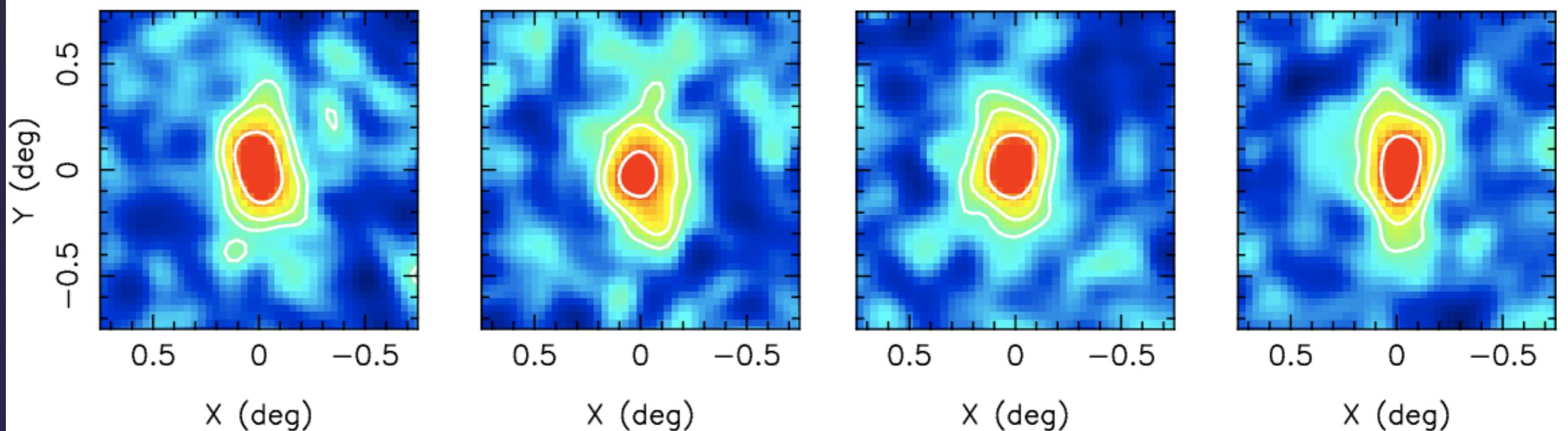


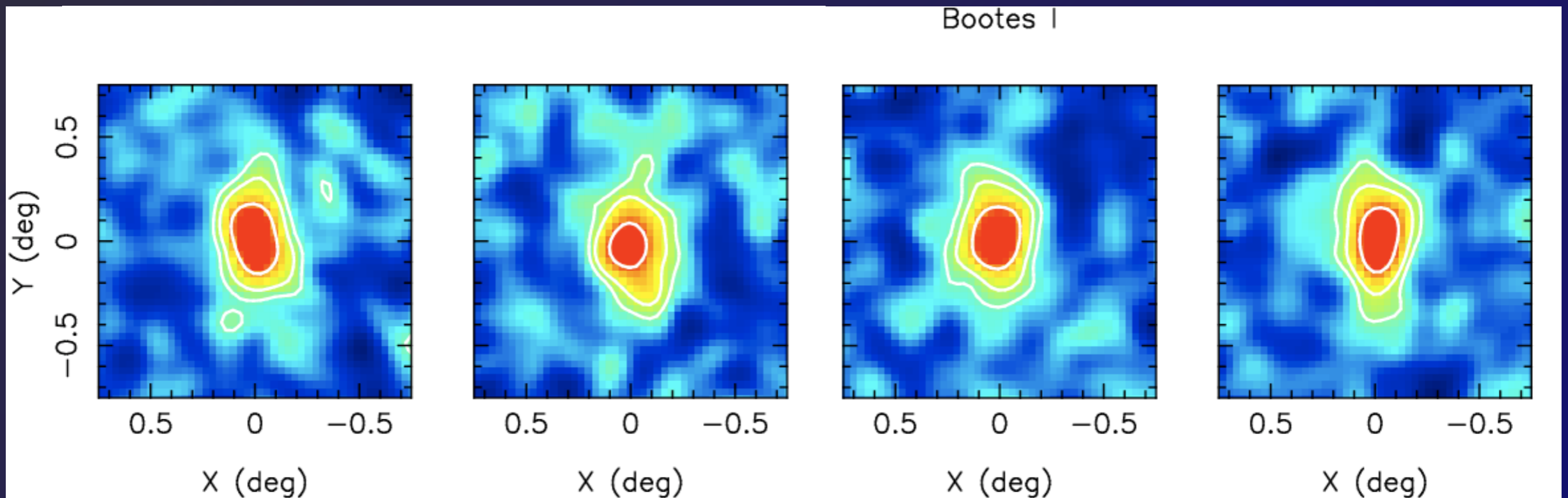
MUSINGS ON FAINT DWARF GALAXIES



Nicolas Martin, Jelte de Jong & Hans-Walter Rix (MPIA)

(2008, ApJ 684, 1075)

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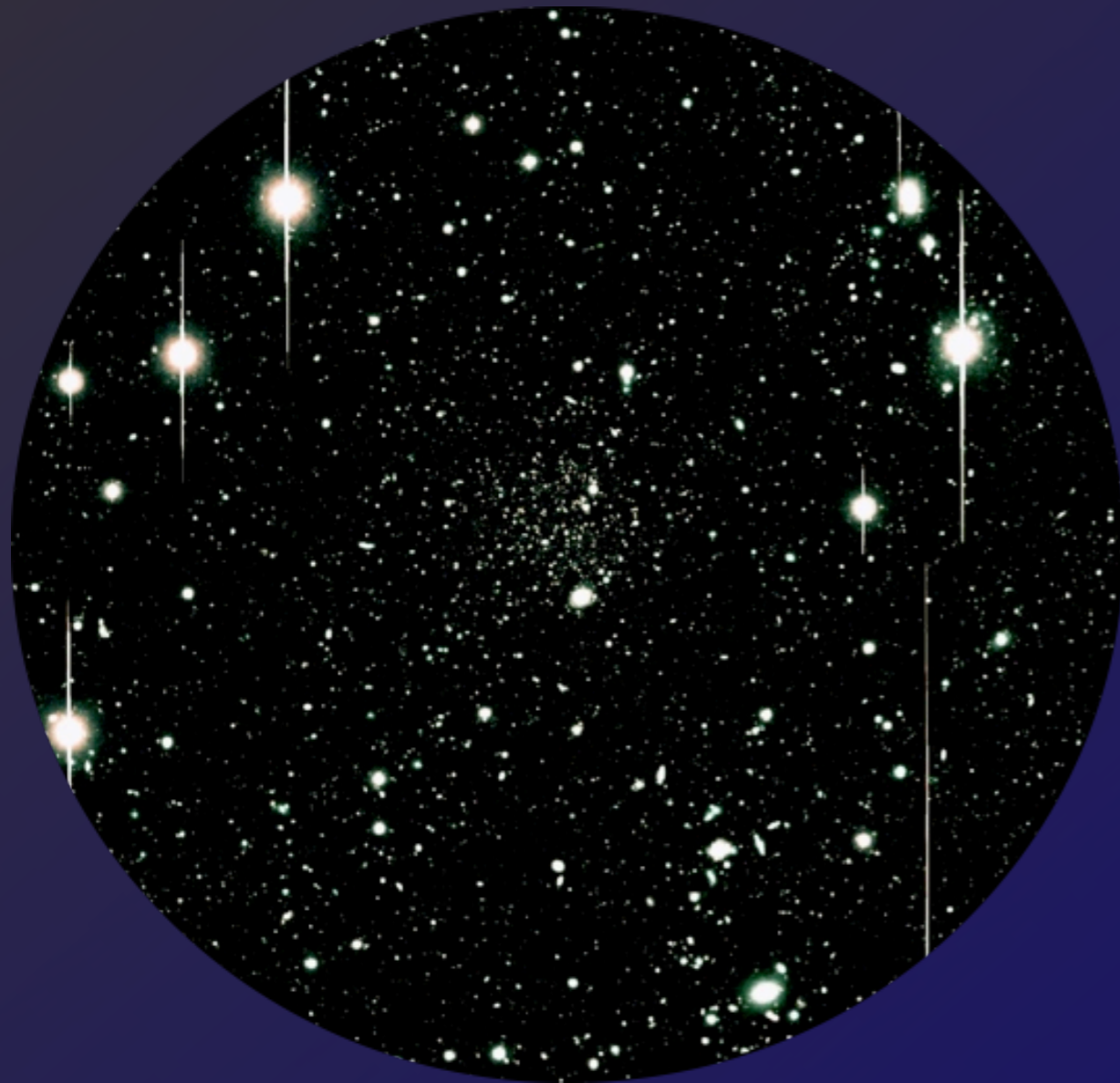
FAINT LOCAL GROUP SATELLITES

