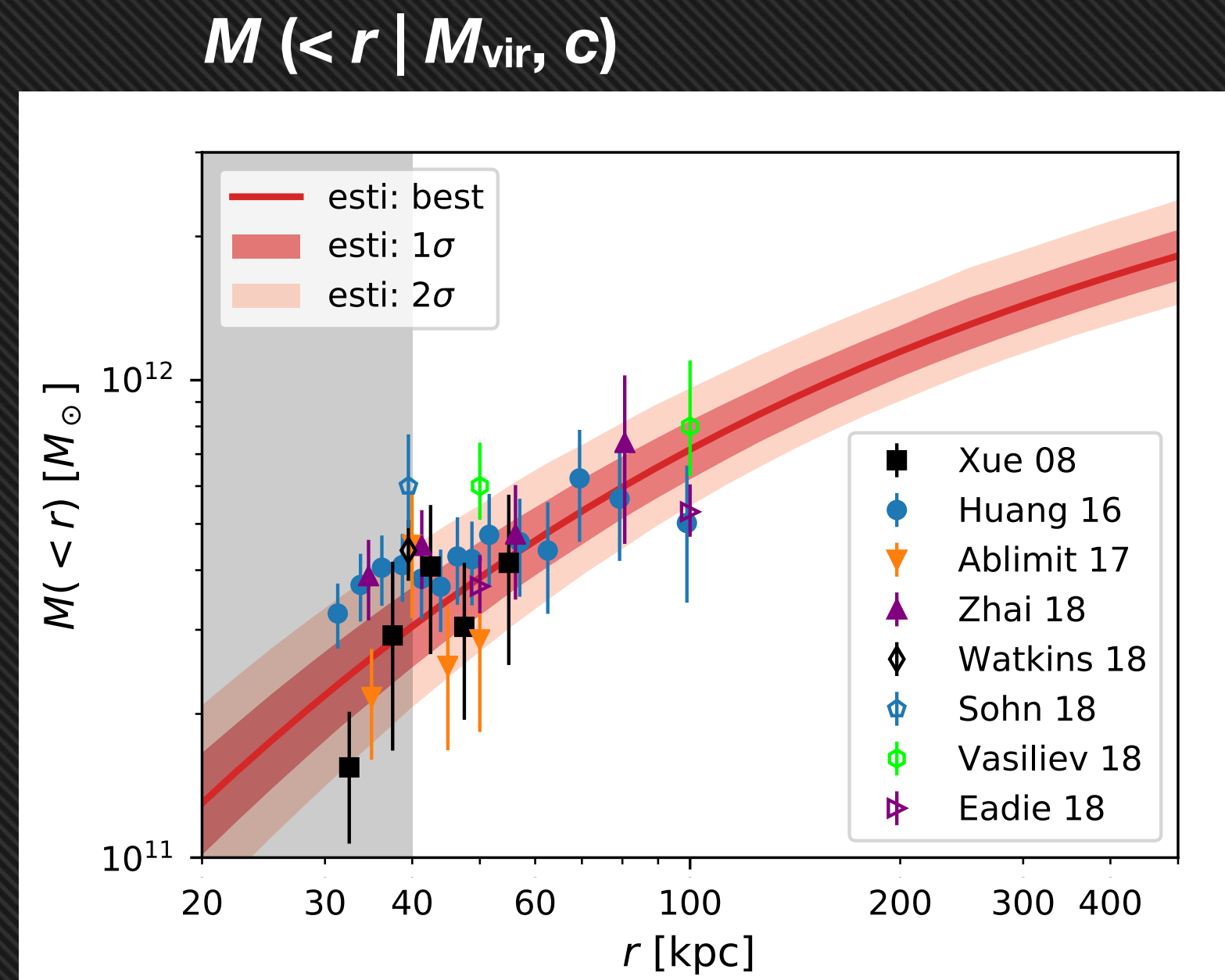




Constrain the Milky Way Mass Profile with Phase Space Distribution of Satellite Galaxies



Zhaozhou Li (李昭洲)

Shanghai Jiao Tong University

Collaborators:

Jiaxin Han (SJTU)

Wenting Wang (SJTU)

Ting Li (Carnegie)

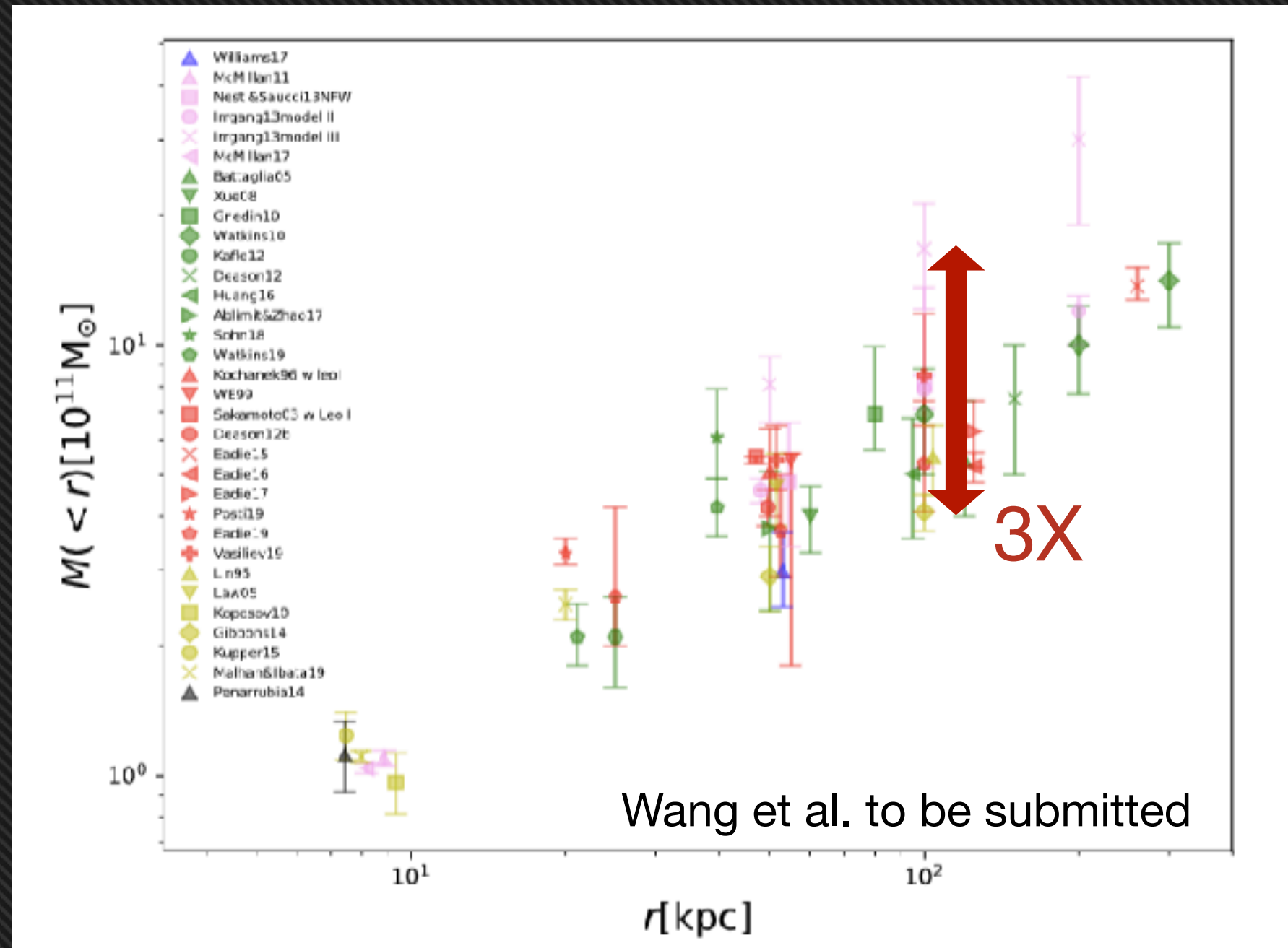
Yongzhong Qian (UMN/TDLi)

Yipeng Jing (SJTU)

Why we care about Milky Way mass & profile

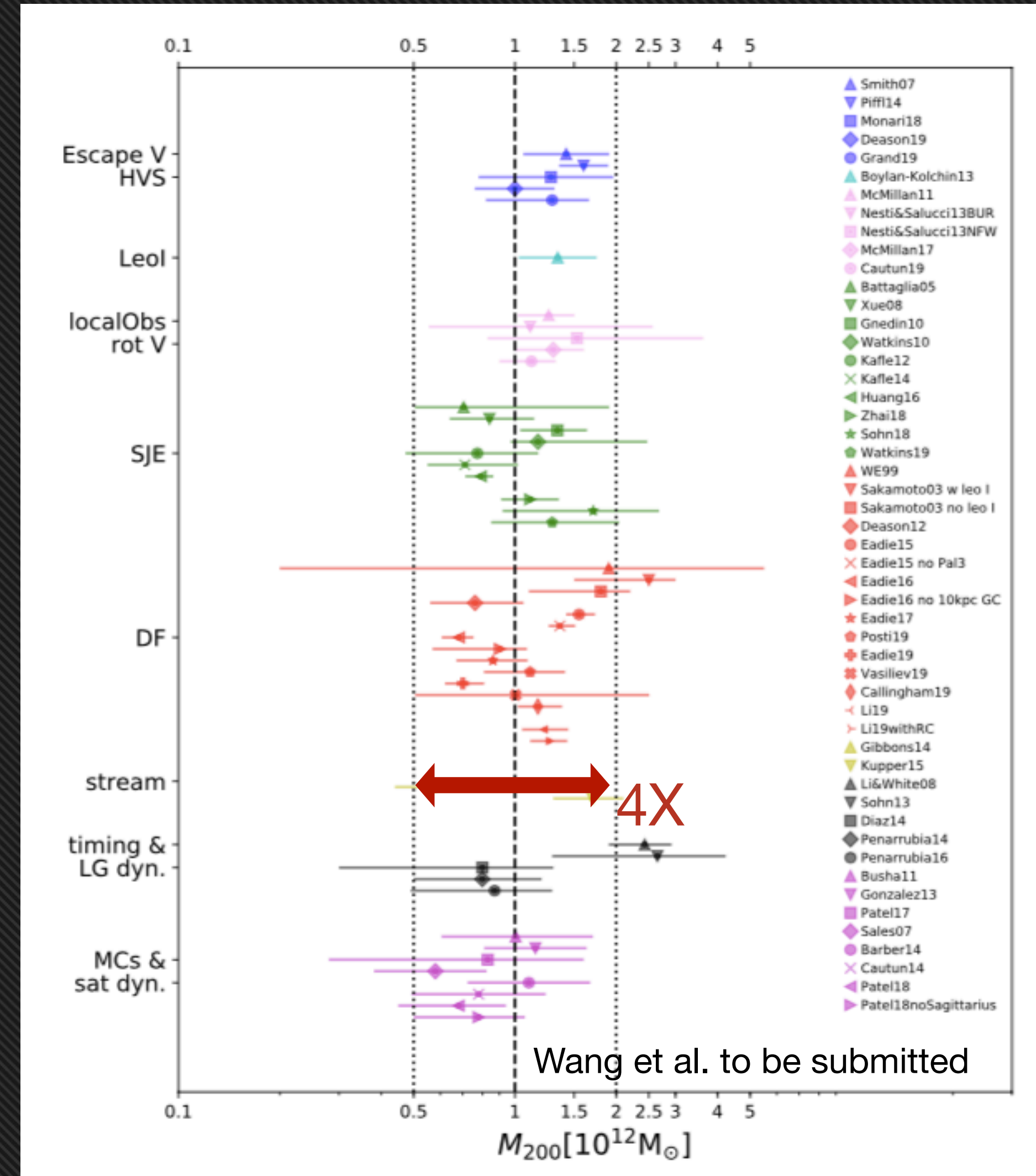
- ▶ Basic properties of our home galaxy
 - ▶ Galactic dynamics
 - ▶ Formation and assembly history of the MW
- ▶ Basis for validating theory predictions from local observation
 - ▶ Dark matter detection
 - ▶ Cosmology, e.g. expected baryon fraction, number of satellites
 - ▶ Gravity Theory

Milky Way (MW) Mass: still very uncertain!

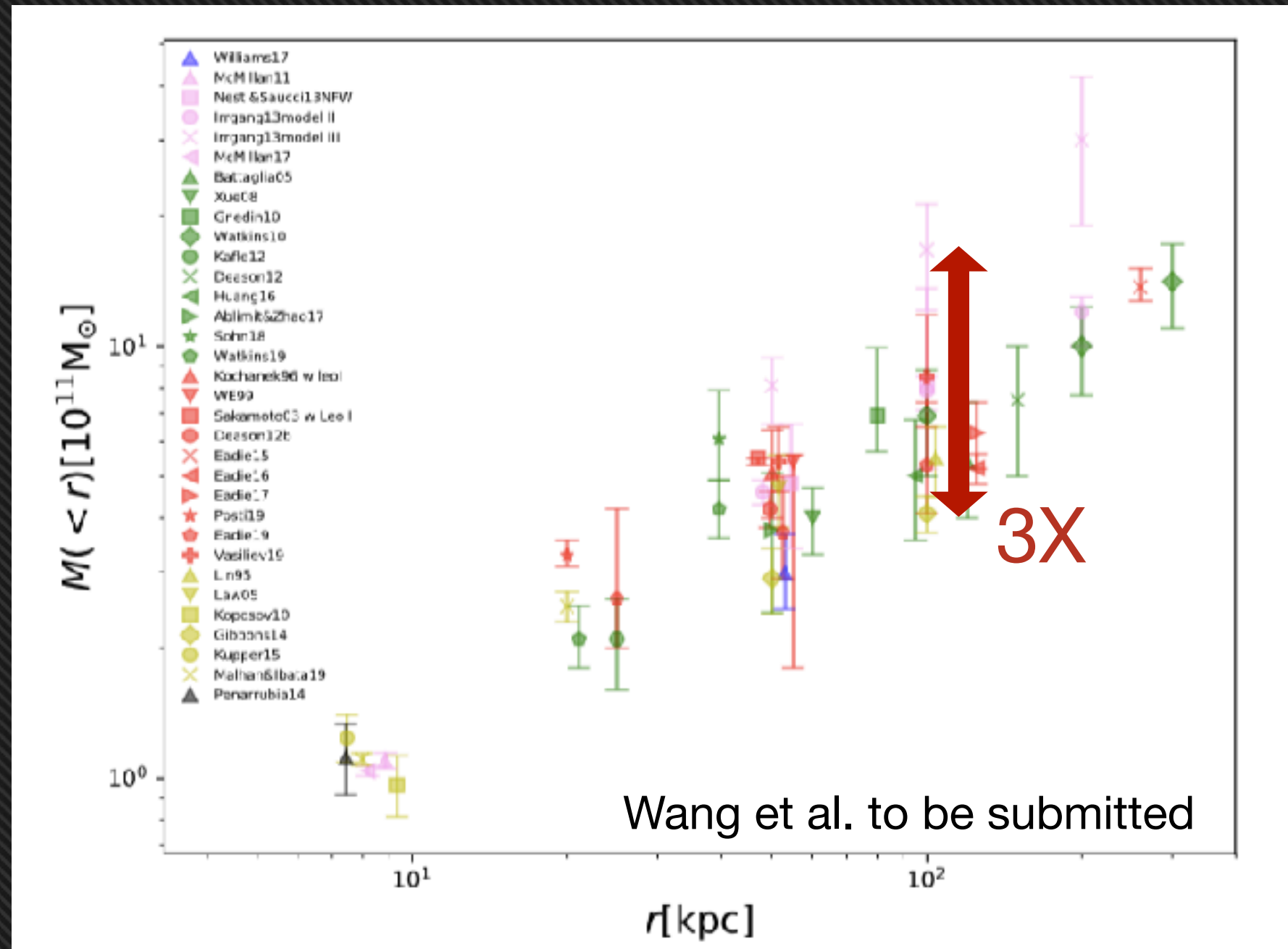


Challenges in data & model

- Virial radius is far beyond tracers like stars
- Poor observations in outer halo
- Model uncertainty: $\beta(r)$, $n(r)$ or form of DF
- Observation error and incompleteness



Milky Way (MW) Mass: still very uncertain!



