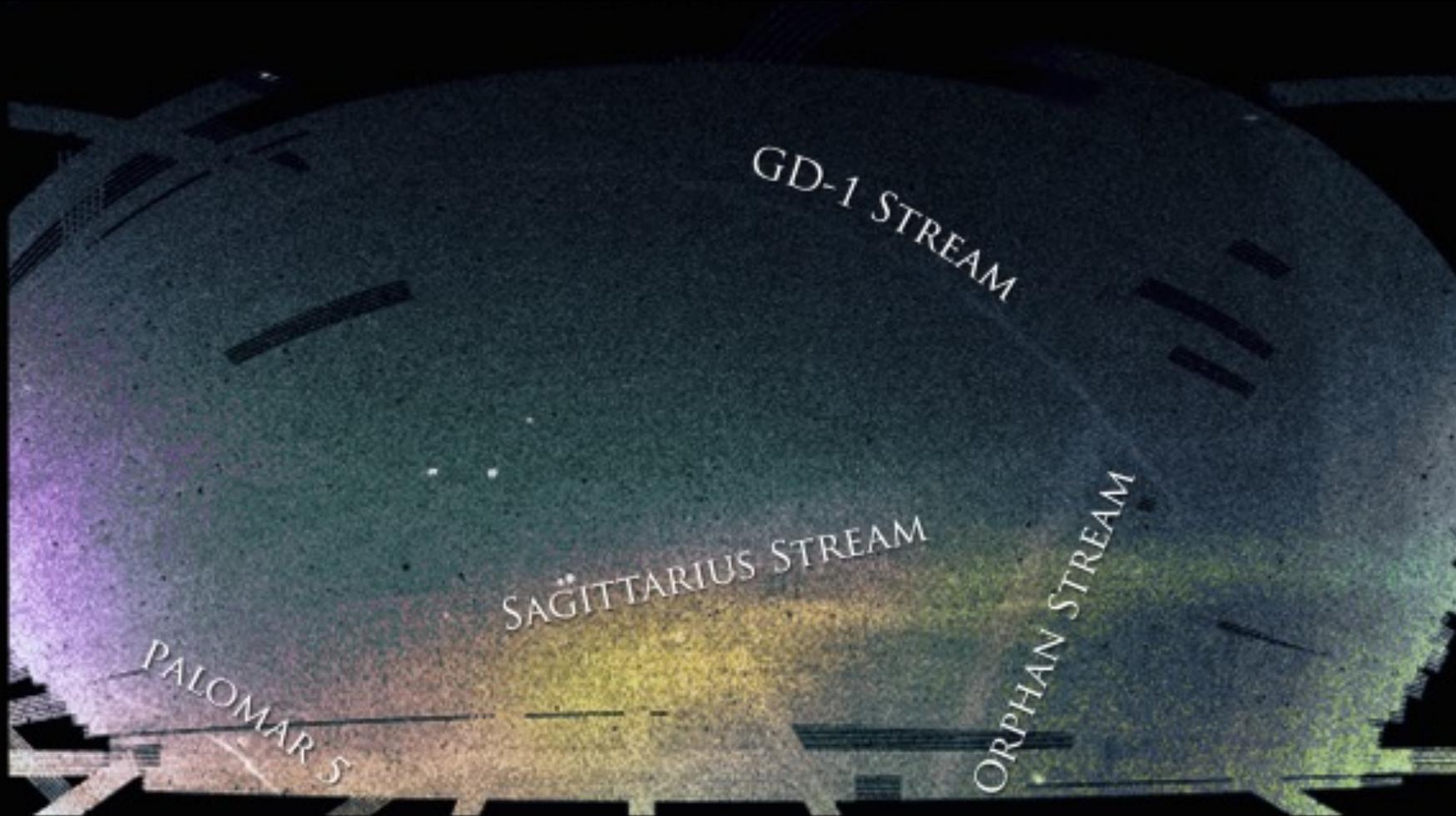


Dwarf
Galaxies





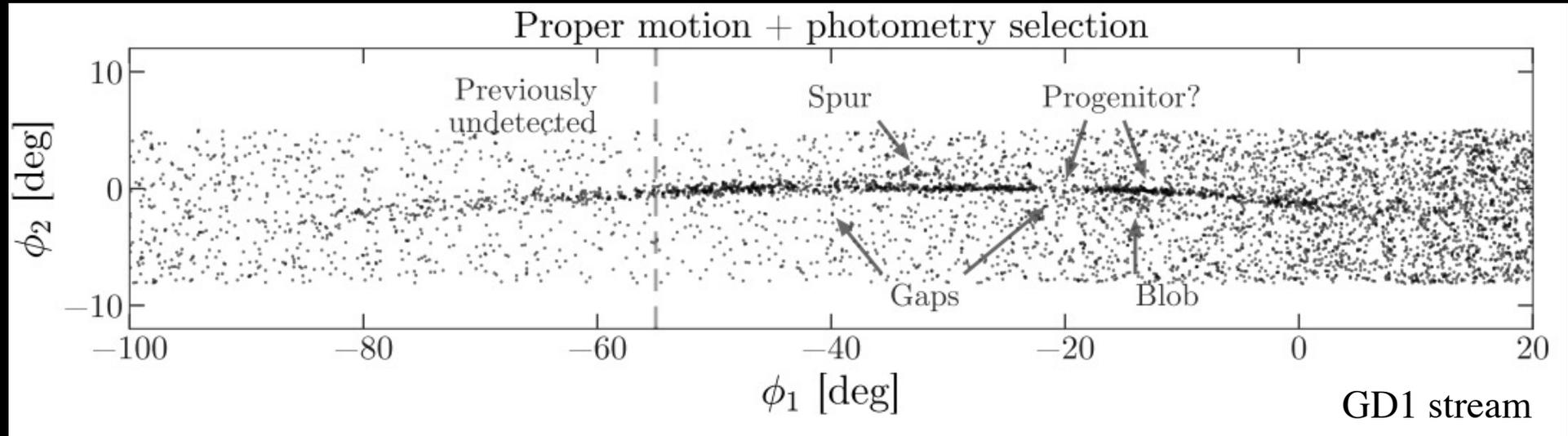
Field of Streams



Credit: Bonaca, Geha, Kallivayalil