Number of **large LG satellites** more than **doubled**

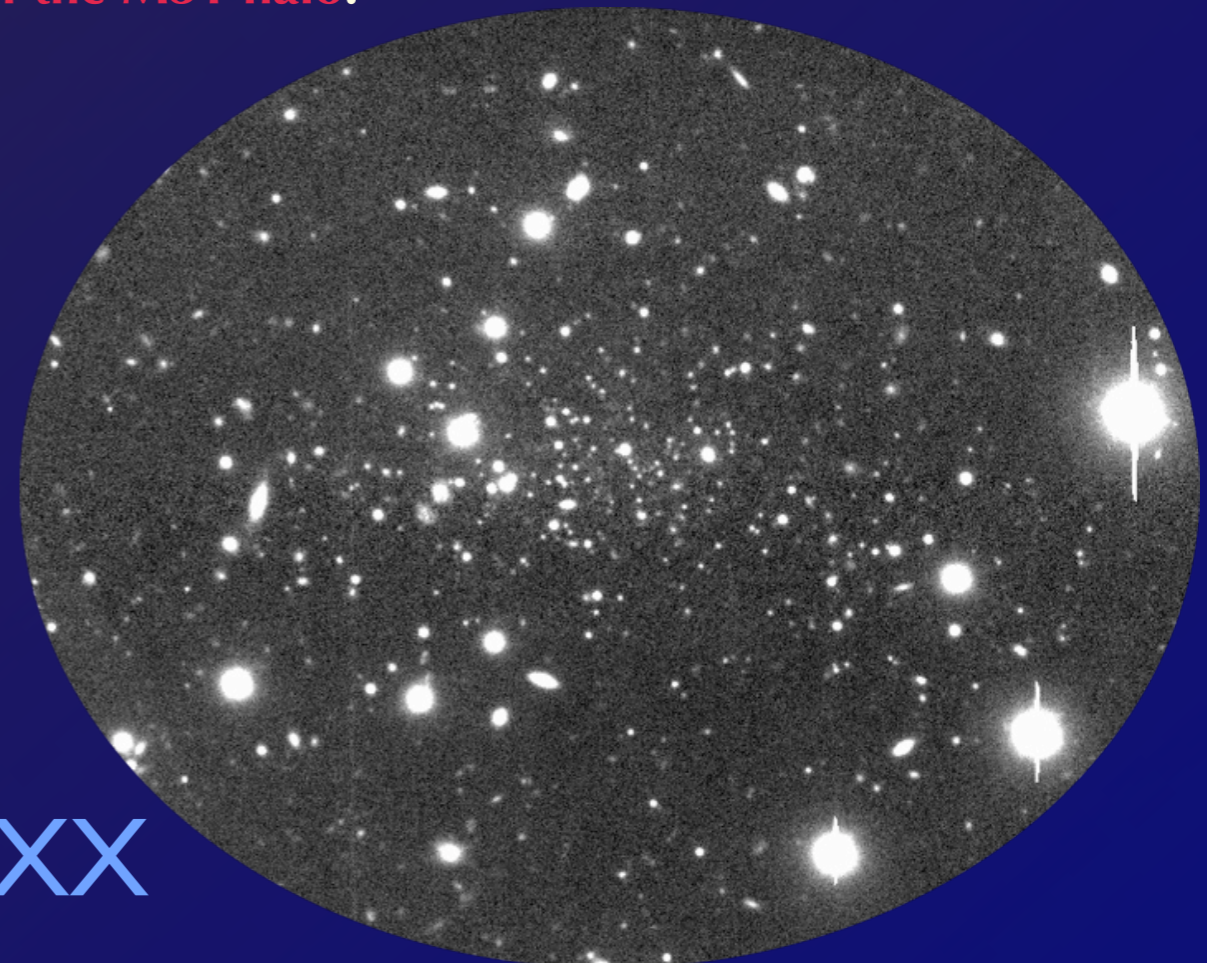
- **MW:** Boötes I (-6.4), Boötes II (-4.2), Canes Venatici I (-8.6), Canes Venatici II (-4.9), Coma Berenices (-4.1), Hercules (-6.6), Leo IV (-5.0), Leo V (-4.3), Leo T (-8.0), SDSSJ1058+2843 (-0.4), Segue 1 (-1.5), Ursa Major I (-5.4), Ursa Major II (-4.1), Willman 1 (-2.6) found in the **SDSS** (**quarter** of the MW halo)

[$M_V = -8.8$ for Draco, previously known faintest MW satellite]

- **M31:** And IX (-8.3), And X (-8.1), And XI (-7.3), And XII (-6.4), And XIII (-6.9), And XIV (-8.5), And XV (-9.4), And XVI (-9.2), And XVII (-8.5), And XVIII (-9.1), And XIX (-9.7), And XX (-6.5) & ... from **large surveys** of roughly a **third of the M31 halo**.