Challenges in data & model

- Virial radius is far beyond tracers like stars

Use satellite galaxies instead

- Poor observations in outer halo

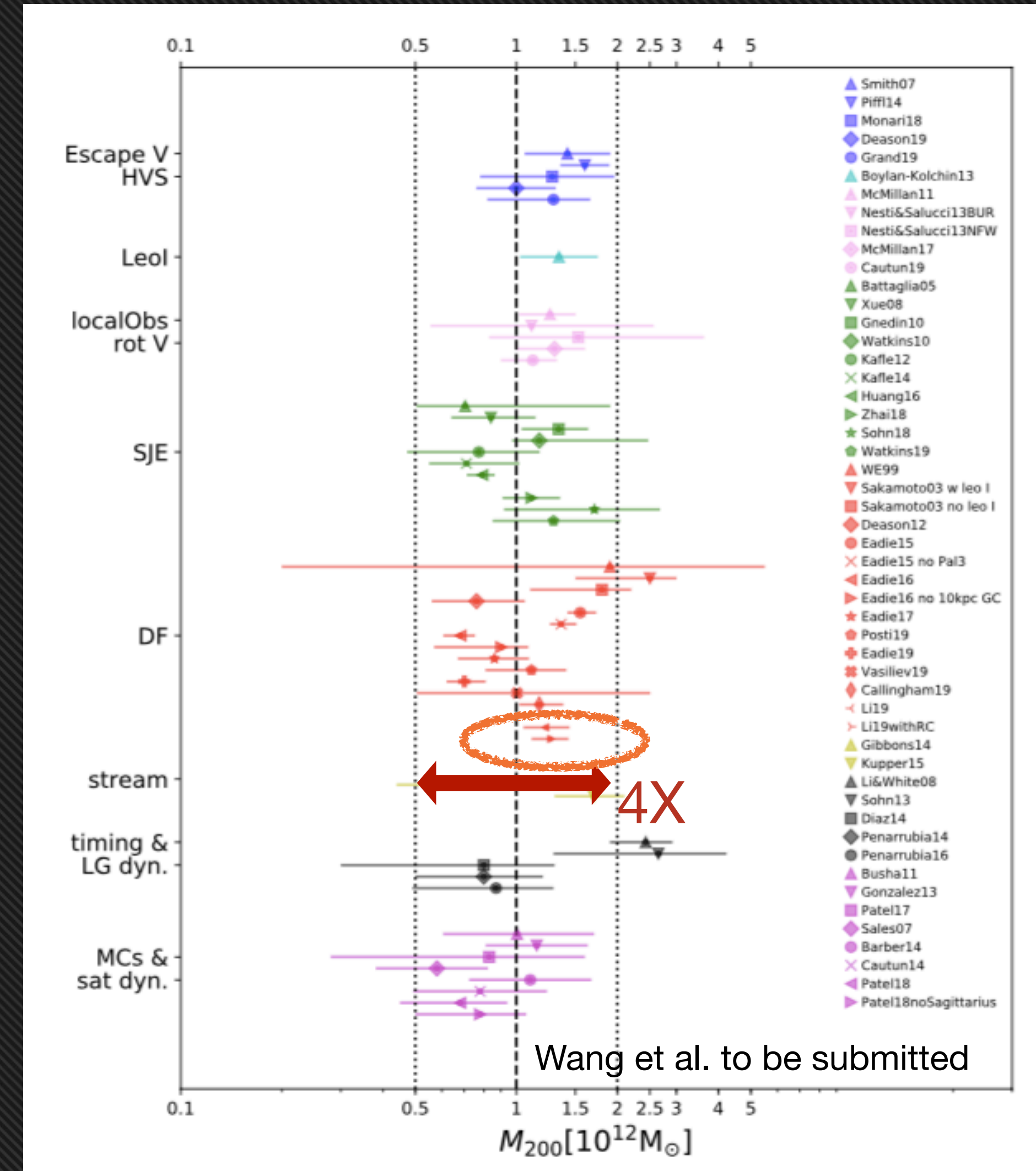
Gaia

- Model uncertainty: $\beta(r)$, $n(r)$ or form of DF

Dynamical model based on simulation

- Observation error and incompleteness

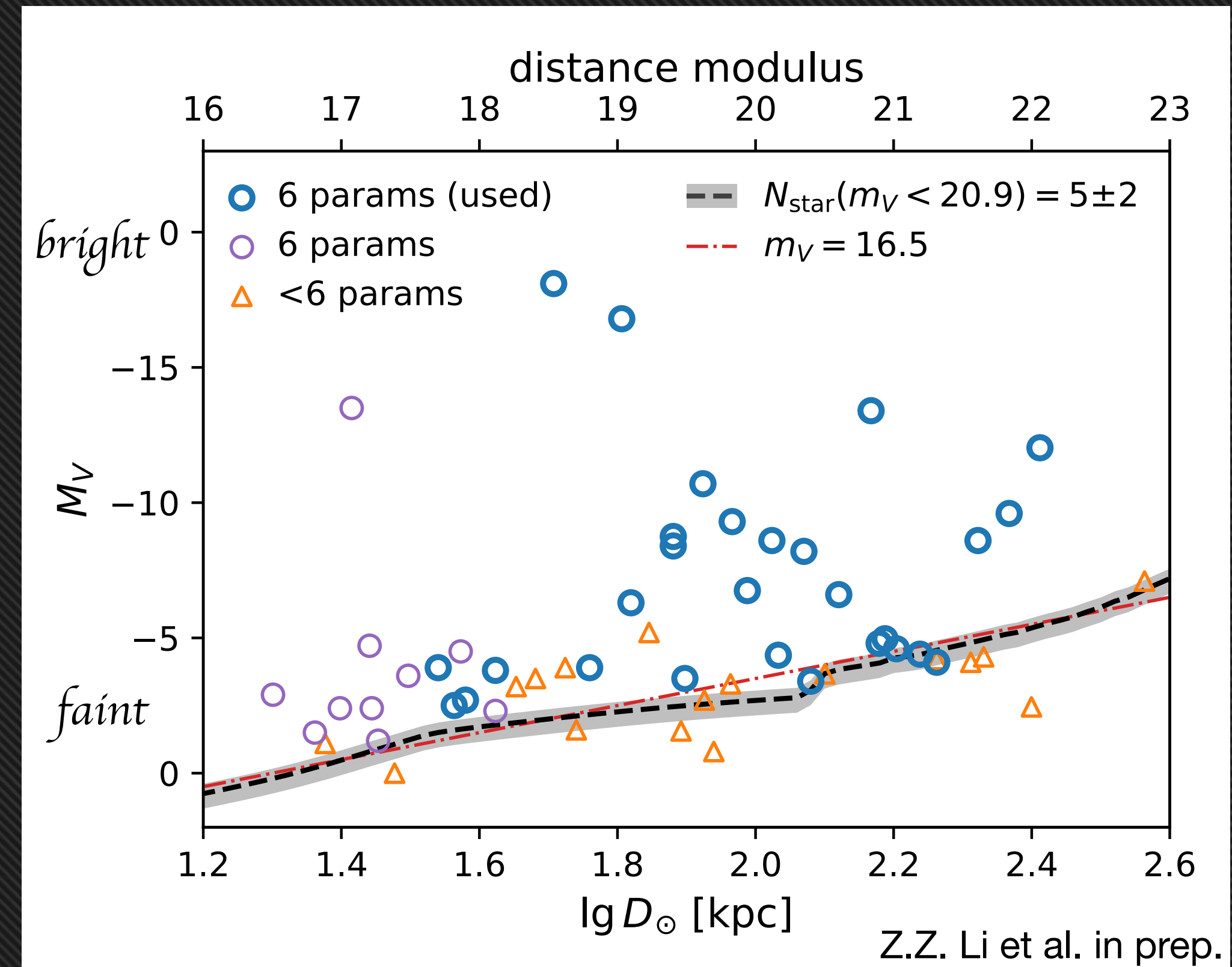
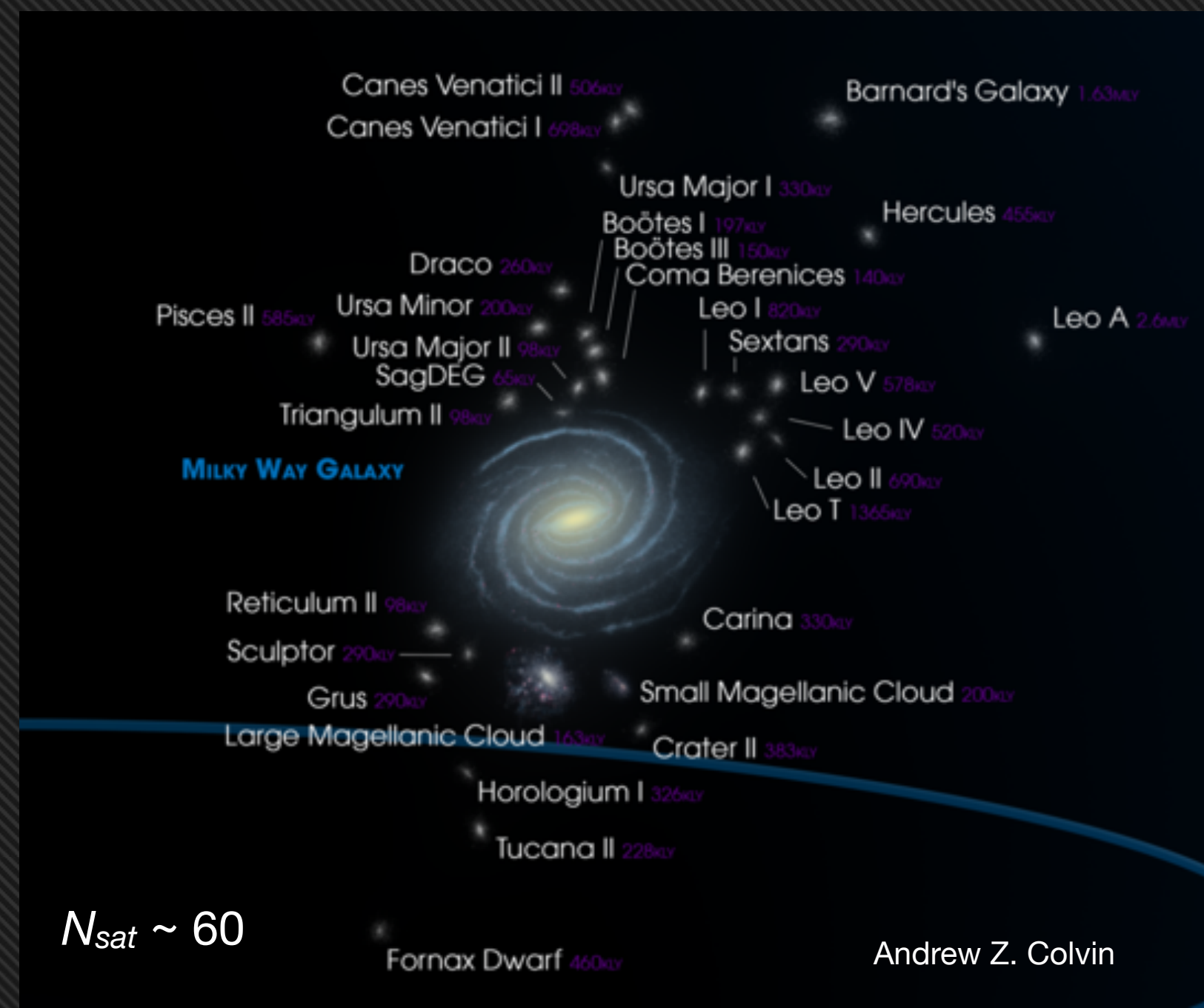
Bayesian statistics



Satellite galaxies: best tracers for MW outer halo

Virtues:

- Extended distribution
the **only** tracers beyond 100 kpc or farther
- Proper motion (pm) available: Gaia
- Well understood population
knowledge from theory and simulations
well phase-mixed (Han+ 2019)
⇒ **physical model based on simulation and equilibrium assumption**



Data:

- 28 satellites with Gaia DR2 pm
40 kpc < r < 280 kpc
- homogeneous selection function:
distant and faint galaxies are missed
 $N_{\text{star}}(M_V < 20.9) \sim 5$

How kinematics reveal mass

Popular methods: Jeans Equation, Distribution Function

Poor observation \Rightarrow heavier **model dependence** for outer halo

Extract info from **simulations** to mitigate model dependence

- ◆ statistics: instantaneous kinematics (*Busha+ 2011, Boylan-Kolchin+ 2013, Patel+2017, ...*), orbit circularity (*Barber+ 2013*), angular momentum (*Patel+2018*)
- ◆ **orbital distribution**: $p(E, L | M)$ (*Li+ 2017, Callingham+ 2018*)
more efficient use of both data and simulation
 E is not observable \Rightarrow calibration required

- ◆ **phase space distribution function**

$$p(\mathbf{r}, \mathbf{v}) = \frac{p(E, L)}{8\pi^2 L T_r(E, L)} \equiv f(E, L)$$

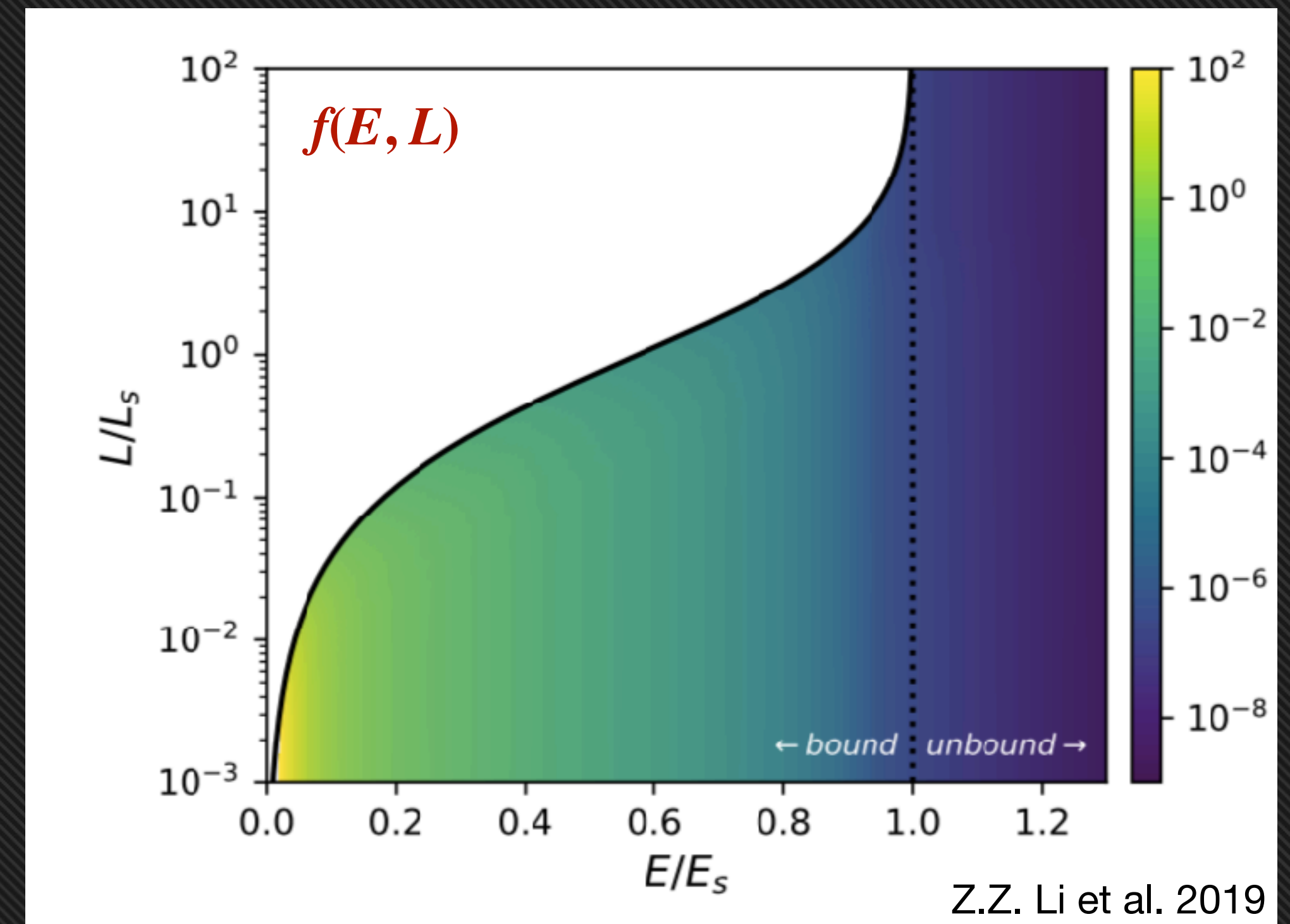
similarity in halo structures \Rightarrow stacking & scaling