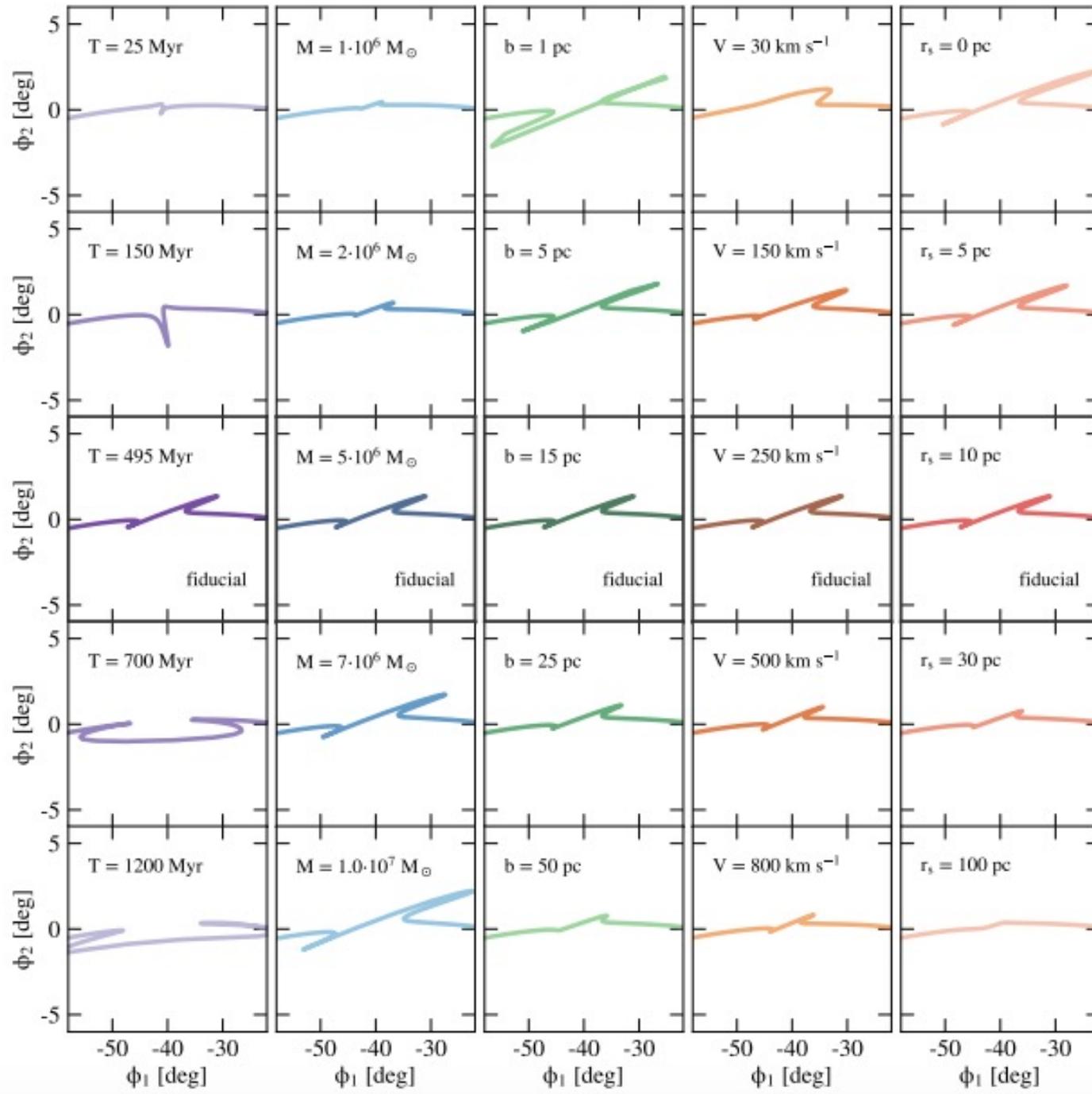
Data: Sloan Digital Sky Survey

Streams: An insight into Dark Matter

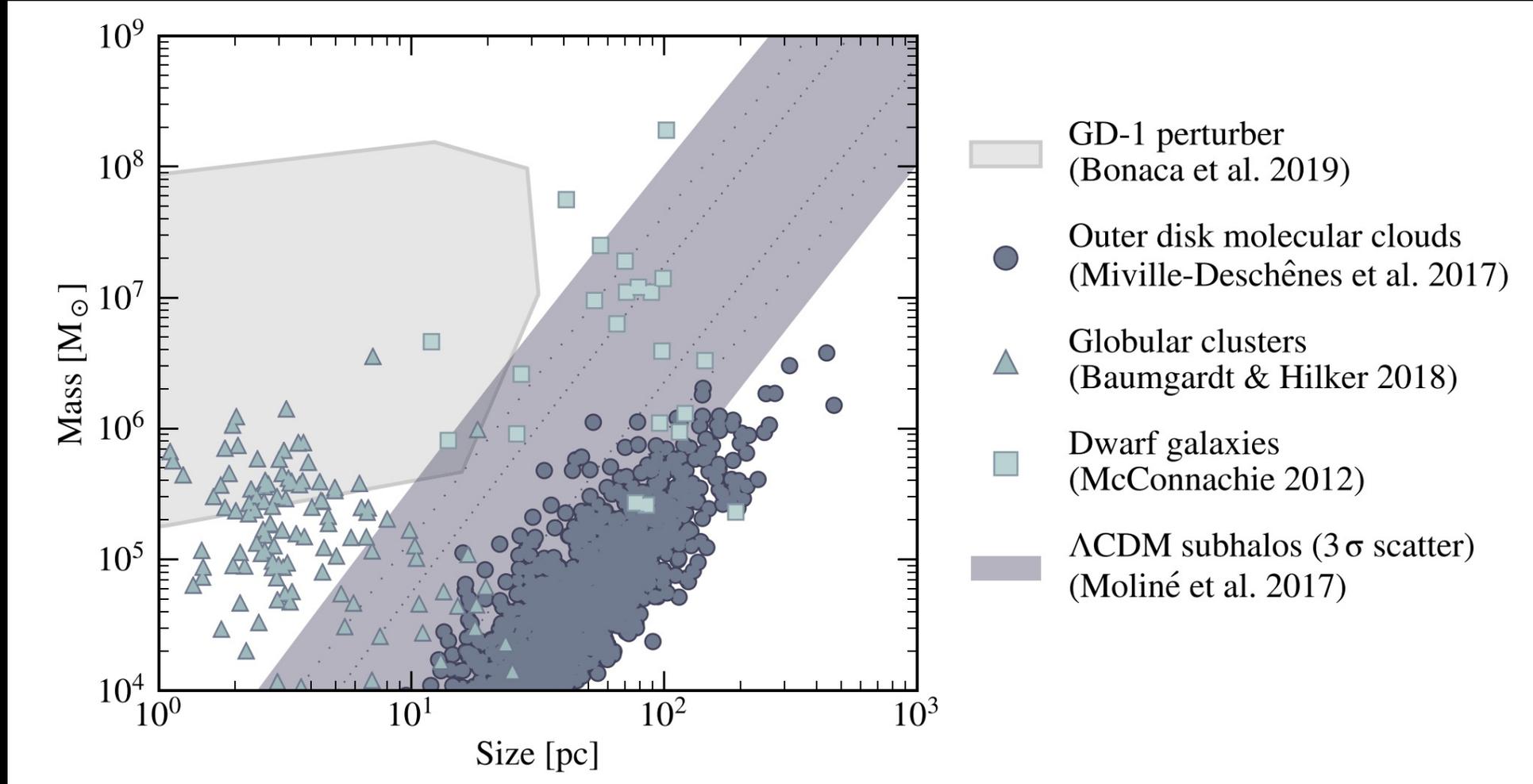