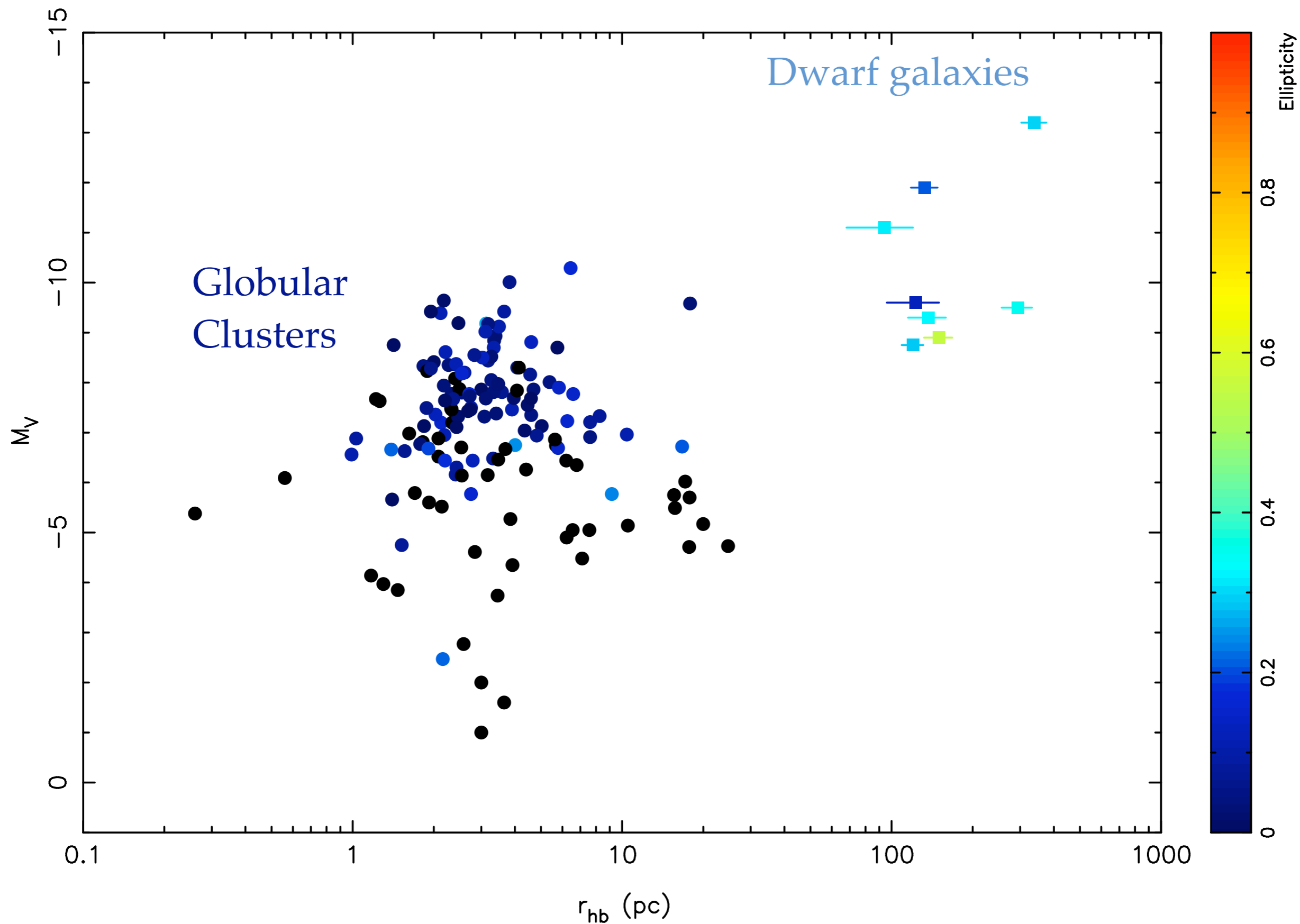


Leo T

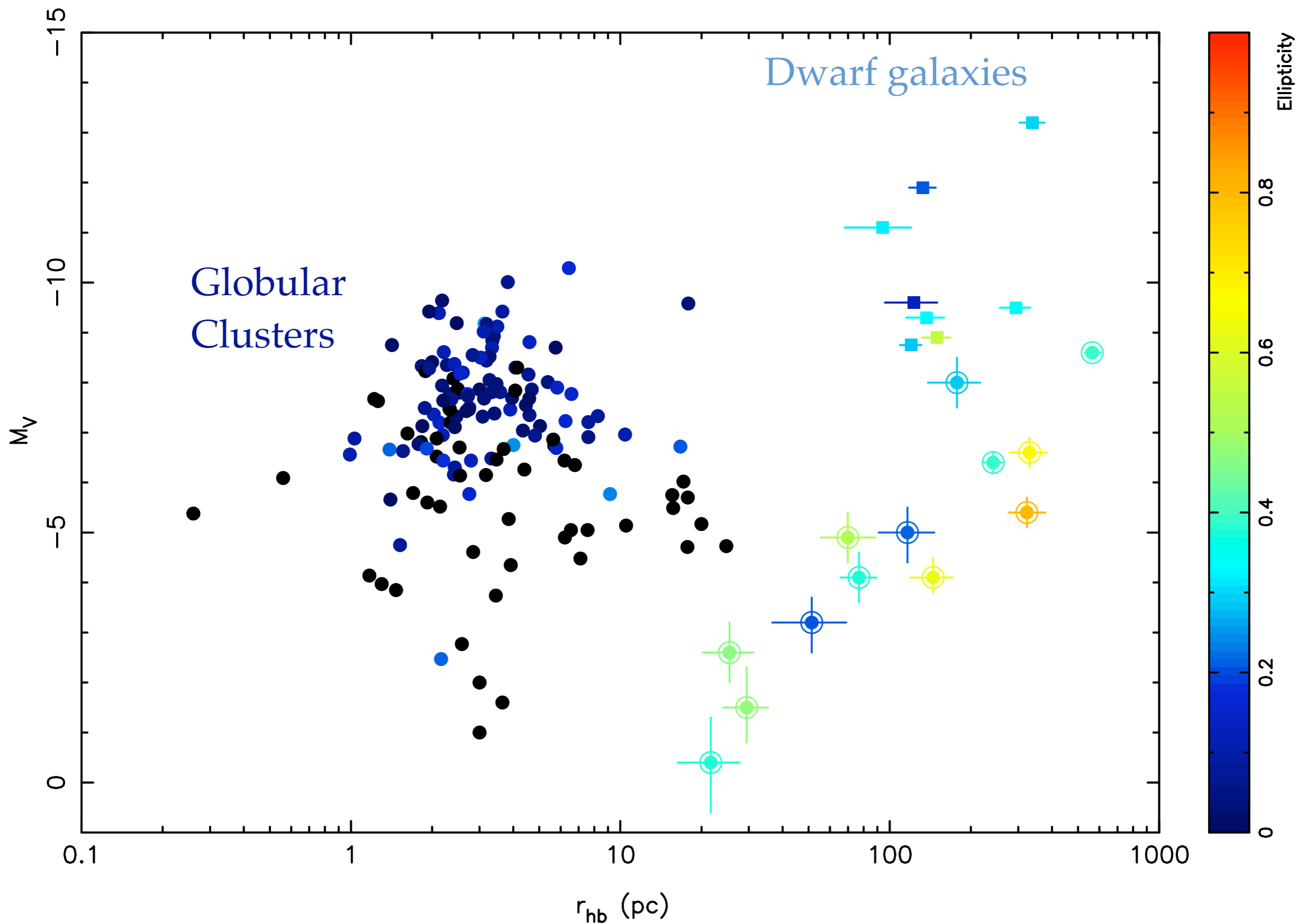


And XX

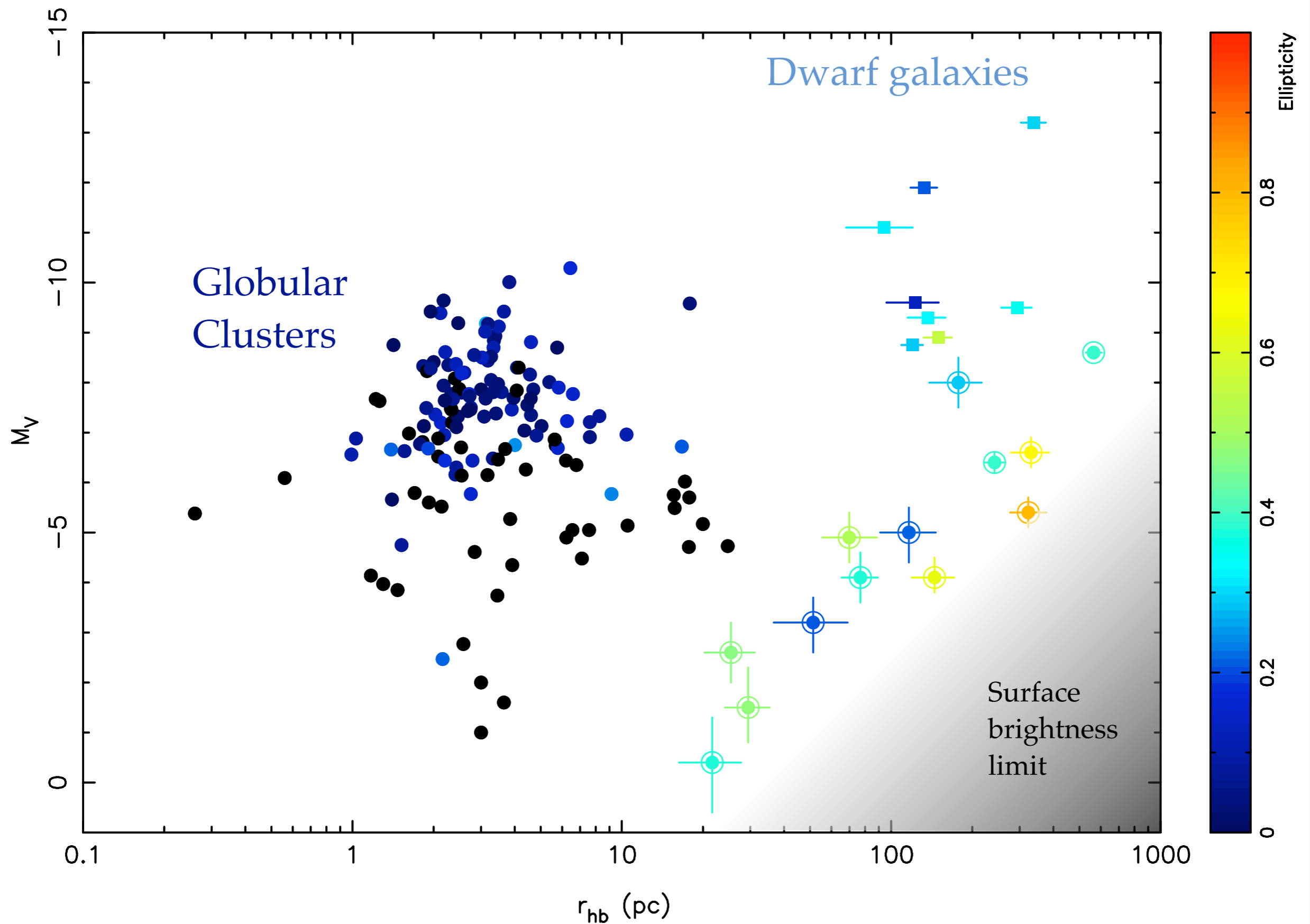
MILKY WAY SATELLITE SYSTEM



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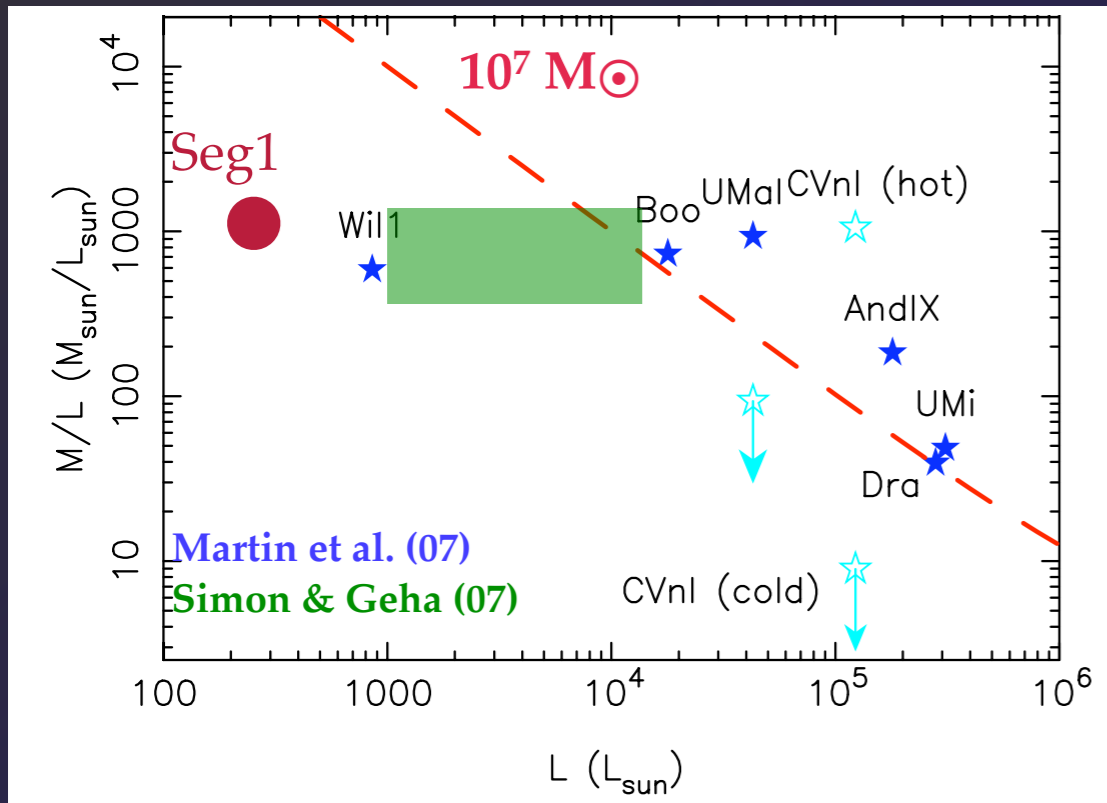


FAINT LOCAL GROUP SATELLITES

Number of **large LG satellites** more than **doubled**

Around both MW and M31

What are they?



- Are they faint dwarf galaxies? Some overlap with previously known galaxies (especially around M31).
- **DM dominated** from spectroscopic surveys (Mass estimates)
- **Why smaller?** (completeness)