$$p(\mathbf{r}, \mathbf{v} | M, c) = \frac{1}{r_s^3 v_s^3} \tilde{f}\left(\frac{E}{v_s^2}, \frac{L}{r_s v_s}\right)$$

complete description to satellite kinematics

precise & unbiased

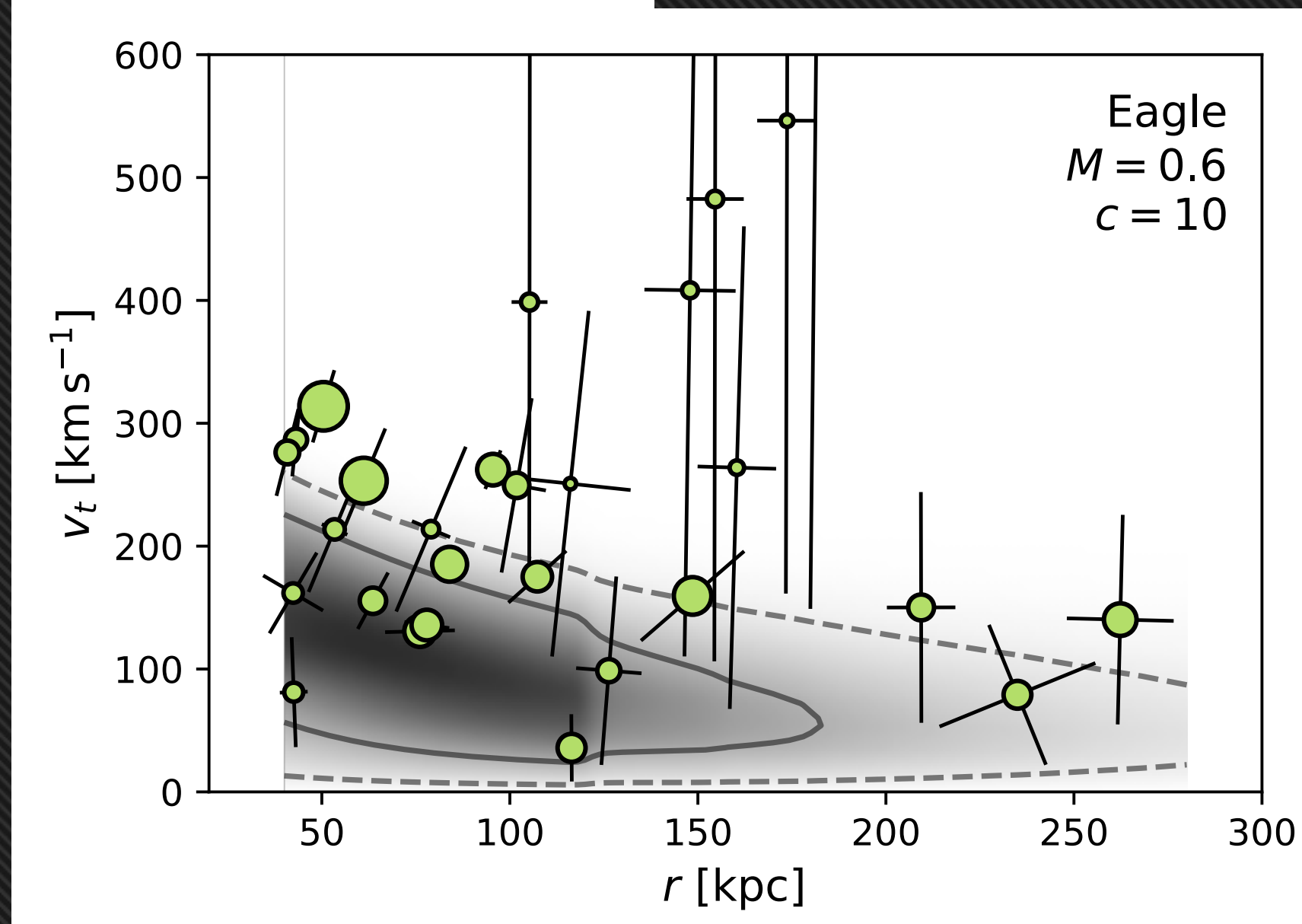
rigorous treatment to observation error and incompleteness



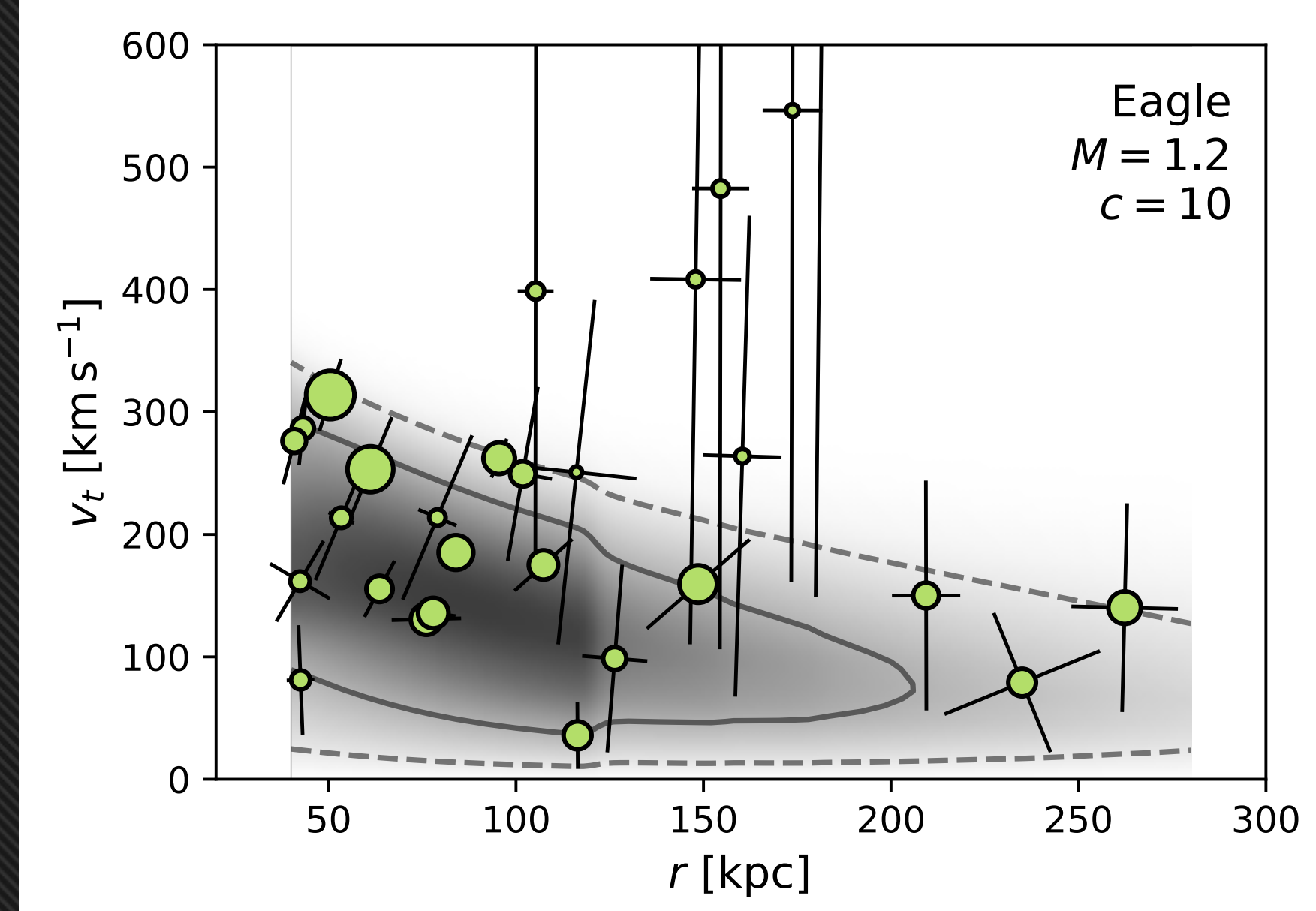
Infer Mass & concentration

Observation (r, v) vs Model $p_{\text{obs}}(r, v | M, c)$

$M = 0.6 \times 10^{12} M_{\odot}$



$M = 1.2 \times 10^{12} M_{\odot}$



Likelihood:

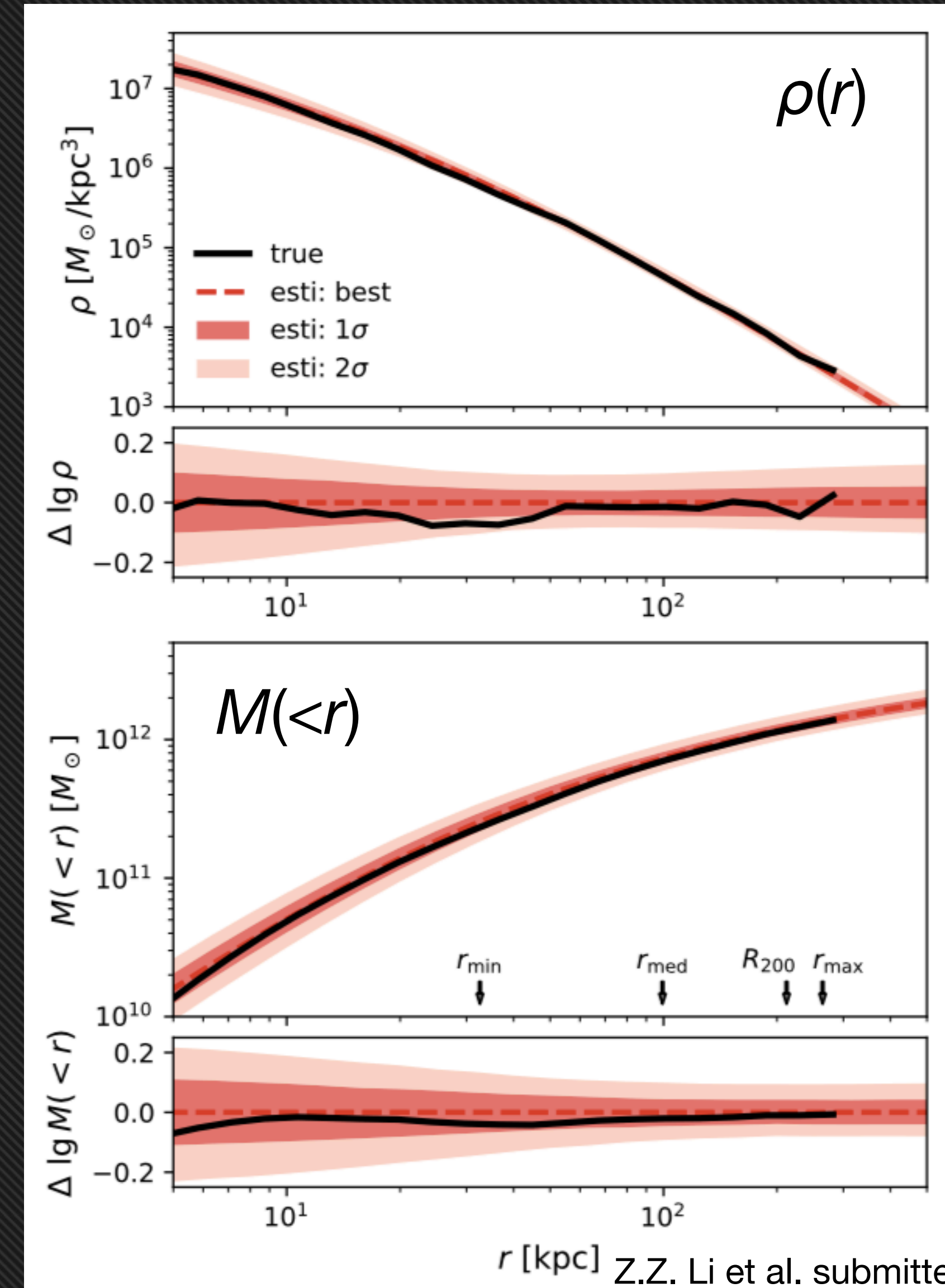
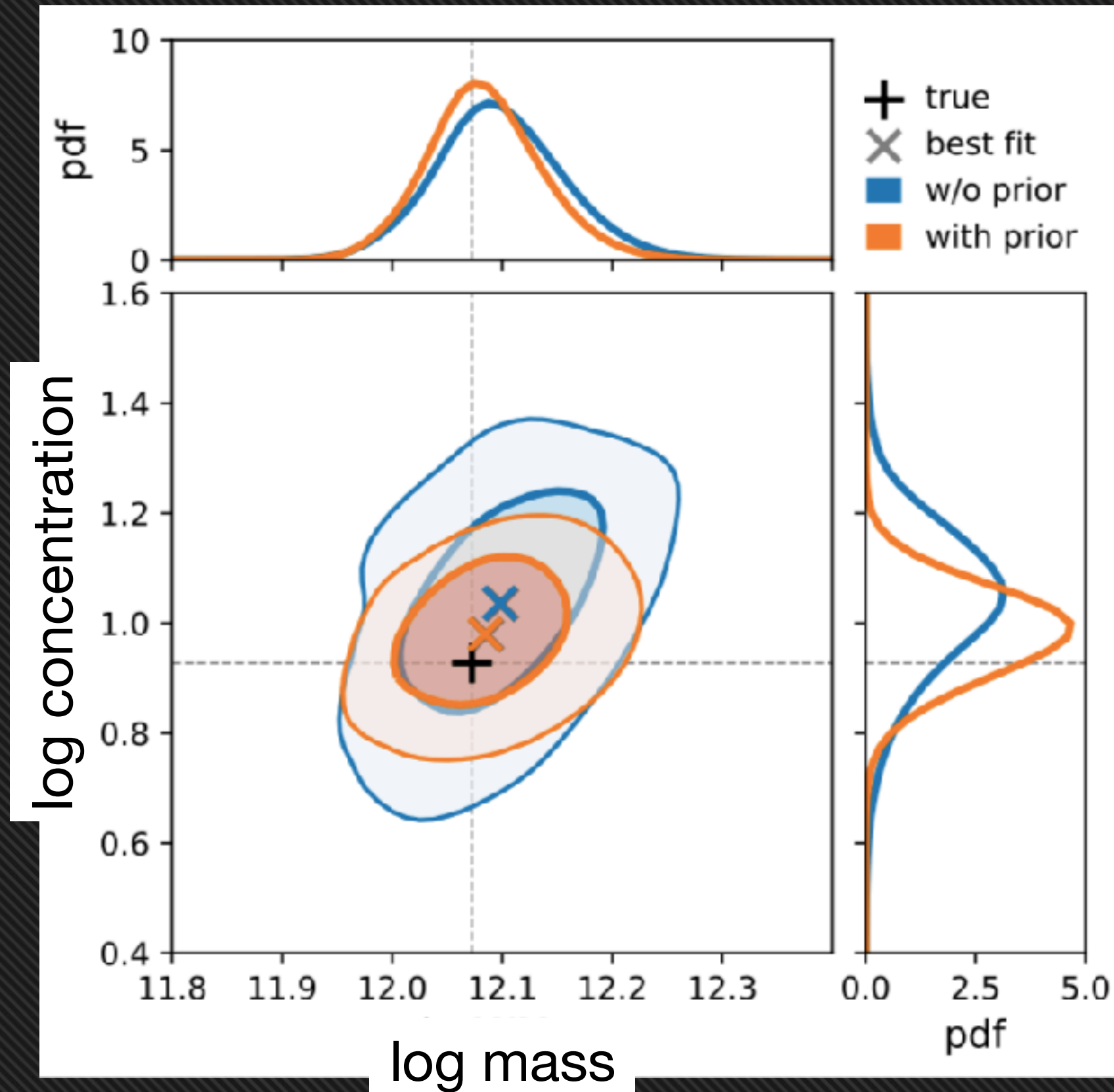
$$p(M, c | \{w\}) \propto \left[\prod_{i=1}^{n_{\text{sat}}} p_{\text{obs}}(w_i | M, c) \right] p(c | M) p(M)$$

Observational effects included

Prior information: e.g. M - c relation, or other indep. measurements

Test with mock sample from simulation

Example using 40 tracers
~15% precision level for halo mass
well recover the mass profile