- Gaps in streams can constrain dark matter subhalo masses, and therefore models of warm dark matter!
- Streams are also used to constrain the potential of the Milky Way.

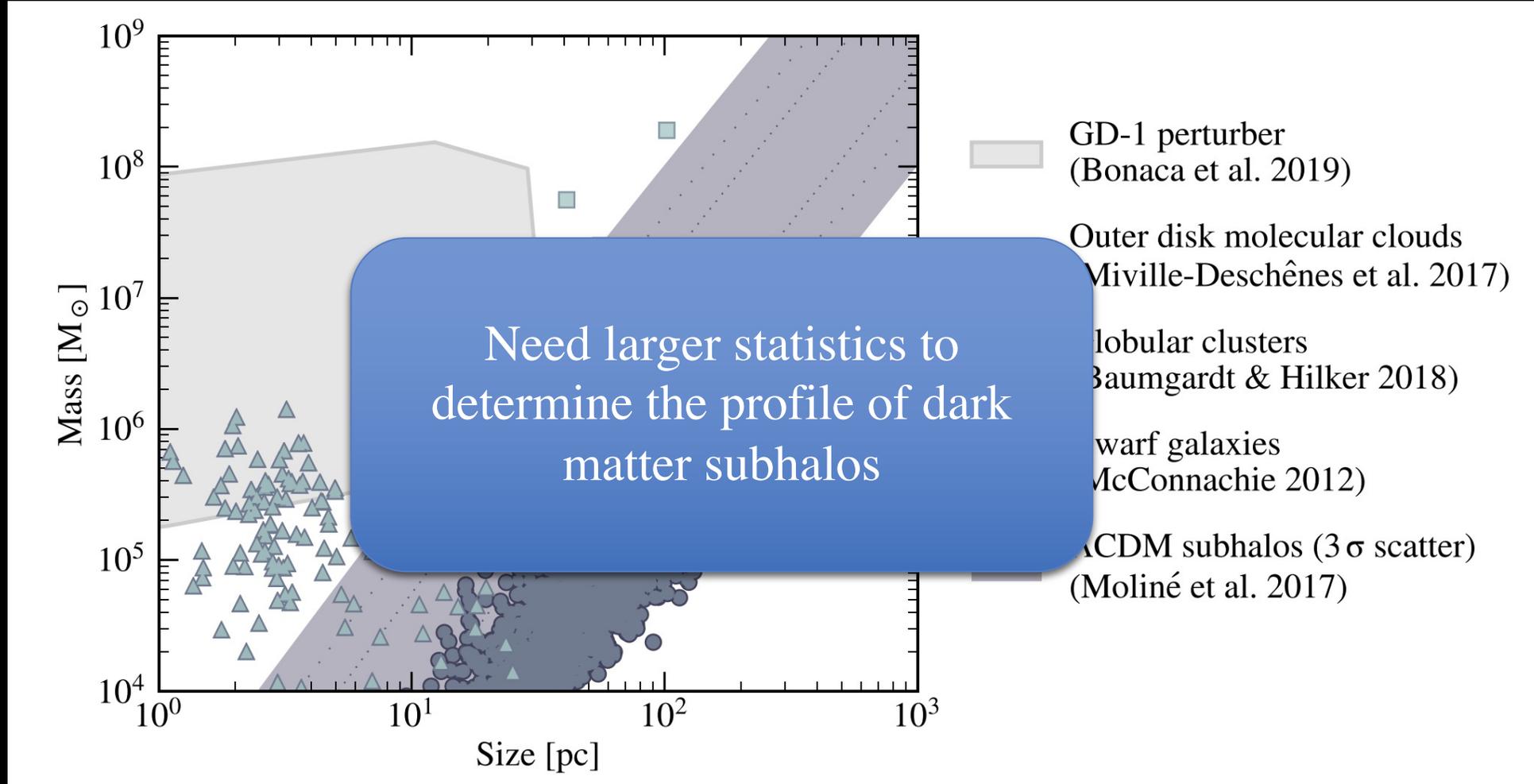
Grillmair & Dionatos (2006b)
Koposov et al. (2010)
Price-Whelan & Bonaca (2018)
Bonaca et al. (2019)