FAINT LOCAL GROUP SATELLITES

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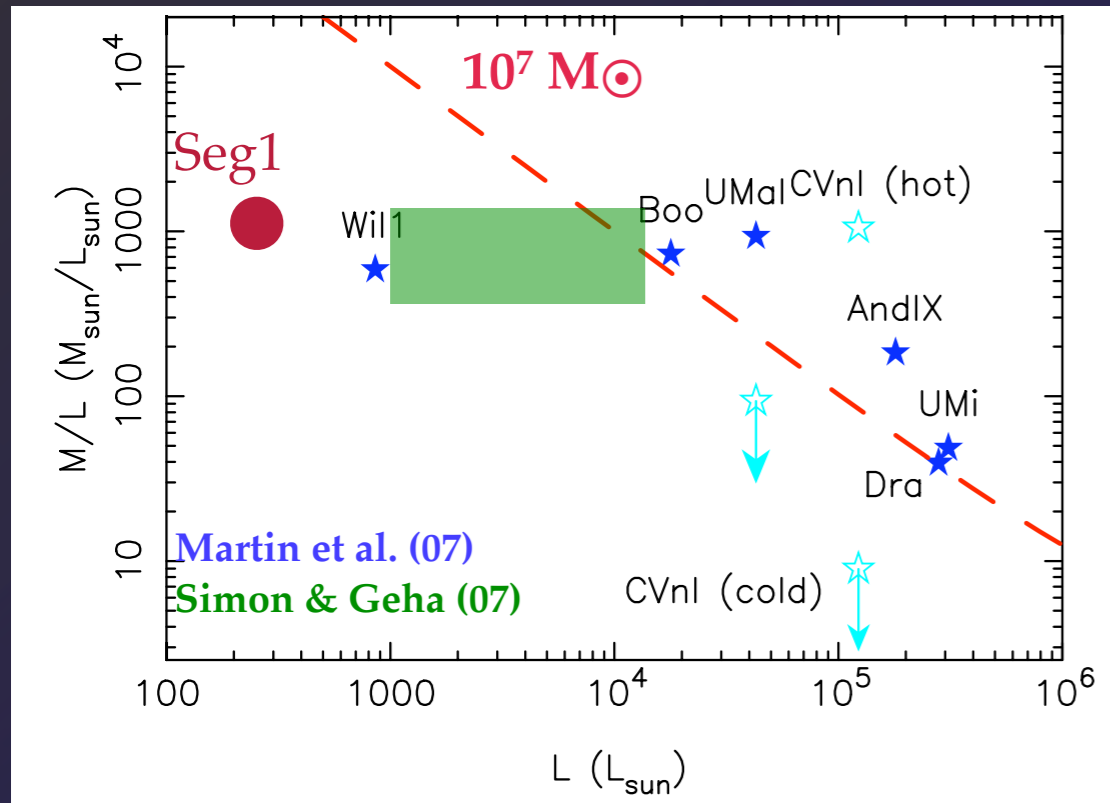
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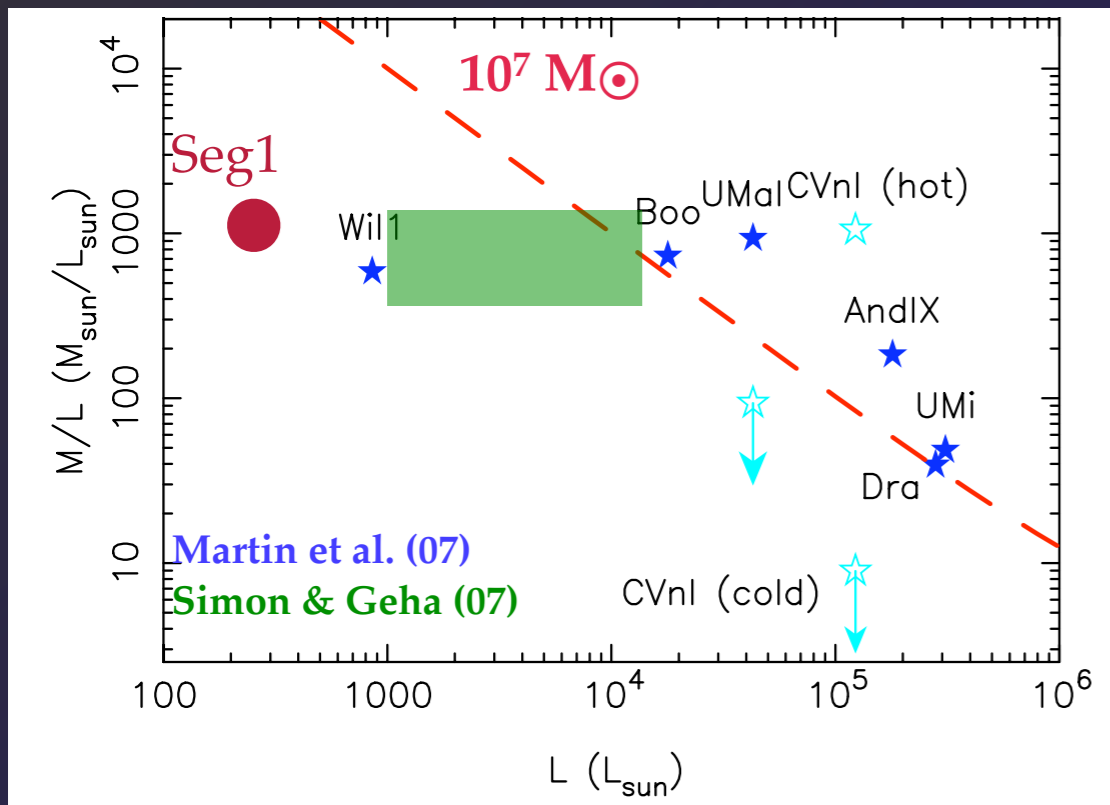
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➔ **What are their properties?**

- Very low number of stars (30 – 400 stars in SDSS)