Method comparison: better than Jeans method

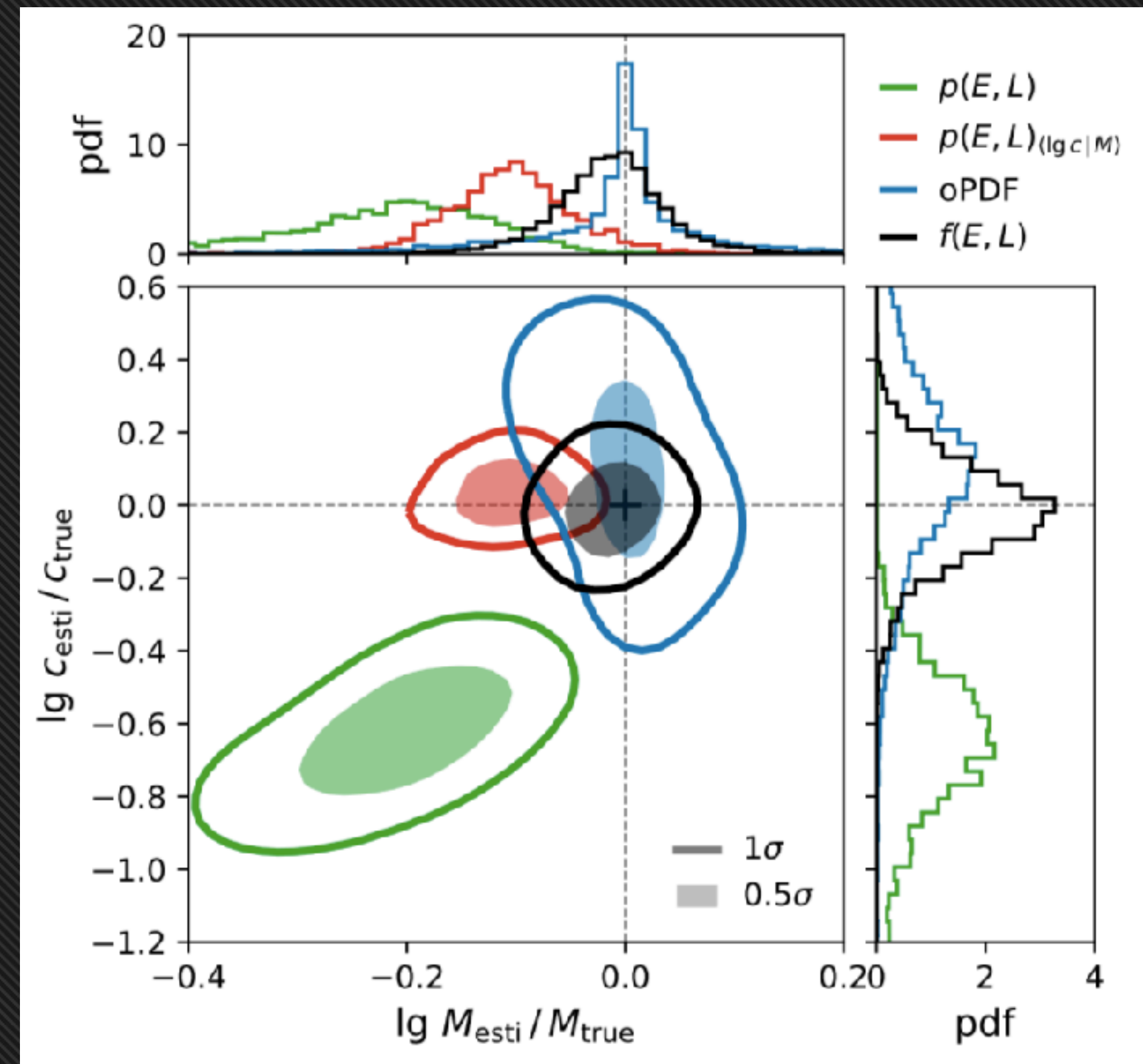
- ✦ $p(E, L | M, c)$: orbital distribution
 - $p(E, L | M)$: Li+ 2017; Callingham+ 2018; M - c relation assumed
- ✦ oPDF (Han+ 2015): based on steady-state assumption representative to **Jeans equation** and **Schwarzschild modeling**
- ✦ $f(E, L | M, c)$: phase space distribution (this work)

phase space distribution = radial distribution along one orbit × orbits distribution

$$p(\mathbf{r}, \mathbf{v}) = \frac{|\mathbf{v}_r|}{8\pi^2 L} p(r | E, L) \times p(E, L)$$

steady state (oPDF) ⇔

$$p(\mathbf{r}, \mathbf{v}) = \frac{p(E, L)}{8\pi^2 L T_r(E, L)} \equiv f(E, L)$$

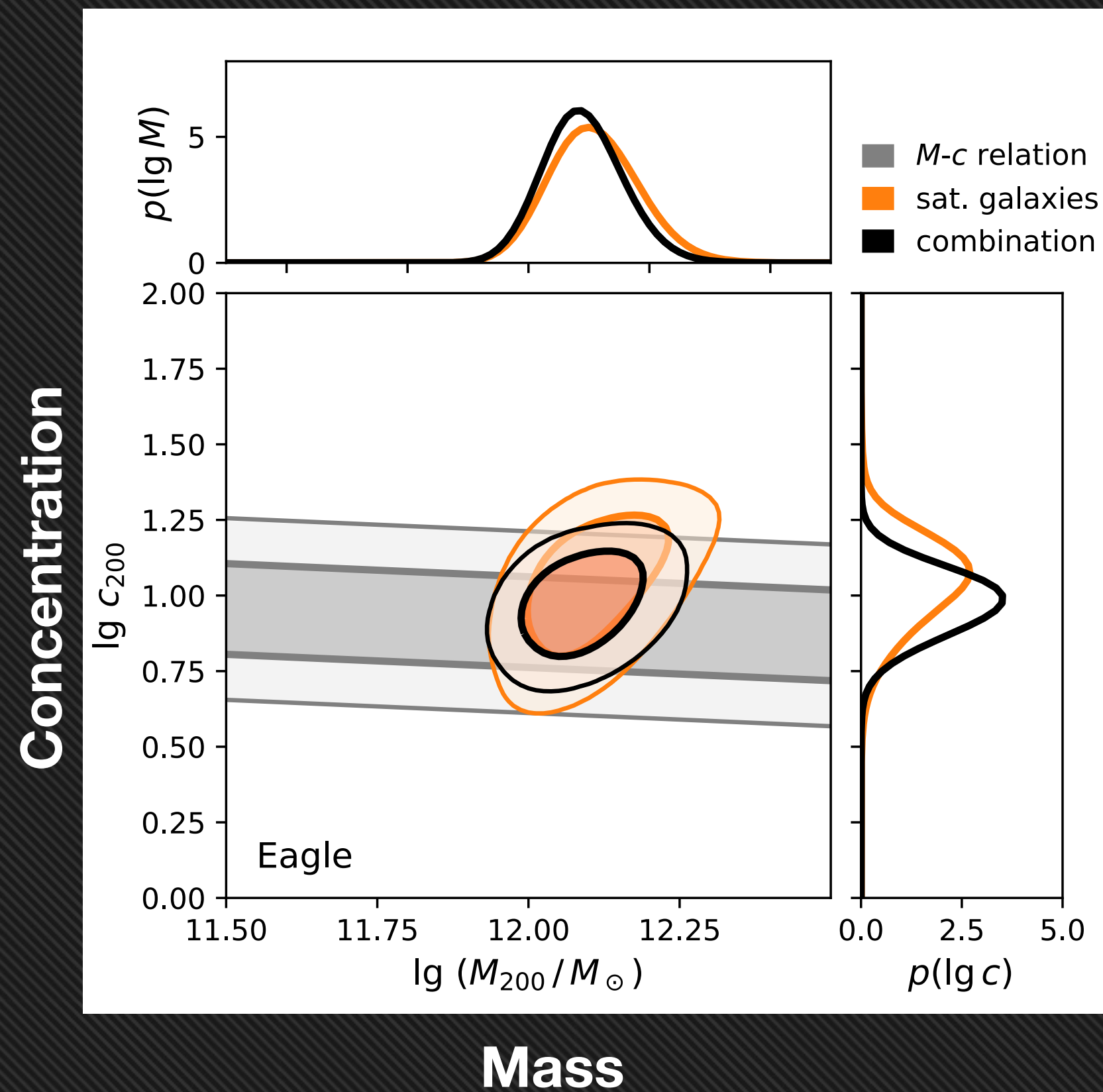


Current best estimation to MW halo mass

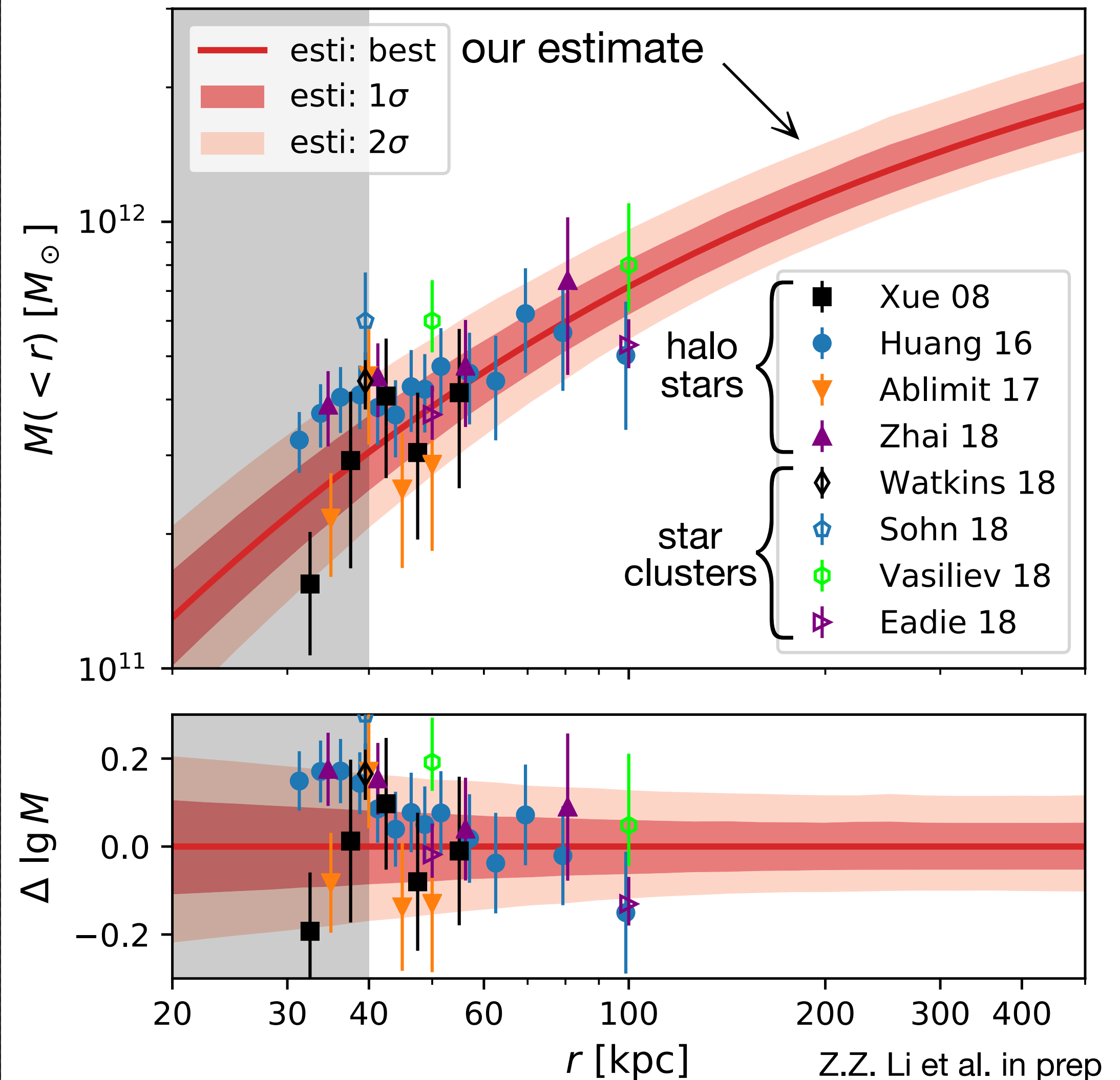
DF model based on **Eagle** simulation
 28 MW satellites with **Gaia** measurement
 ⇒ Constrain total halo mass within **20%**

$$M = 1.23^{+0.21}_{-0.18} \times 10^{12} M_{\odot}$$

$$c = 9.4^{+2.8}_{-2.1}$$

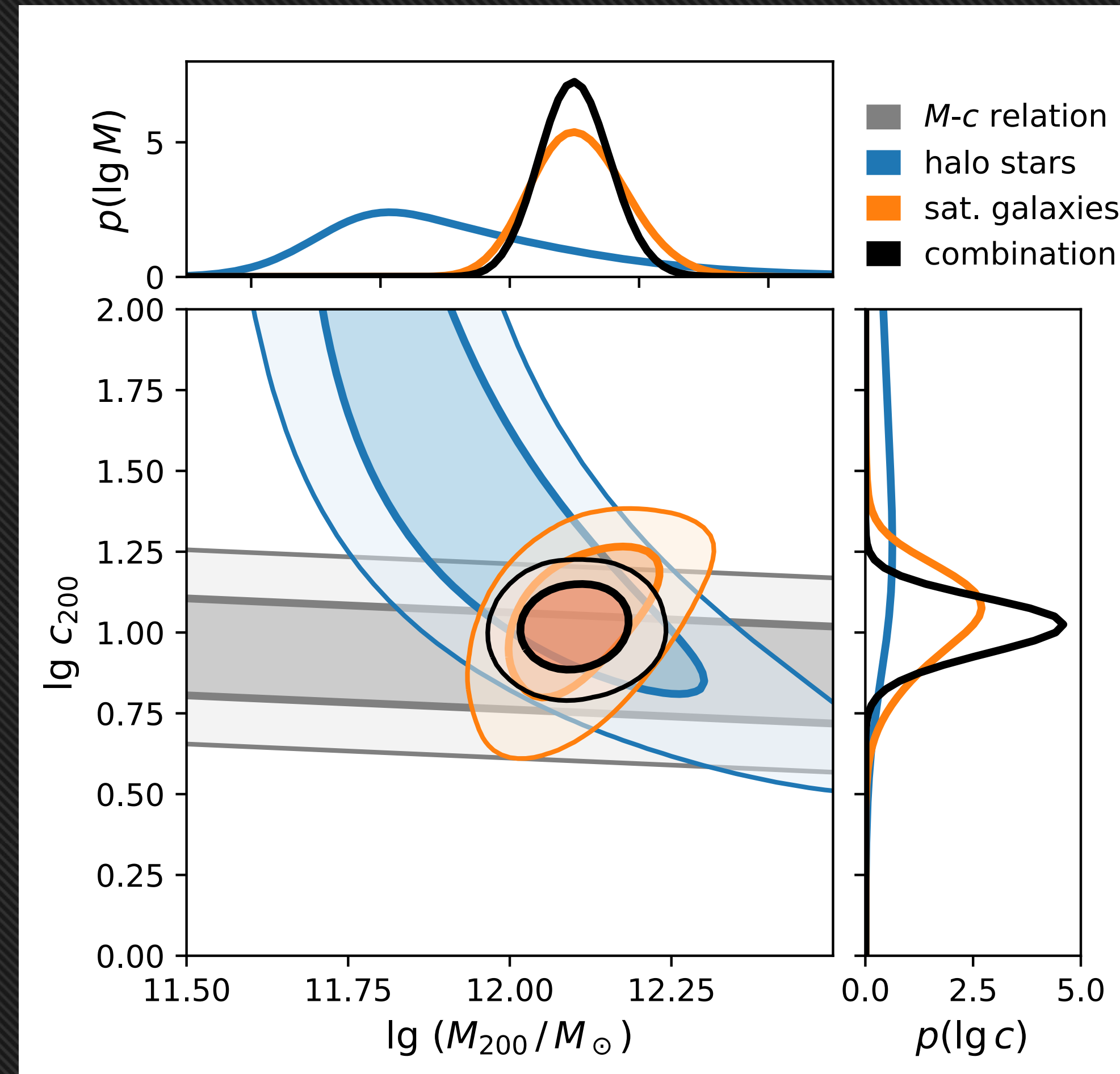


Inferred MW mass profile



Even better: using multiple tracer populations

Satellite Galaxies
 ⇨ Total halo mass
 Satellites + Stars
 ⇨ Shape of potential



| | Satellites | | Satellites + Halo Stars | |
|-------|------------------------|--|-------------------------|------------------------|
| | flat prior | M - c relat. | flat prior | M - c relat. |
| M^a | $1.29^{+0.24}_{-0.20}$ | $1.23^{+0.21}_{-0.18}$ | $1.27^{+0.17}_{-0.15}$ | $1.26^{+0.17}_{-0.15}$ |
| c | $11.0^{+4.8}_{-3.3}$ | $9.4^{+2.8}_{-2.1}$ | $11.7^{+3.2}_{-2.5}$ | $10.4^{+2.3}_{-1.9}$ |