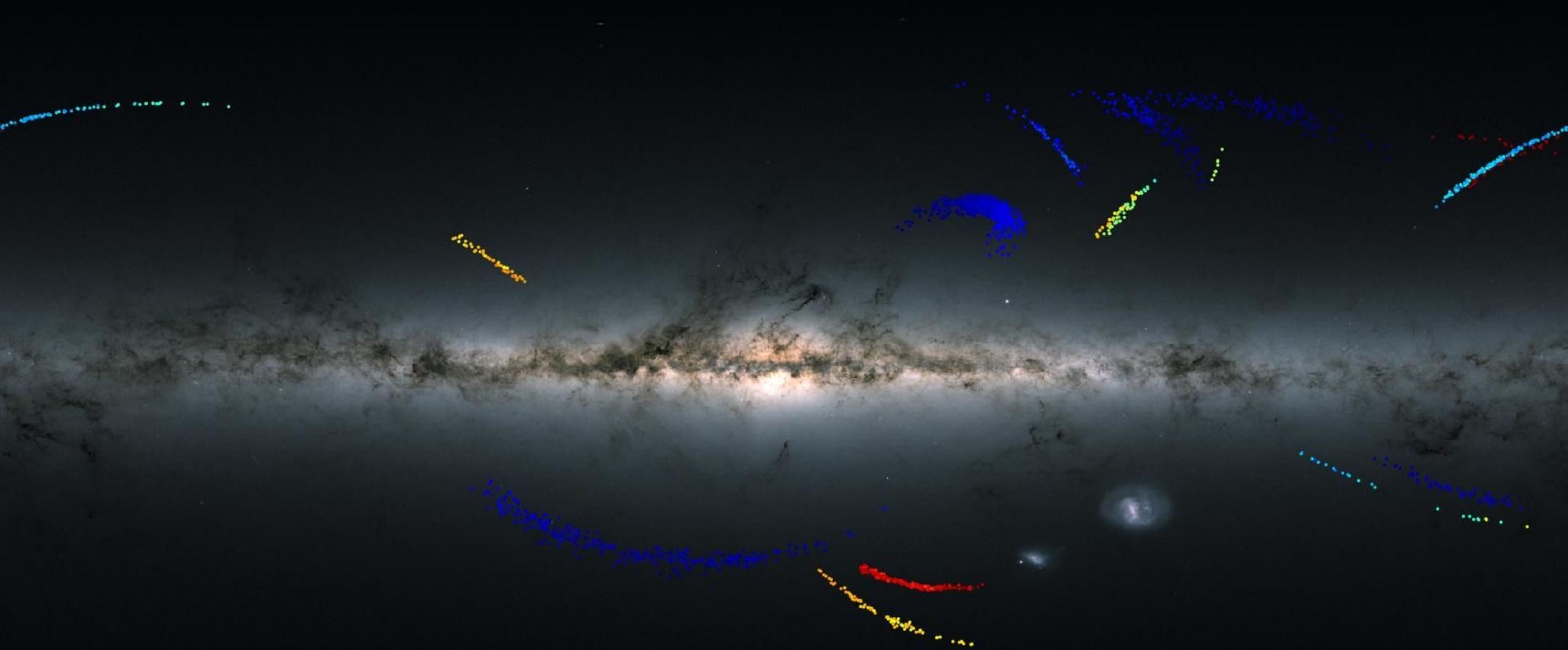
GD-1 Perturber?