- **Reliable structural parameters?** ($\alpha_0, \delta_0, \theta, \epsilon, r_h$) + $N^* \rightarrow M_V$? Are they really distorted?



FAINT LOCAL GROUP SATELLITES



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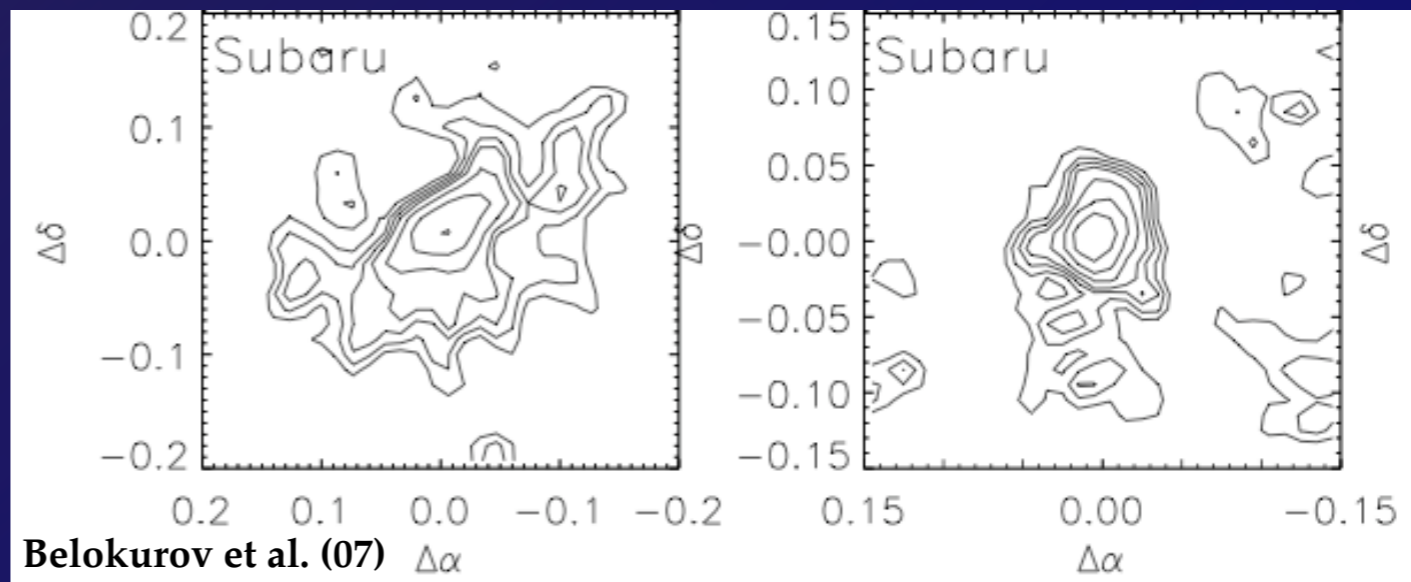
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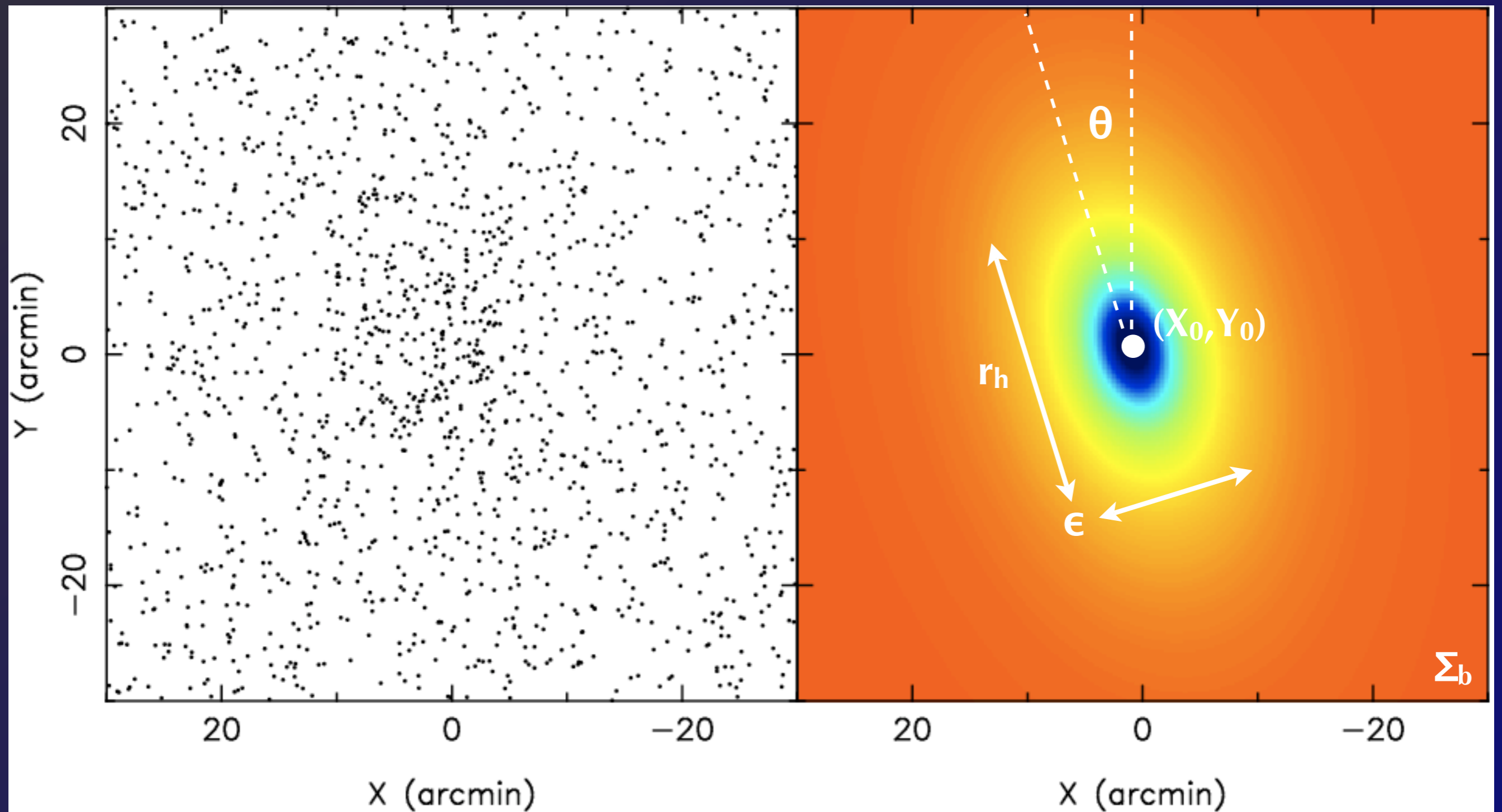