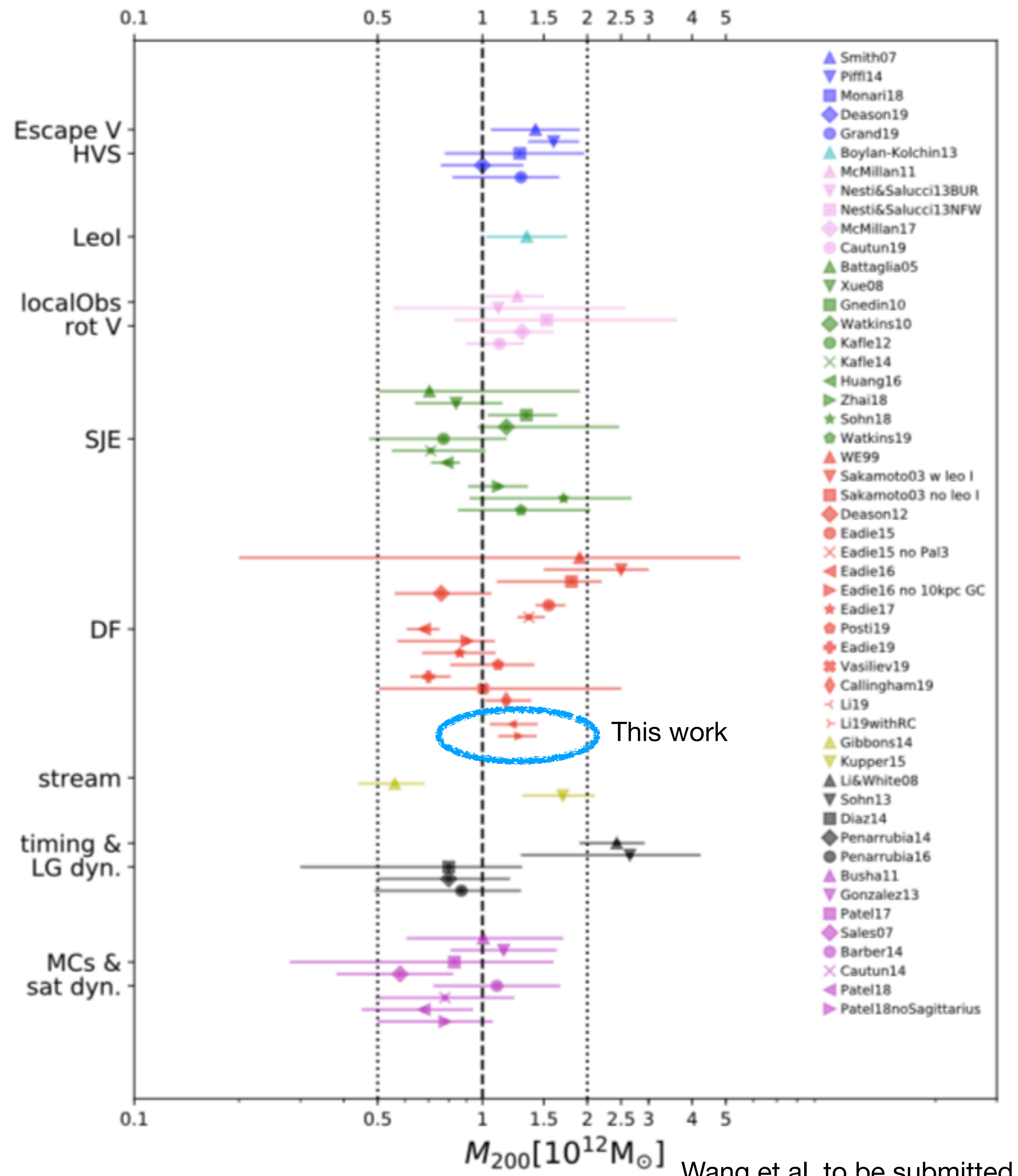
^a The unit of M is $10^{12} M_{\odot}$.

M - c relation: Dutton & Macciò 2014

Halo stars: 40 to 80 kpc

Rotation curve measured by Huang+ 2016

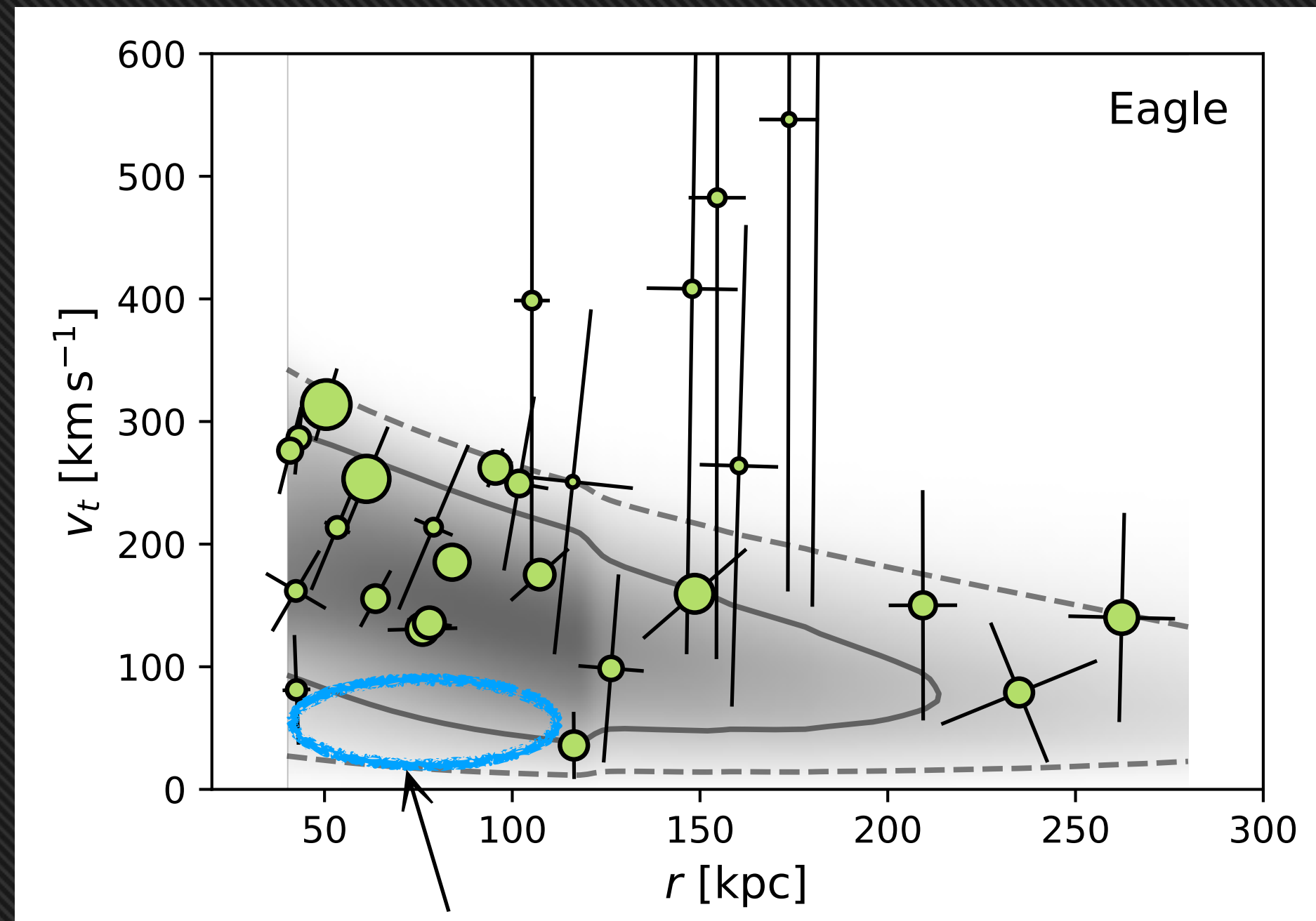
Satellites: 40 to 300 kpc



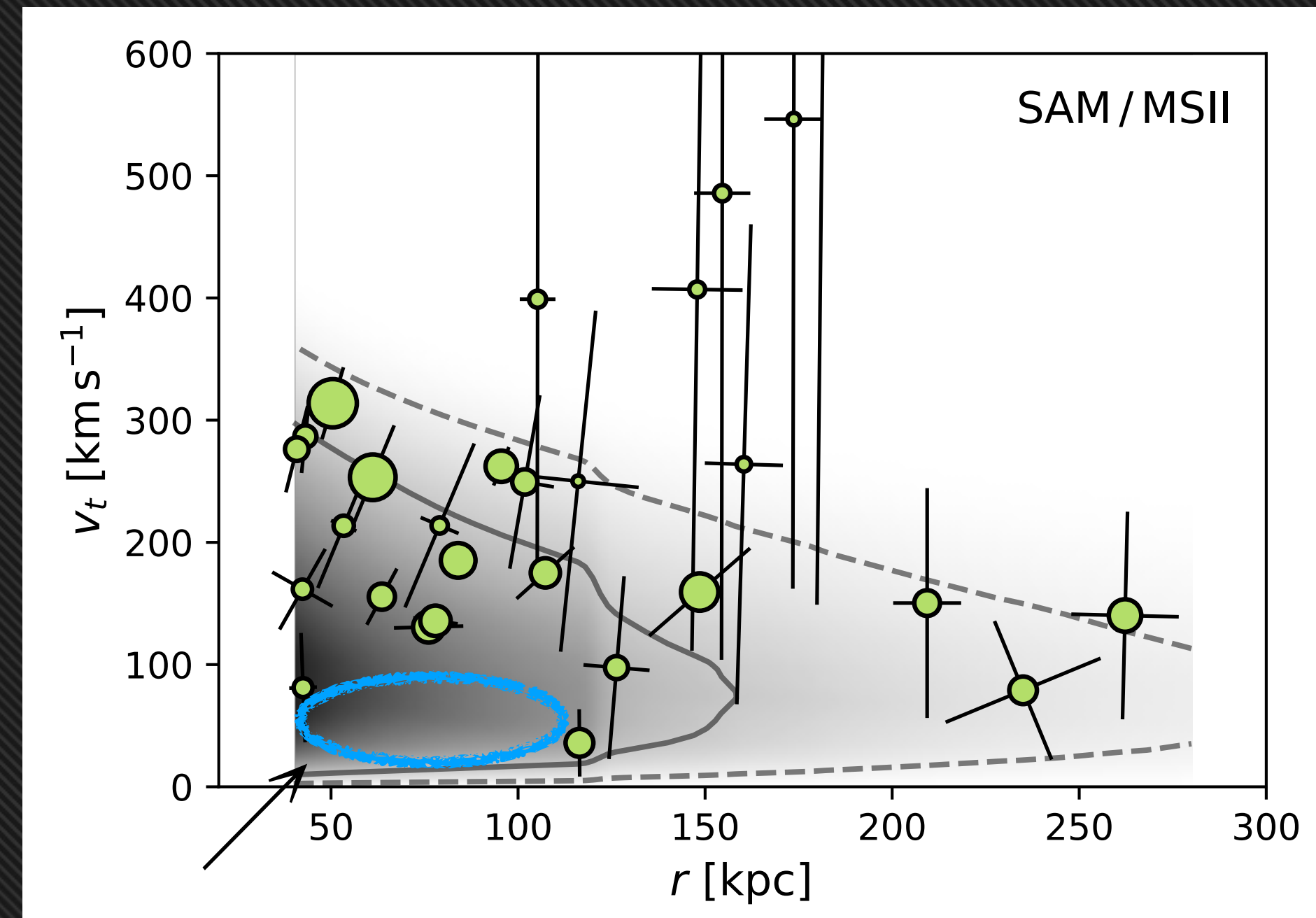
Hydro simulation matches MW satellites better

The Bayesian Evidence of Eagle is 25 times higher than SAM

Hydro simulation



Semi-analytical model on dark simulation



Orbits of small pericenter distance

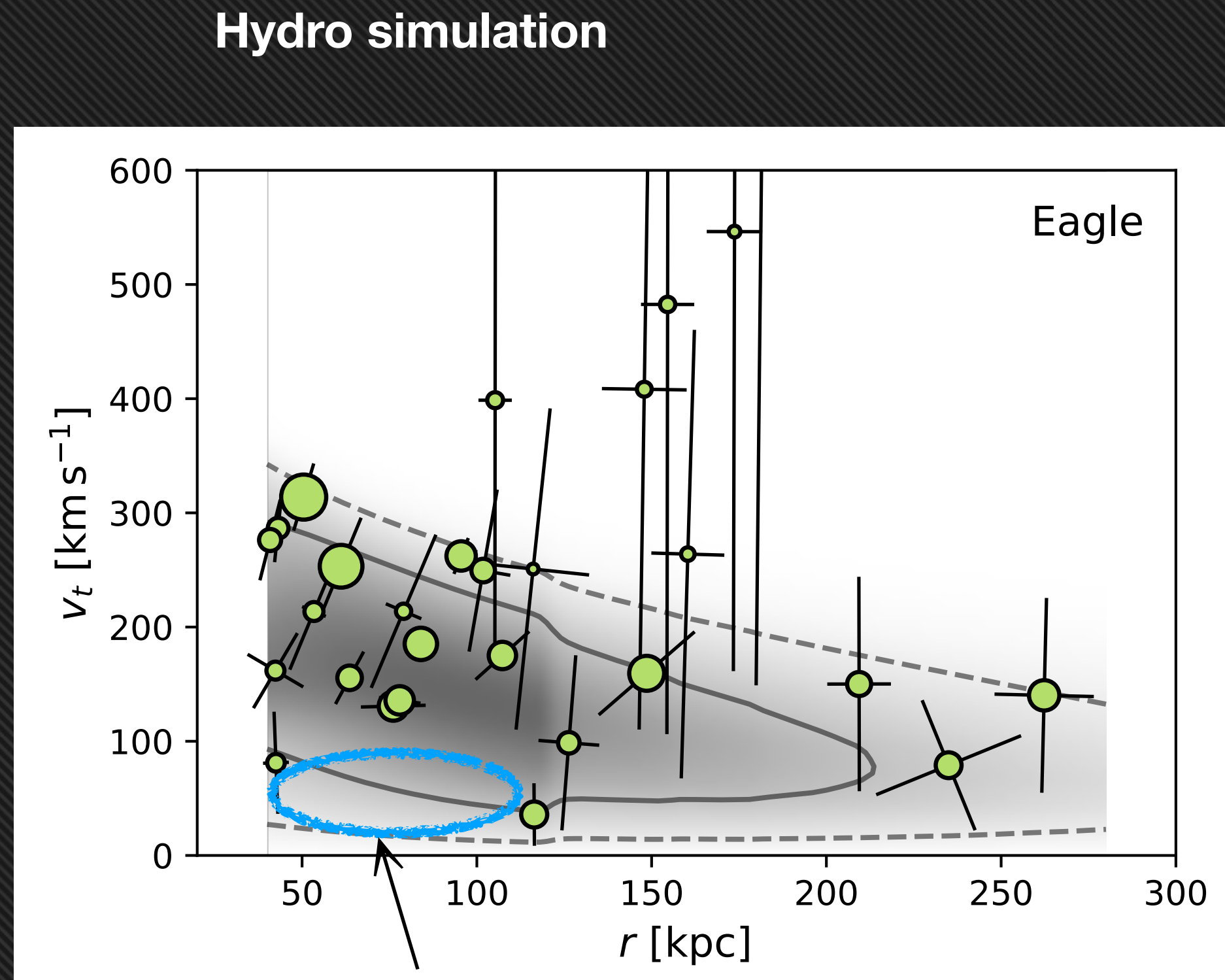
Enhanced satellite disruption due to stellar disc in hydro and real world

Circles: observed satellites

Shade: expected distribution predicted by DF based on simu.