GD-1 Perturber?



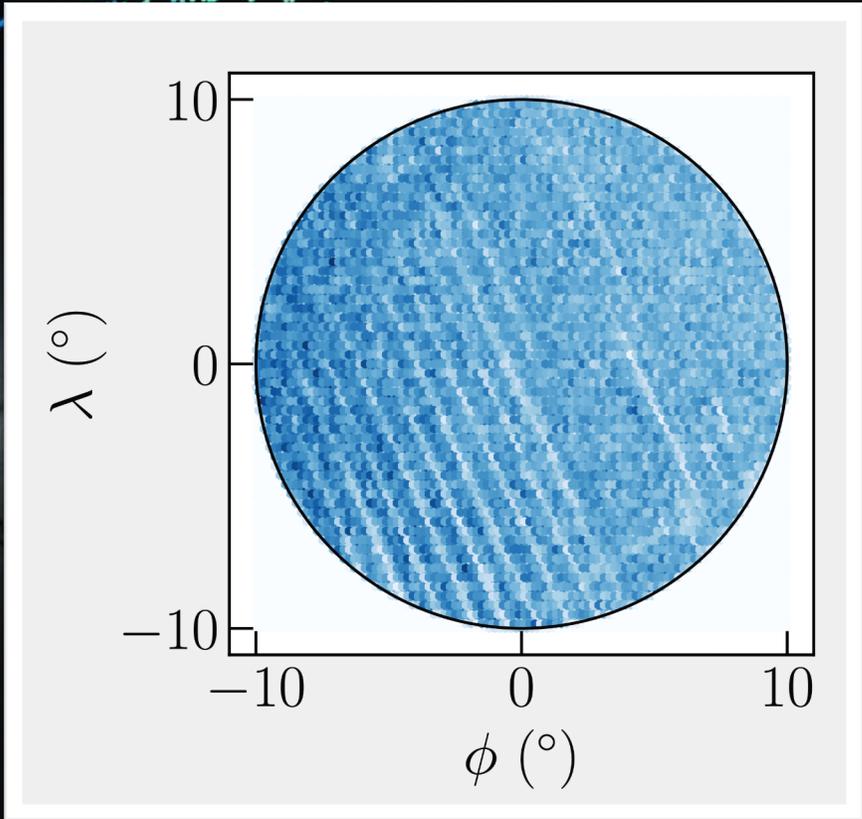
Need to increase statistics of streams



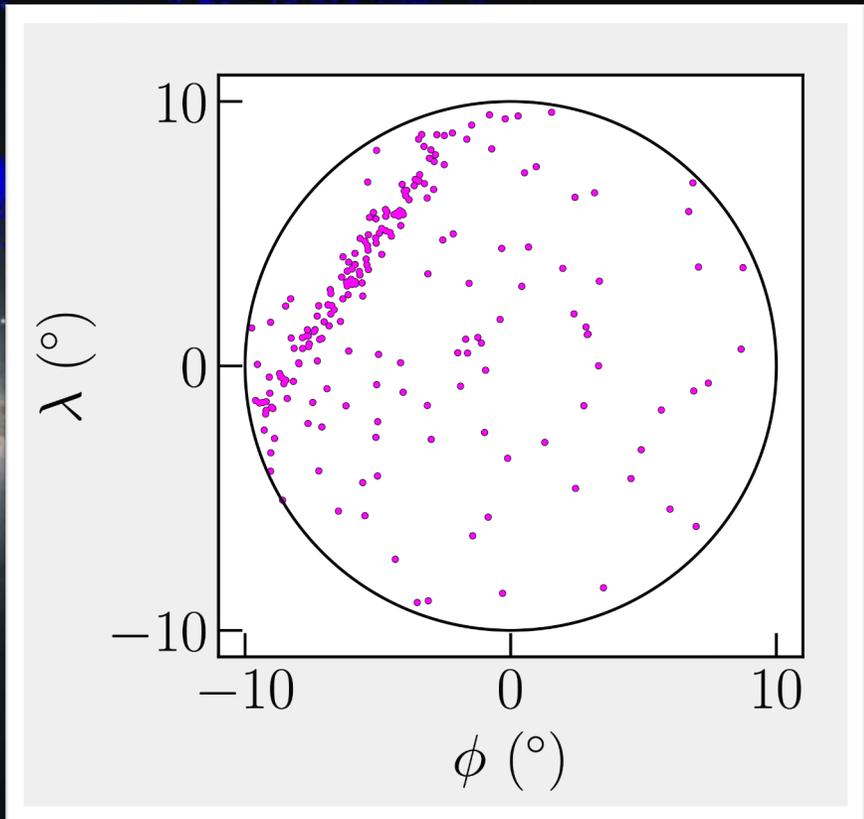
For a large dataset of streams, see <https://github.com/cmateu/galstreams>