STRUCTURAL PROPERTIES?

Martin, de Jong & Rix (2008b)

From SDSS data, **homogeneous structural parameters and properties:**

- Best model with exponential density profile



FITTING STRUCTURAL PROPERTIES

Previous structural parameter estimates **with assumptions:**

- **smoothing** (pixel size, smoothing kernel, background threshold)
- $\epsilon=0\dots$
- + $\Sigma(r)$ model

Maximum Likelihood, only $\Sigma(r)$ model:

$$\mathcal{L}(p_1, p_2, \dots, p_j) = \prod_i \ell_i(p_1, p_2, \dots, p_j)$$

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$$r_i = \left(\left(\frac{1}{1 - \epsilon} ((X_i - X_0) \cos \theta - (Y_i - Y_0) \sin \theta) \right)^2 + \left((X_i - X_0) \sin \theta + (Y_i - Y_0) \cos \theta \right)^2 \right)^{1/2}$$

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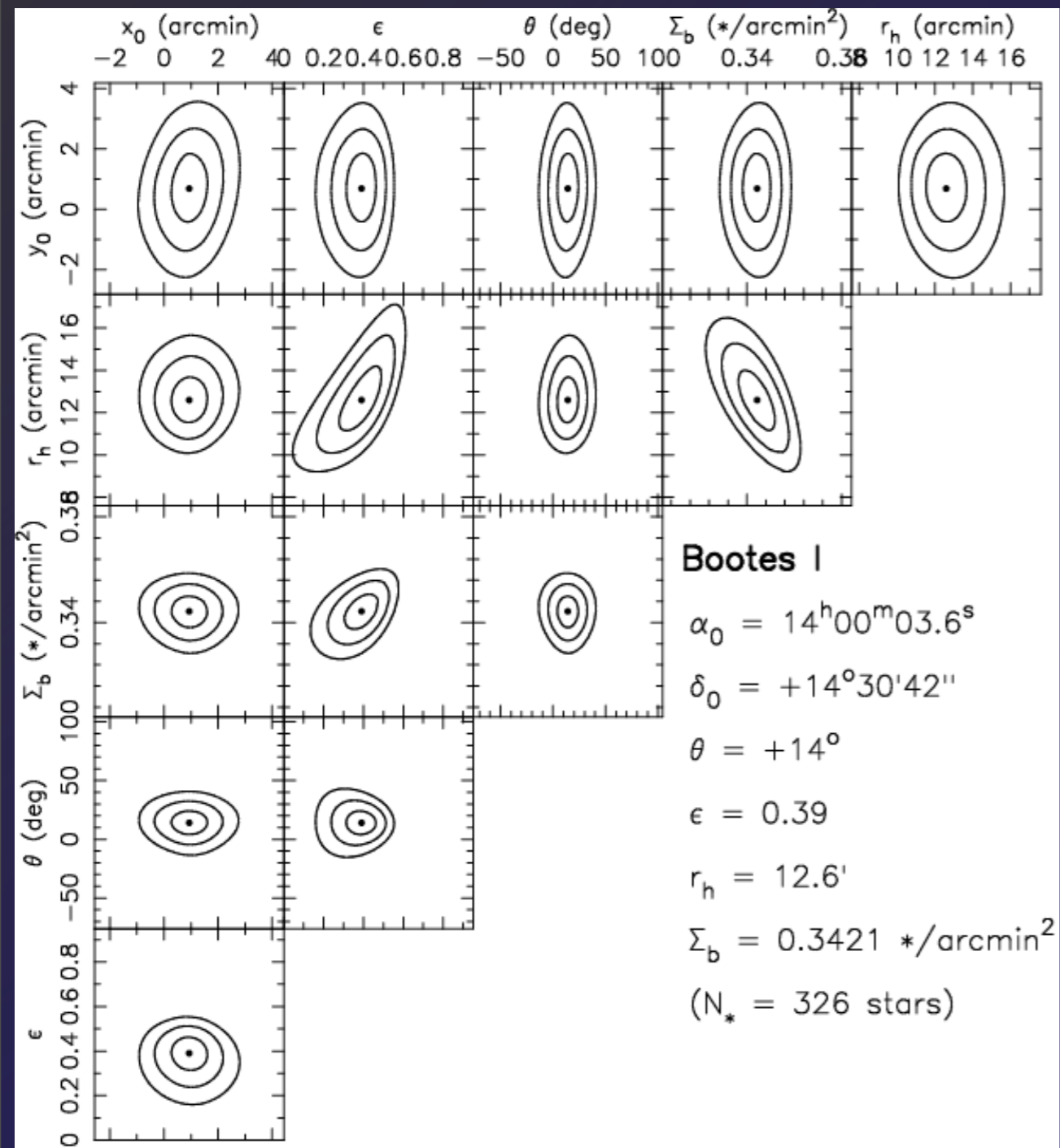
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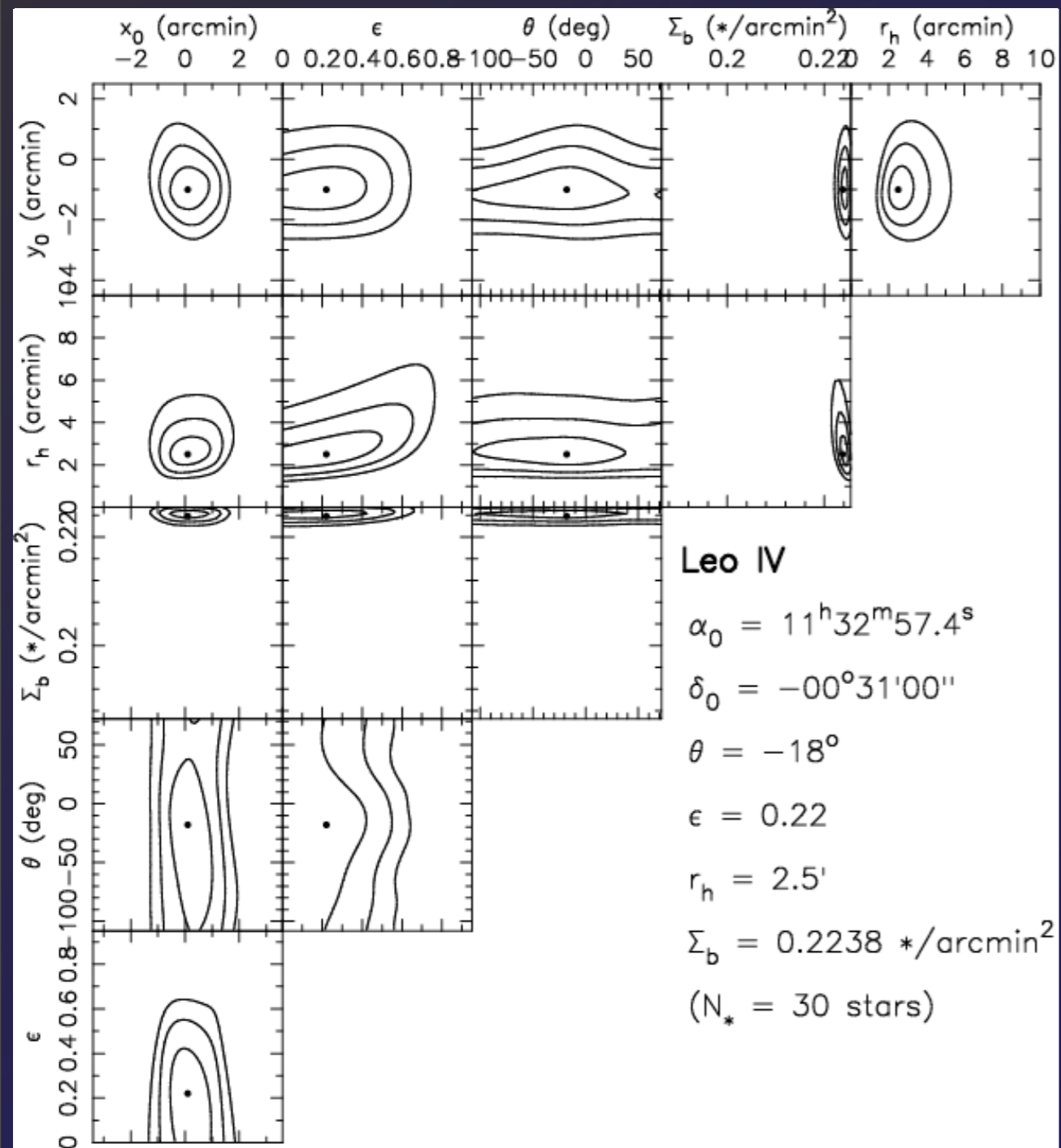
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For all stars i around a satellite, compute $\mathcal{L}(p_1, p_2, \dots, p_j)$ over a grid of
 (p_1, p_2, \dots, p_j)