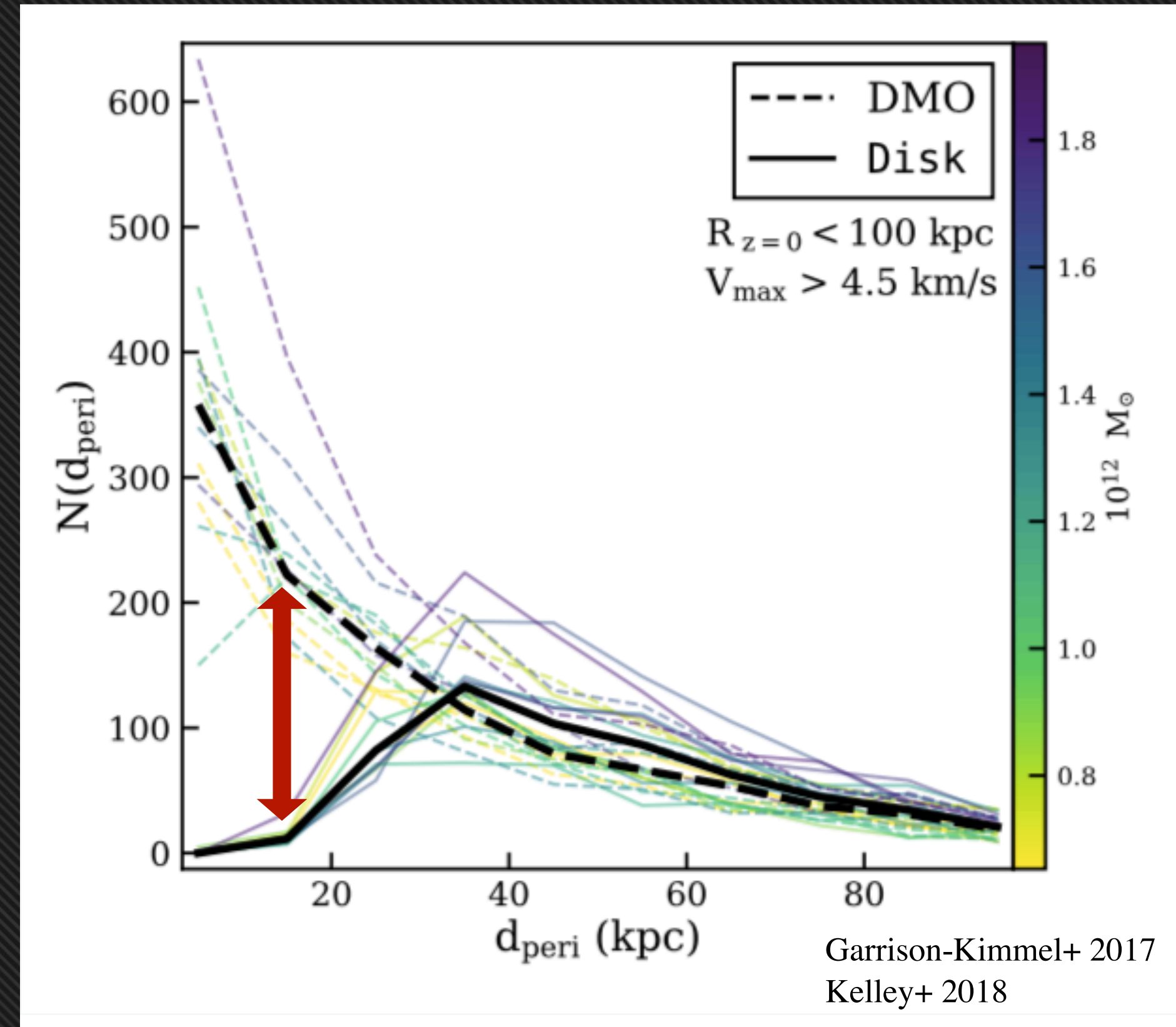
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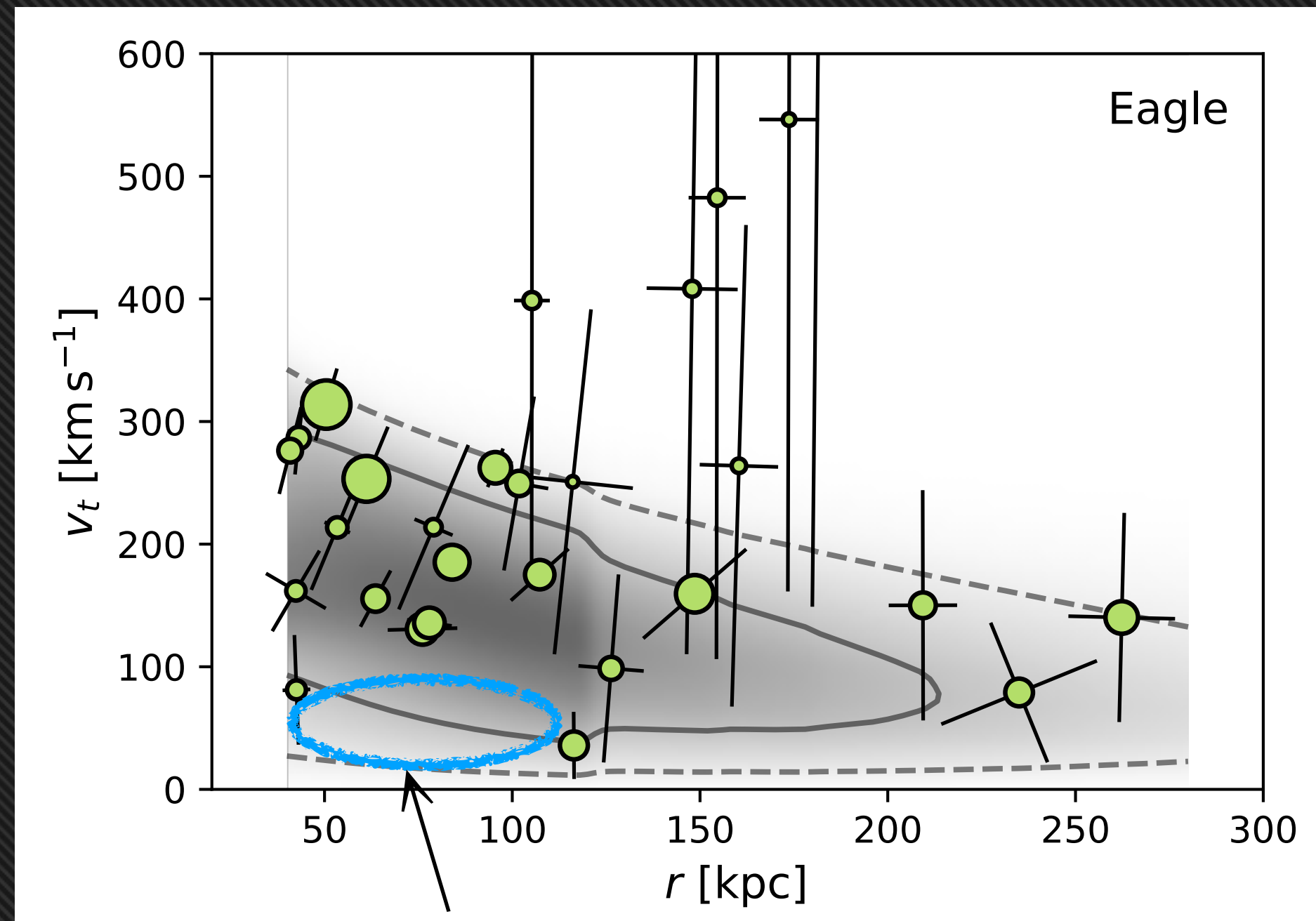
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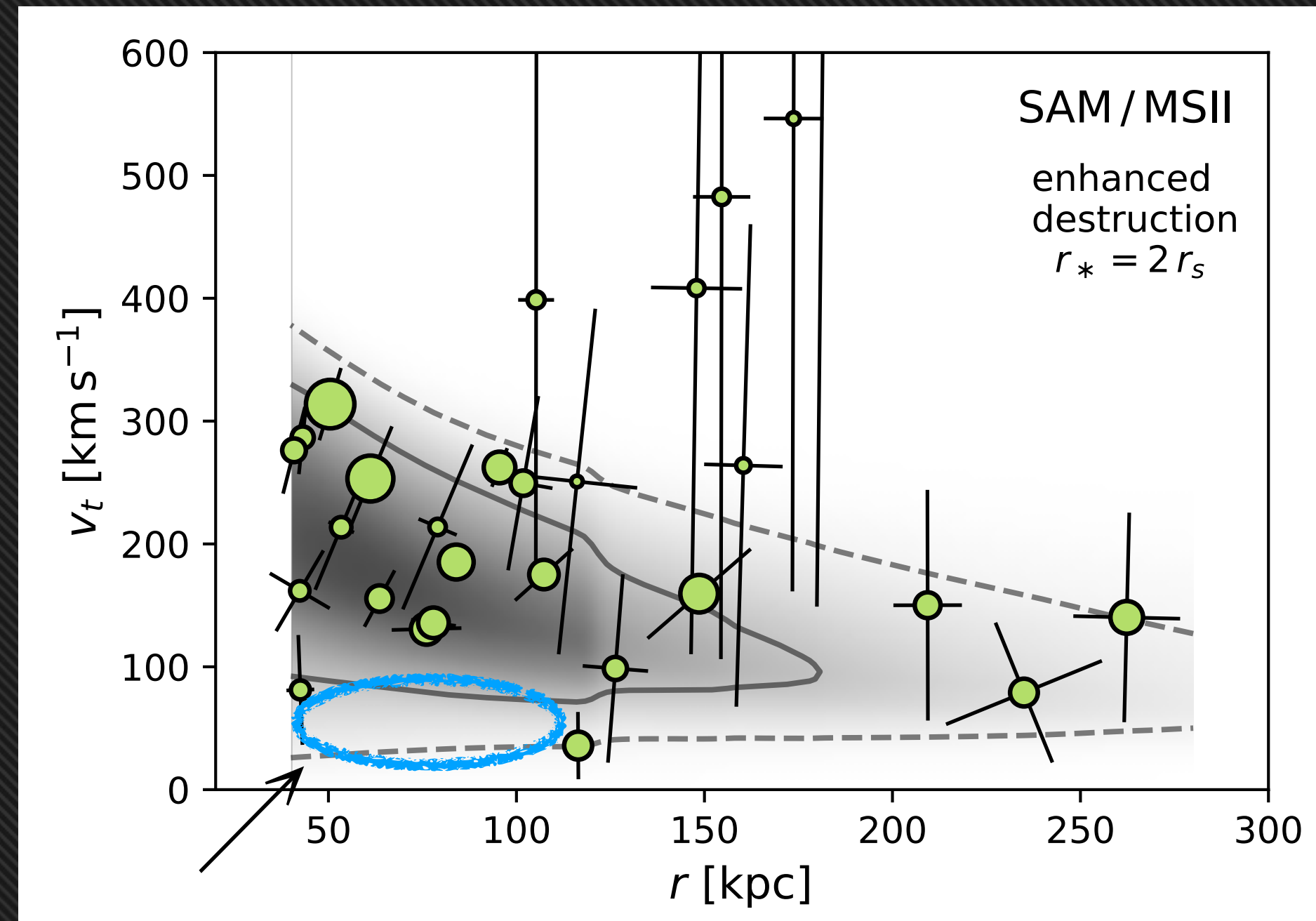
Hydro simulation matches MW satellites better

Mimic the extra disruption by manually remove a fraction of orbits with small d_{peri} in SAM

Hydro simulation



Semi-analytical model on dark simulation with enhanced disruption



Orbits of small pericenter distance

Enhanced satellite disruption due to stellar disc in hydro and real world

Circles: observed satellites

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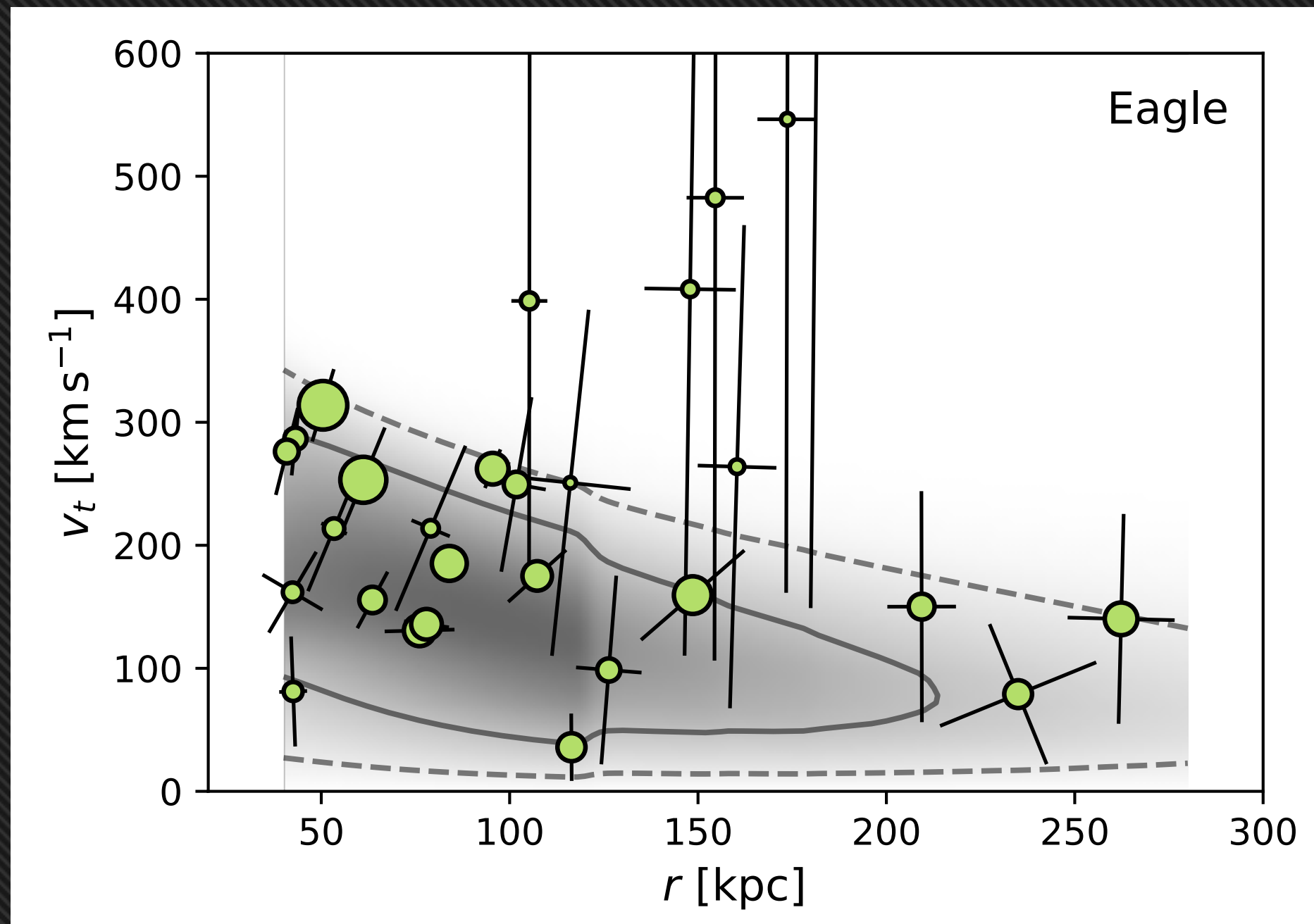
Bonus: reducing uncertainty in satellite orbits