Galaxy Picture
Credit : ESA/Gaia/DPAC
Stellar Streams
Malhan et al. (2018), Ibata et al. (2019)

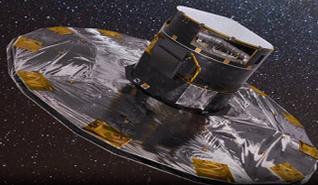
Need to increase statistics of streams



Via Machinae
A machine learning algorithm that searches for anomalies in a multi-dimensional distribution.



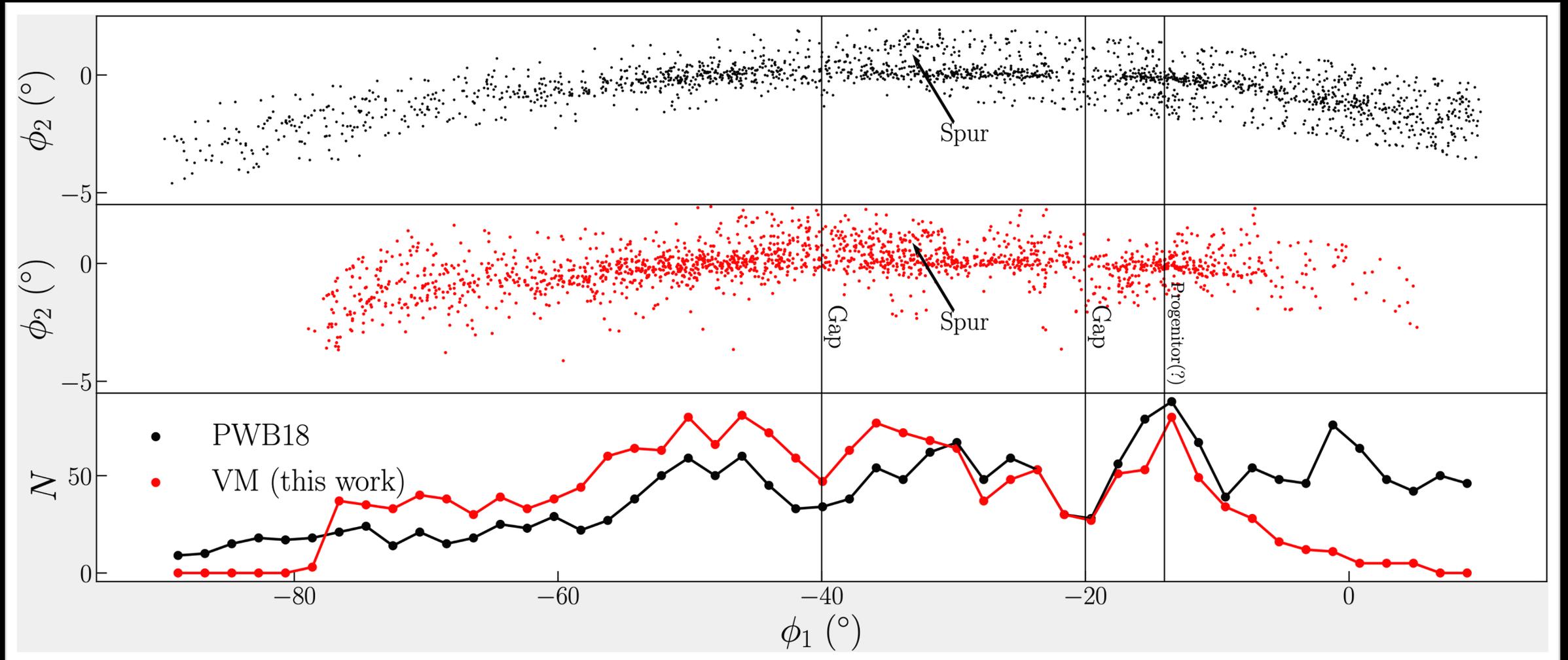