Satellite orbits reveal their infall history and the assembly of MW

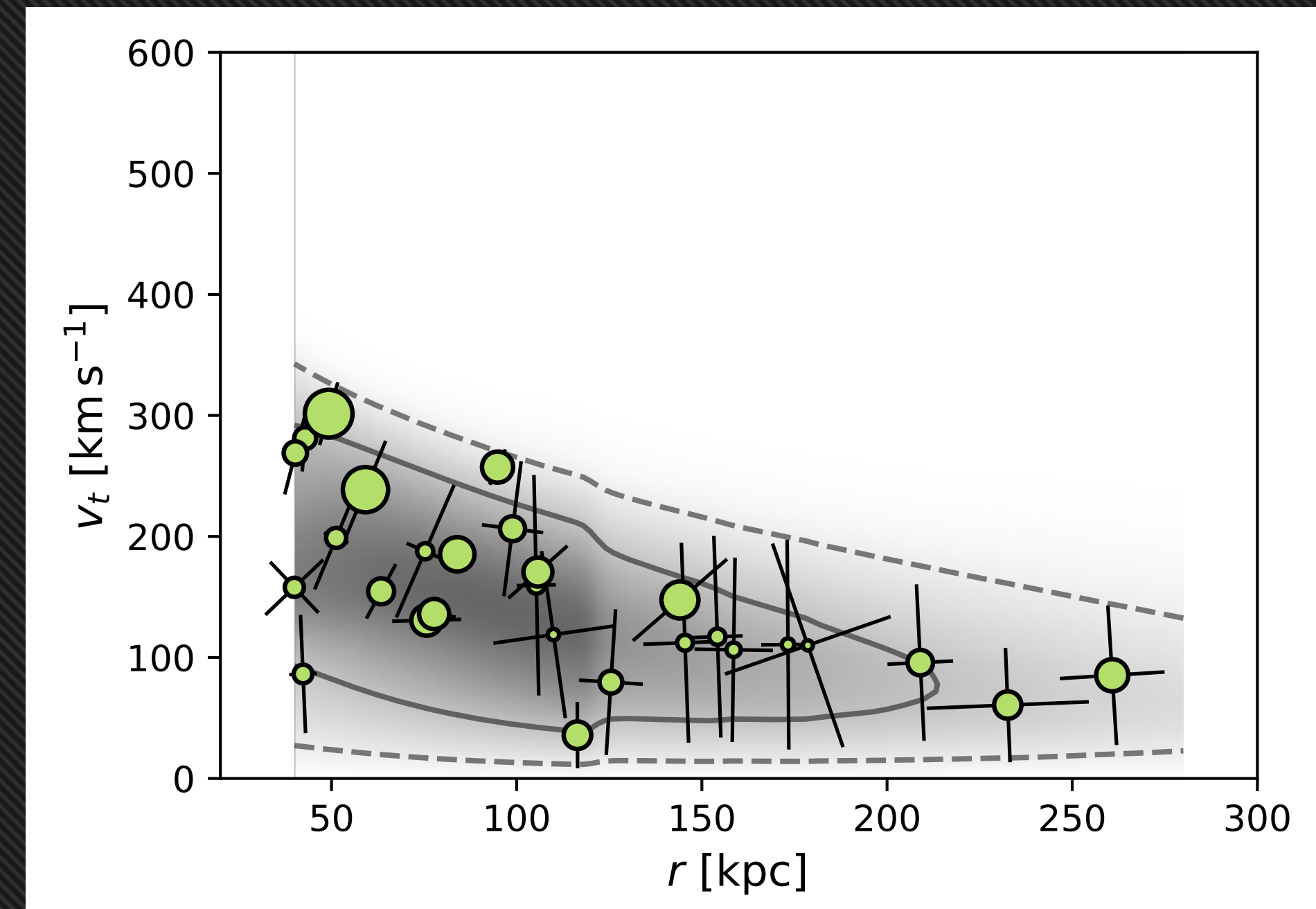
The usage is blocked by large uncertainty in proper motion

Using DF model as prior can **reduce the uncertainty** in satellite orbits

Raw measurements



Posterior kinematics



Summary and Outlook

- Current **BEST** estimation to MW halo mass
 - ✓ best **tracer** for outer halo: satellite galaxies
 - ✓ best **data** available: 28 satellites with *Gaia* DR2 proper motion
 - ✓ realistic **model**: physical DF from simulation → wide usage
 - ✓ rigorous **statistics**: Bayes for selection function, observational error

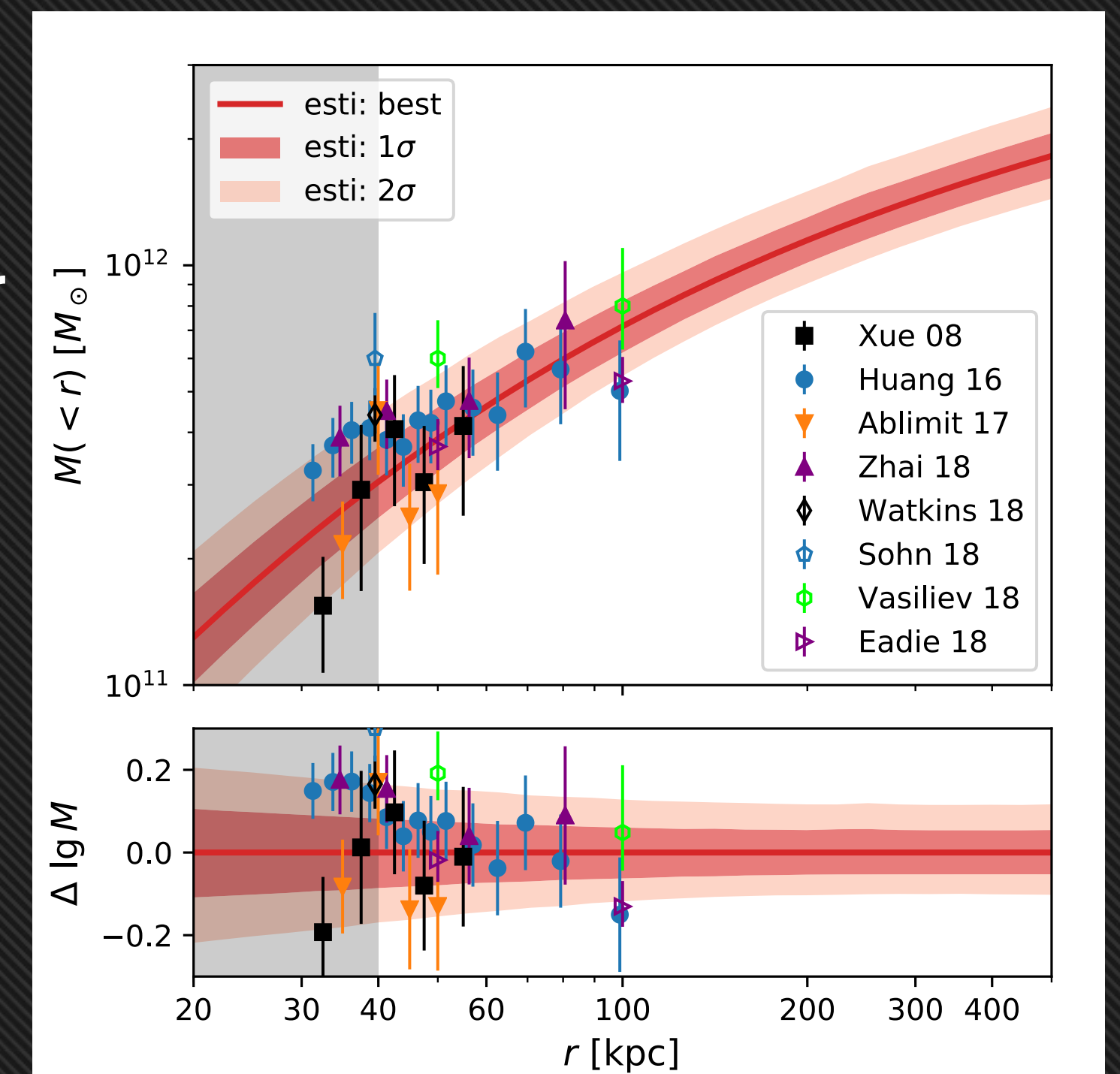
Li et al. 2017 ApJ, 850, 116

Li et al. 2019 arxiv:1910.11257

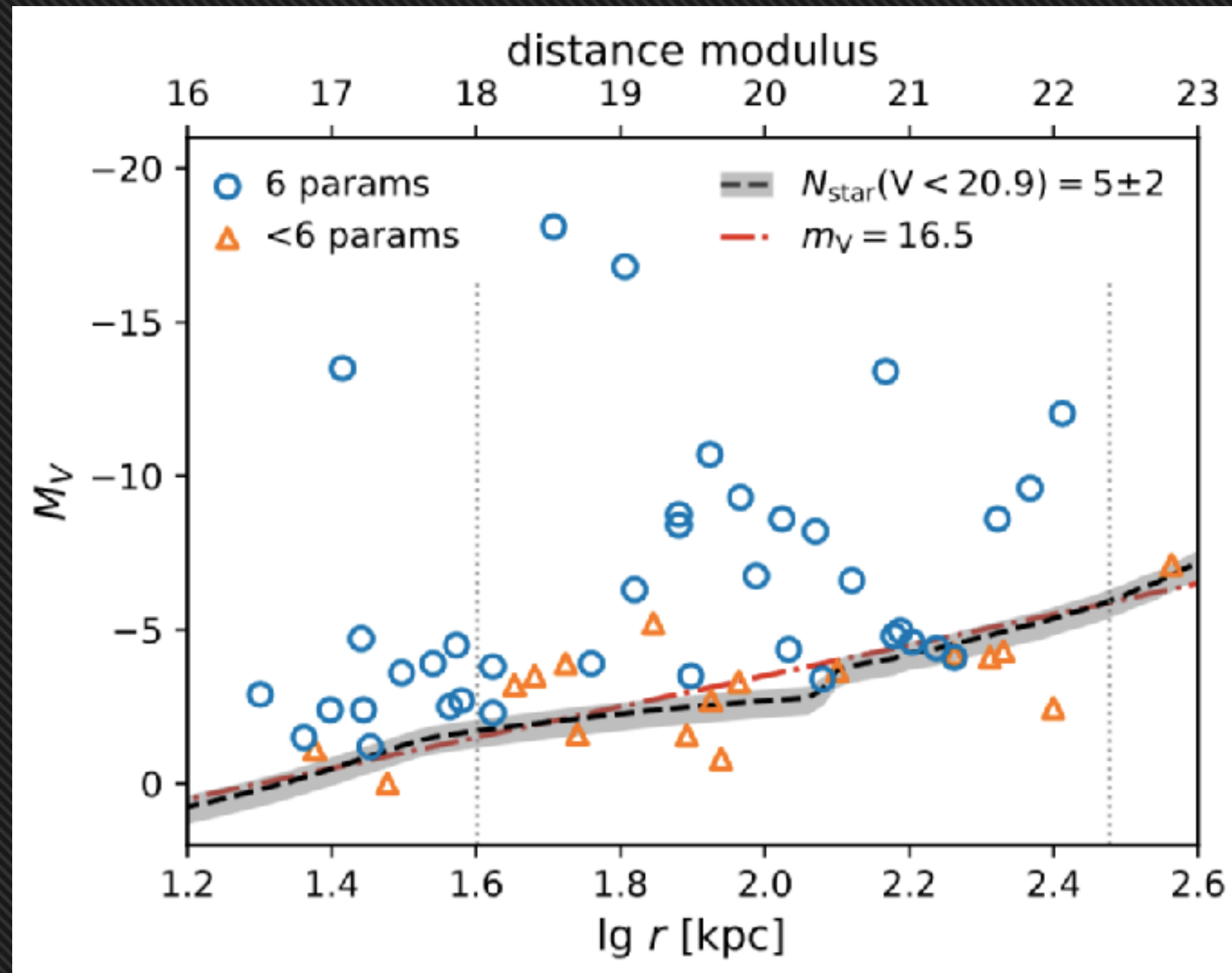
Li et al. 2019 in prep, coming soon!

- Hydro simulations indeed match satellite dynamics better
- Future improvements in MW mass
 - more satellites
 - combination with other tracers (e.g. halo stars, star clusters)
 - peculiarities of MW and its history
- The DF construction method can apply to other tracers or galaxy groups & clusters

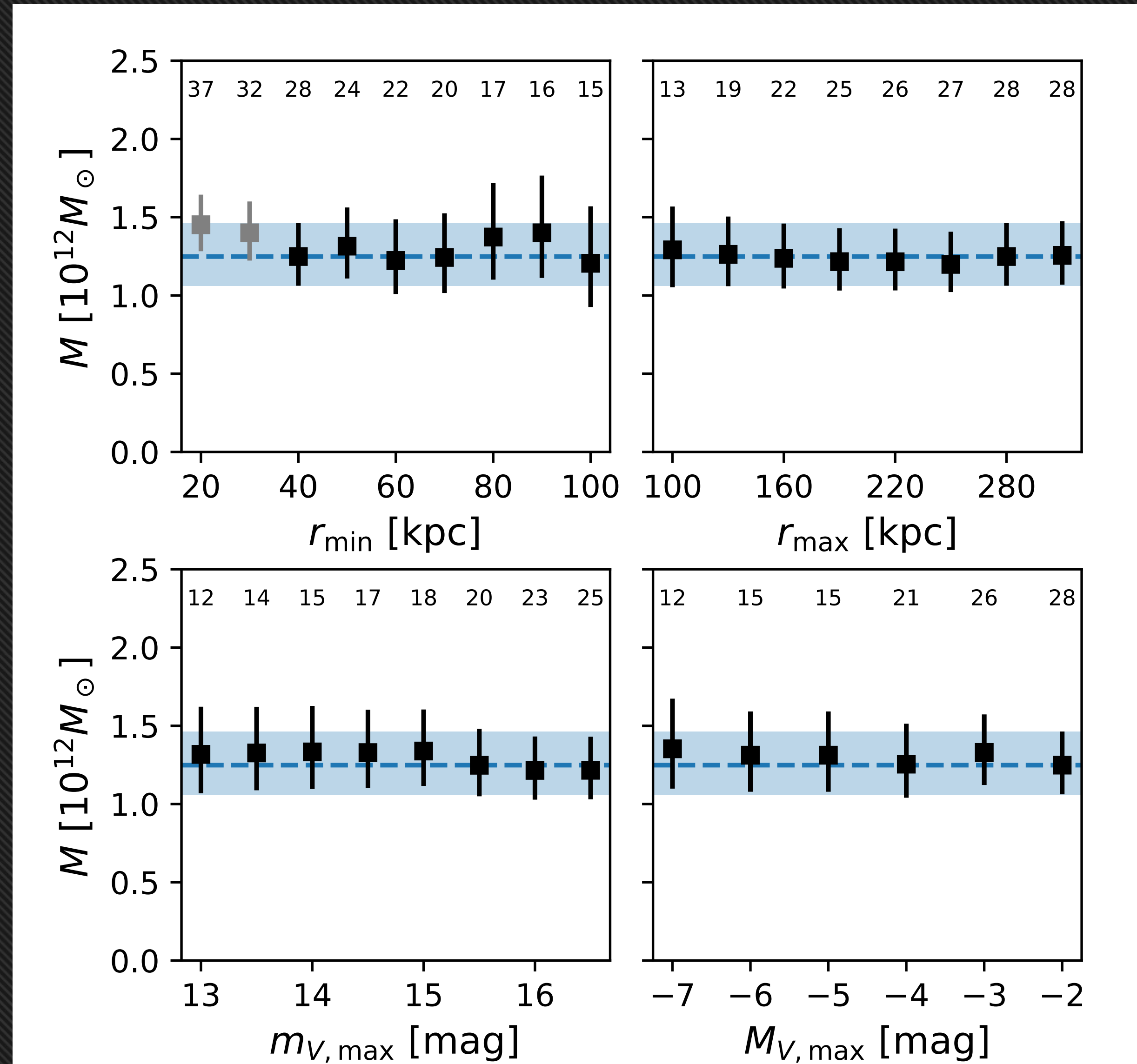
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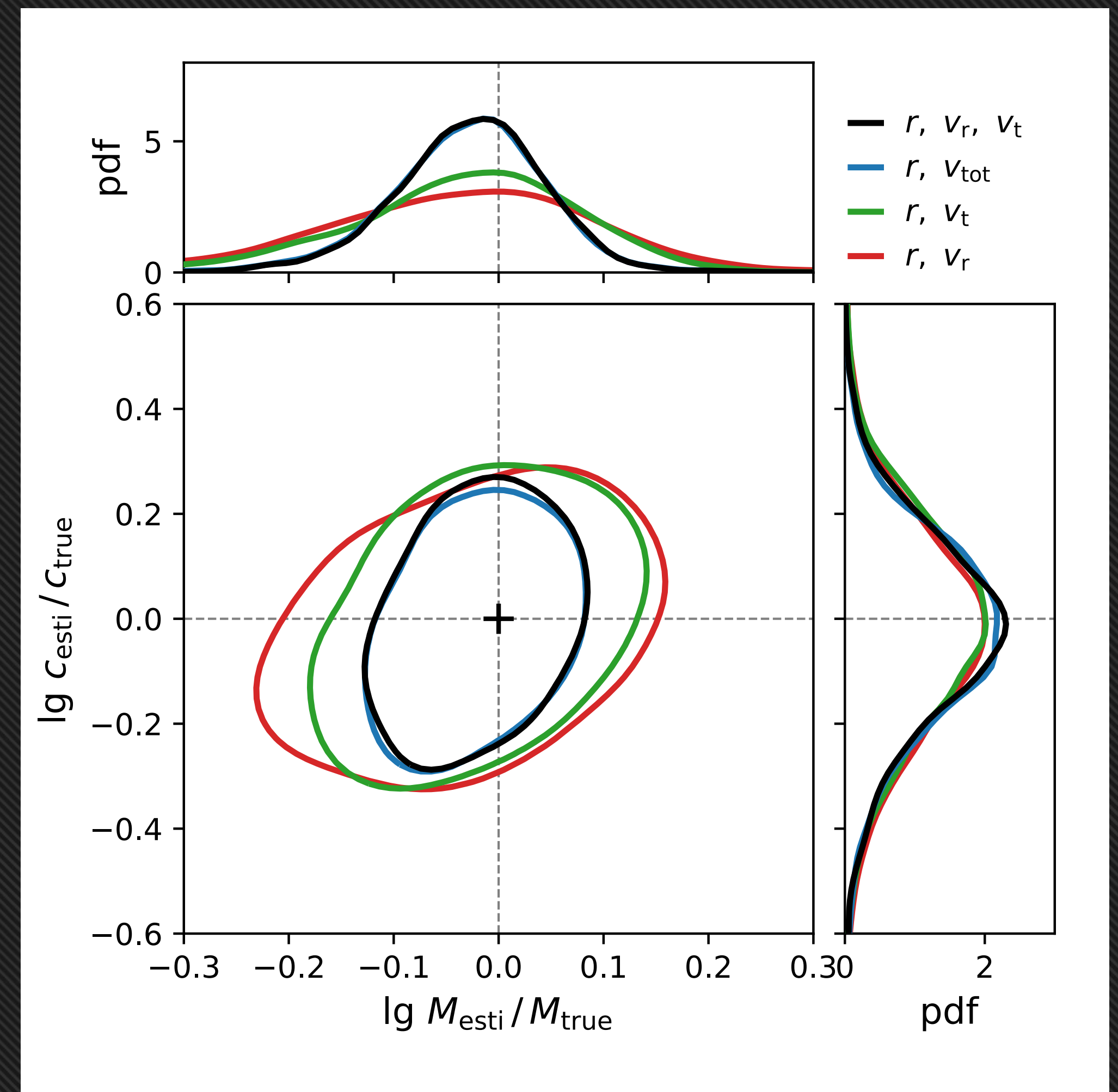
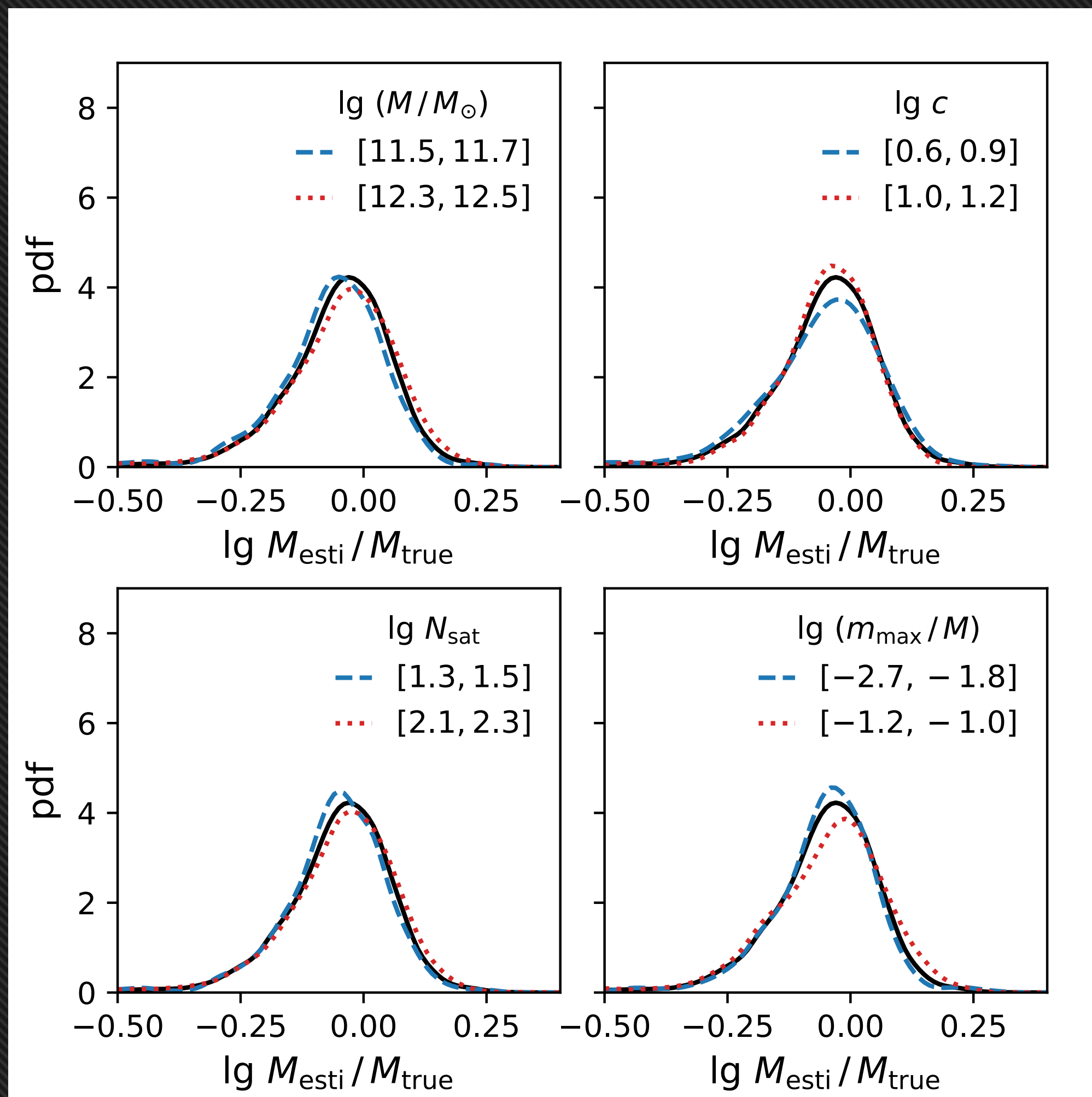


Varying satellite sample selection



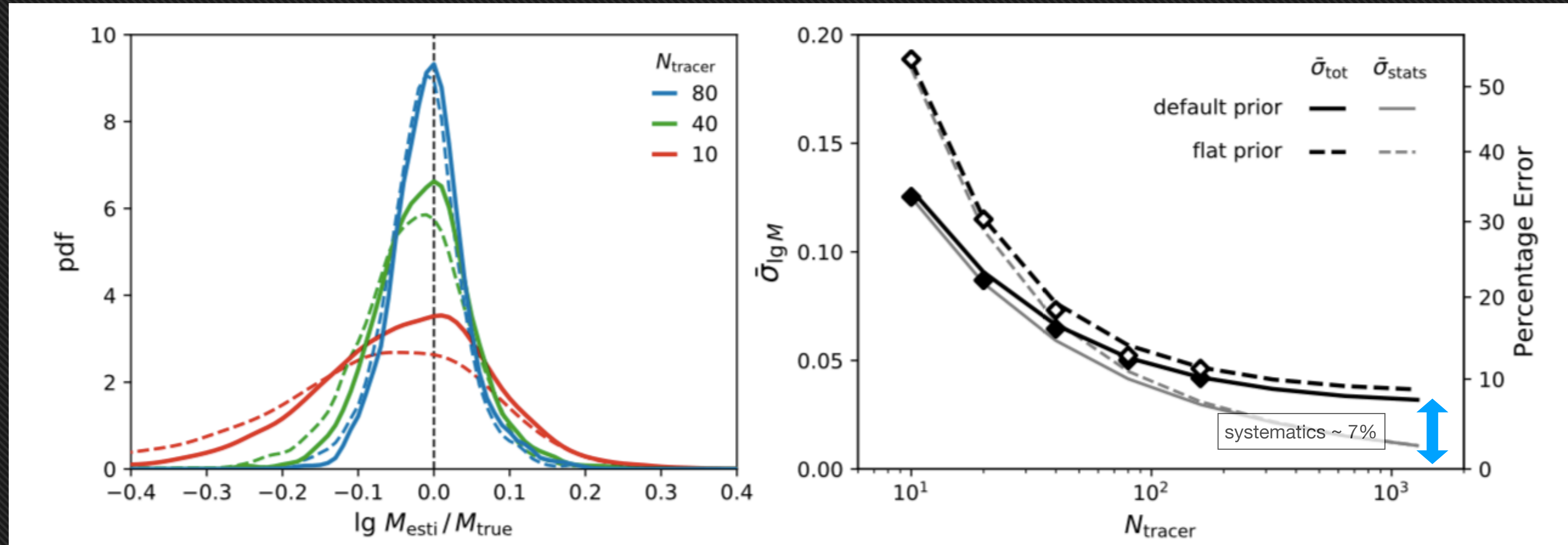
The mass estimation is robust against various sample selection criteria





General performance with mock sample

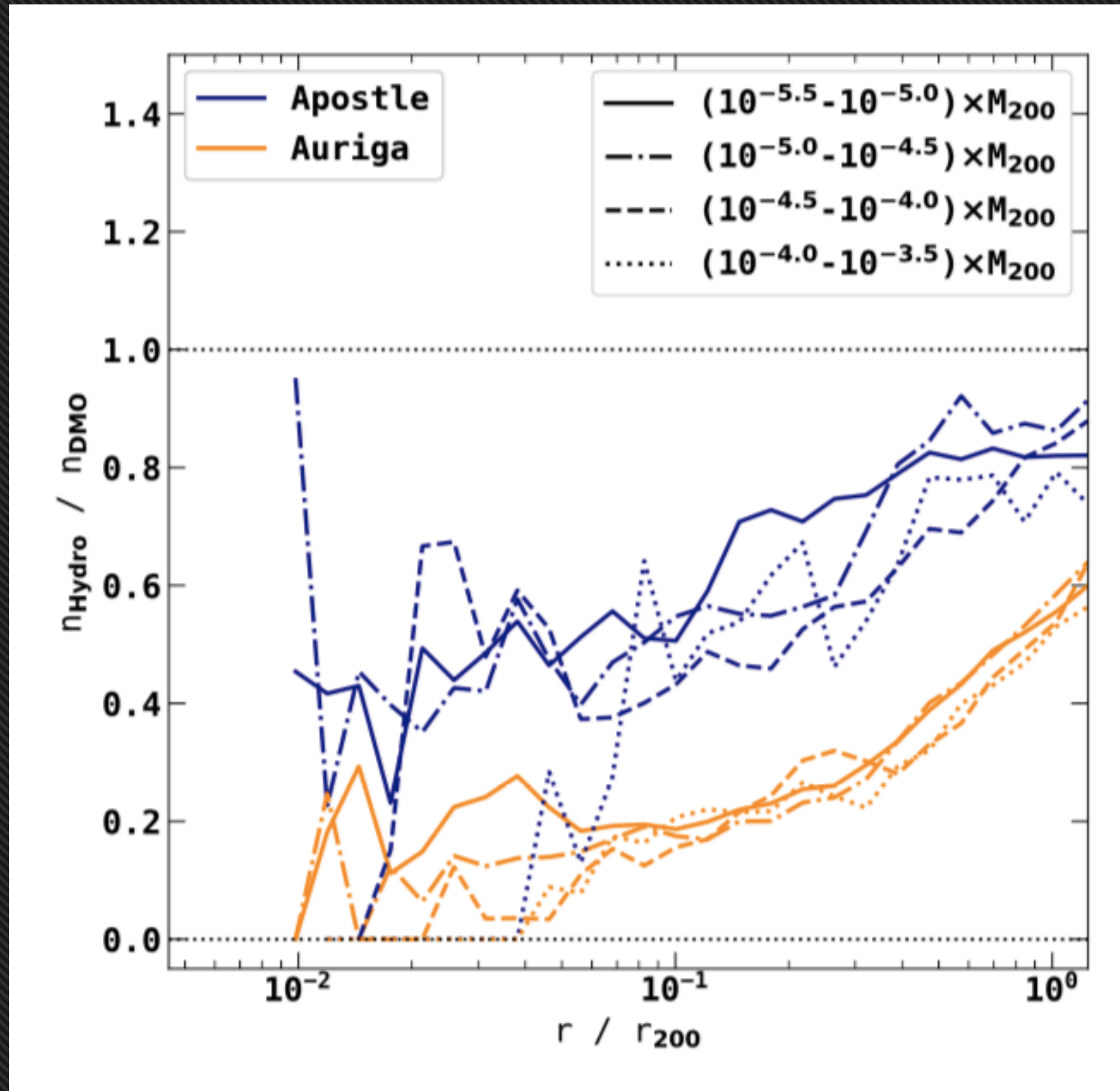
Varying the number of tracers



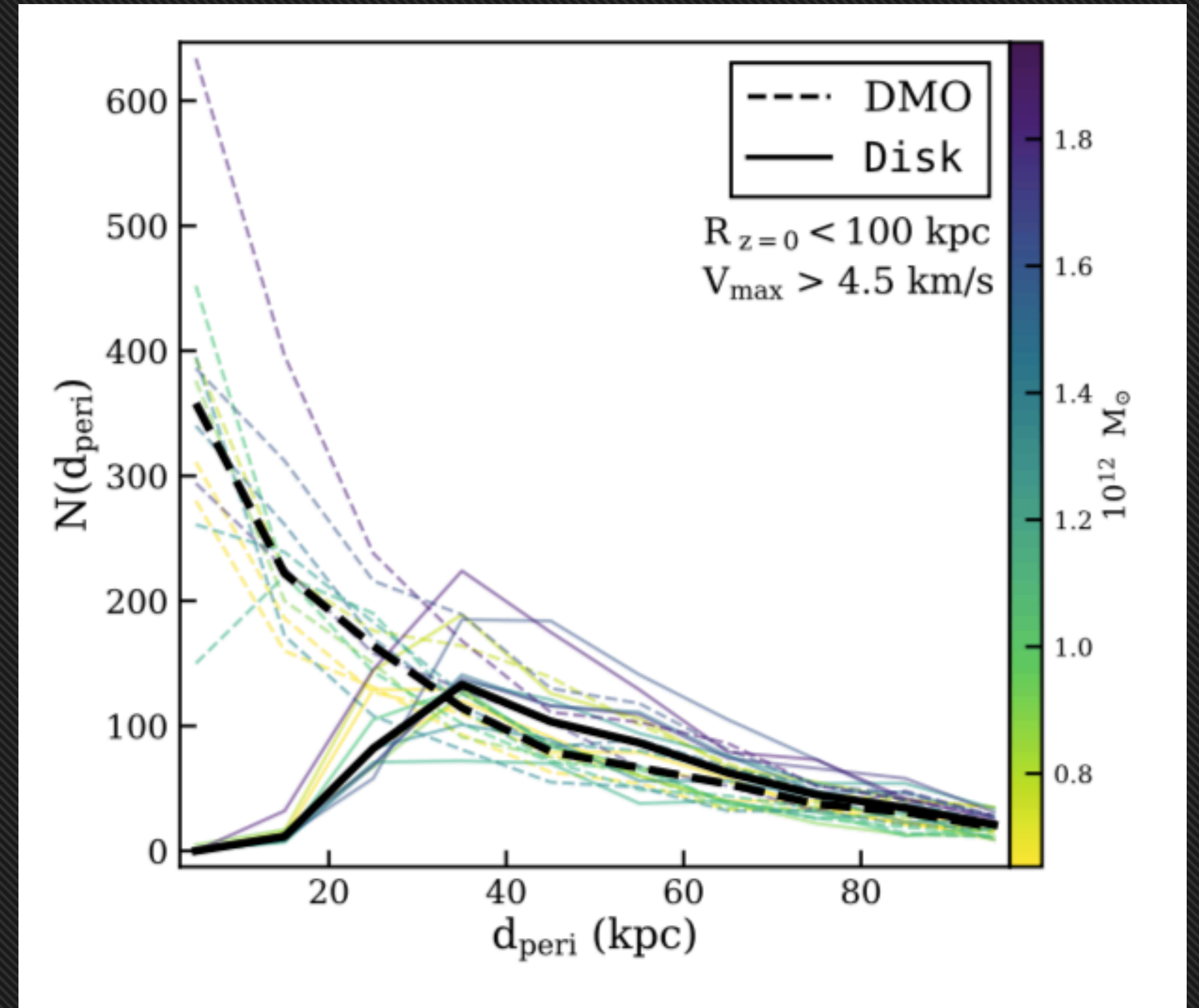
Mass

- Unbiased
- Low systematics (<10%)
negligible comparing to current
statistical uncertainty level

Enhanced satellite disruption due to stellar disc in hydro simulations



Richings 2018



Garrison-Kimmel 2017, Kelley 2018