Gaia



Galaxy Picture
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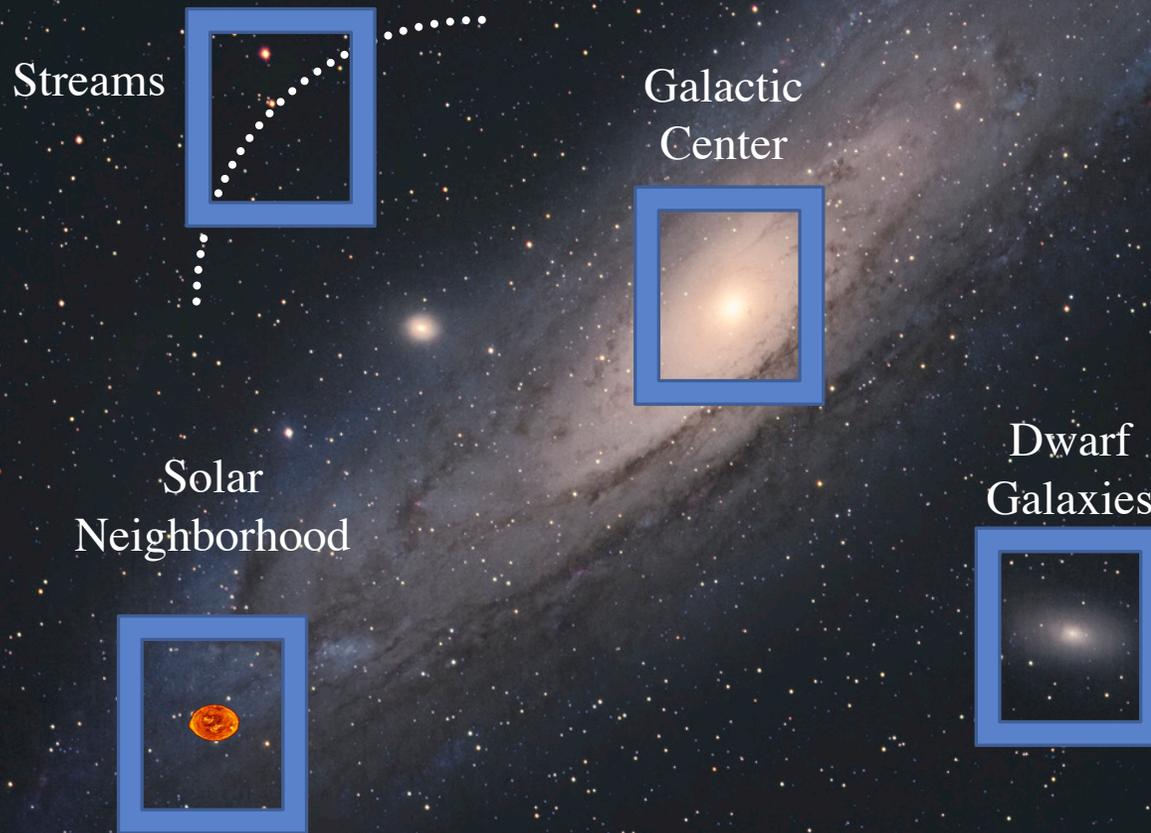
Shih, Buckley, Necib, Tamasas (2021)

Apply Via Machinae on GD-1



This Talk:

How to map out the Dark Matter phase space distribution on Galactic Scales at key locations.





Thank you!