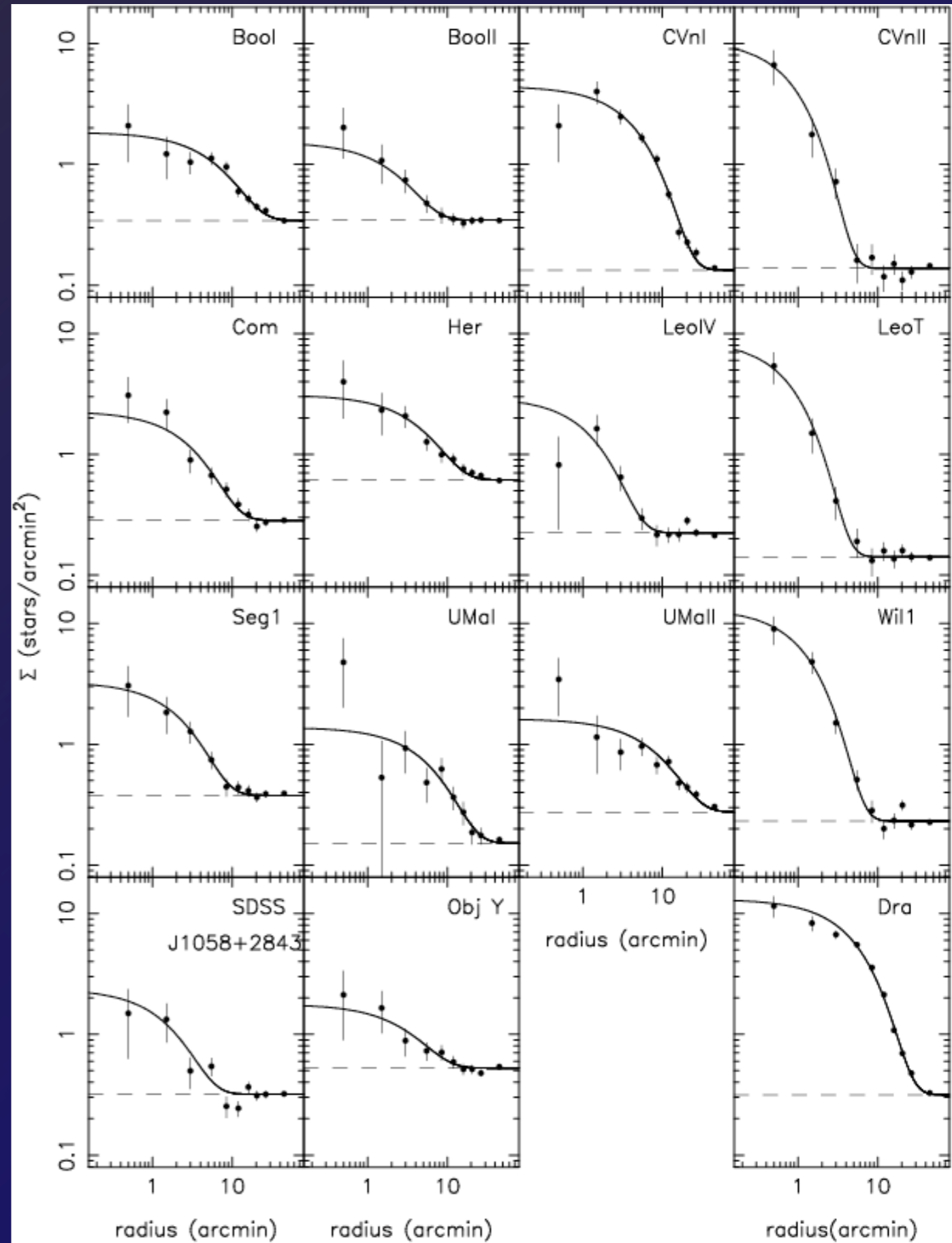
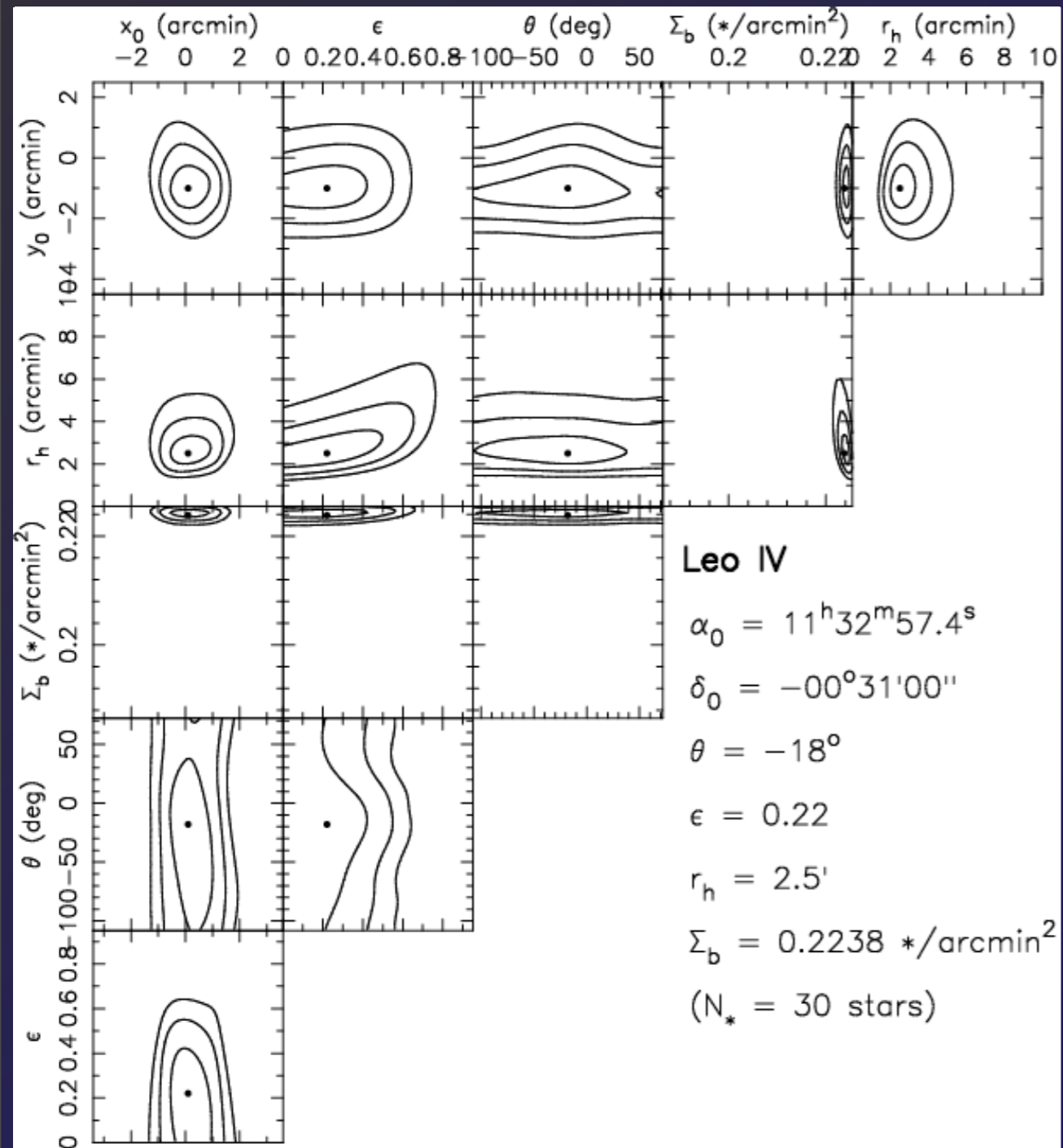
ML FITS



ML FITS



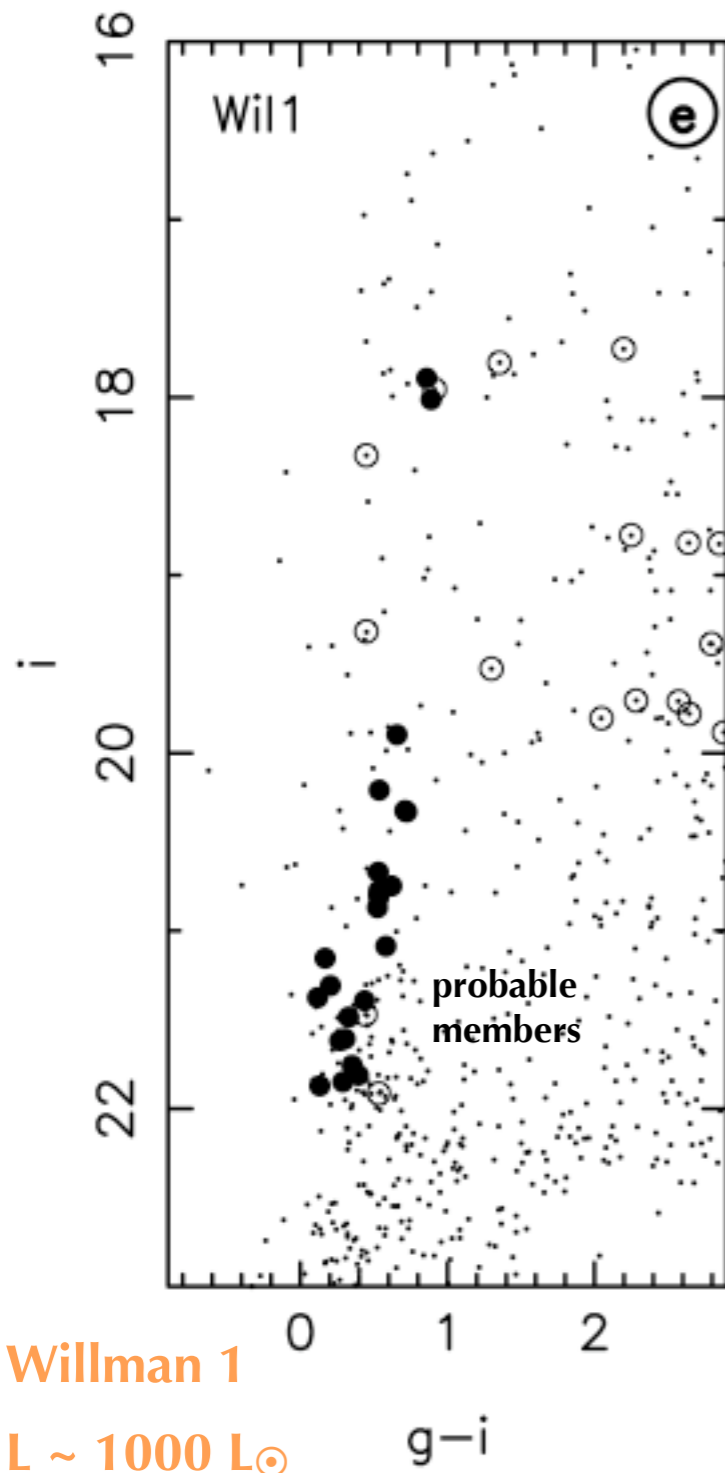
ML FITS



MAGNITUDES

M_V measured from member stars' luminosity suffers from 'CMD shot-noise'

ML gives N^* + stellar population models (de Jong et al. 07) from the same dataset → typical M_V



Willman 1

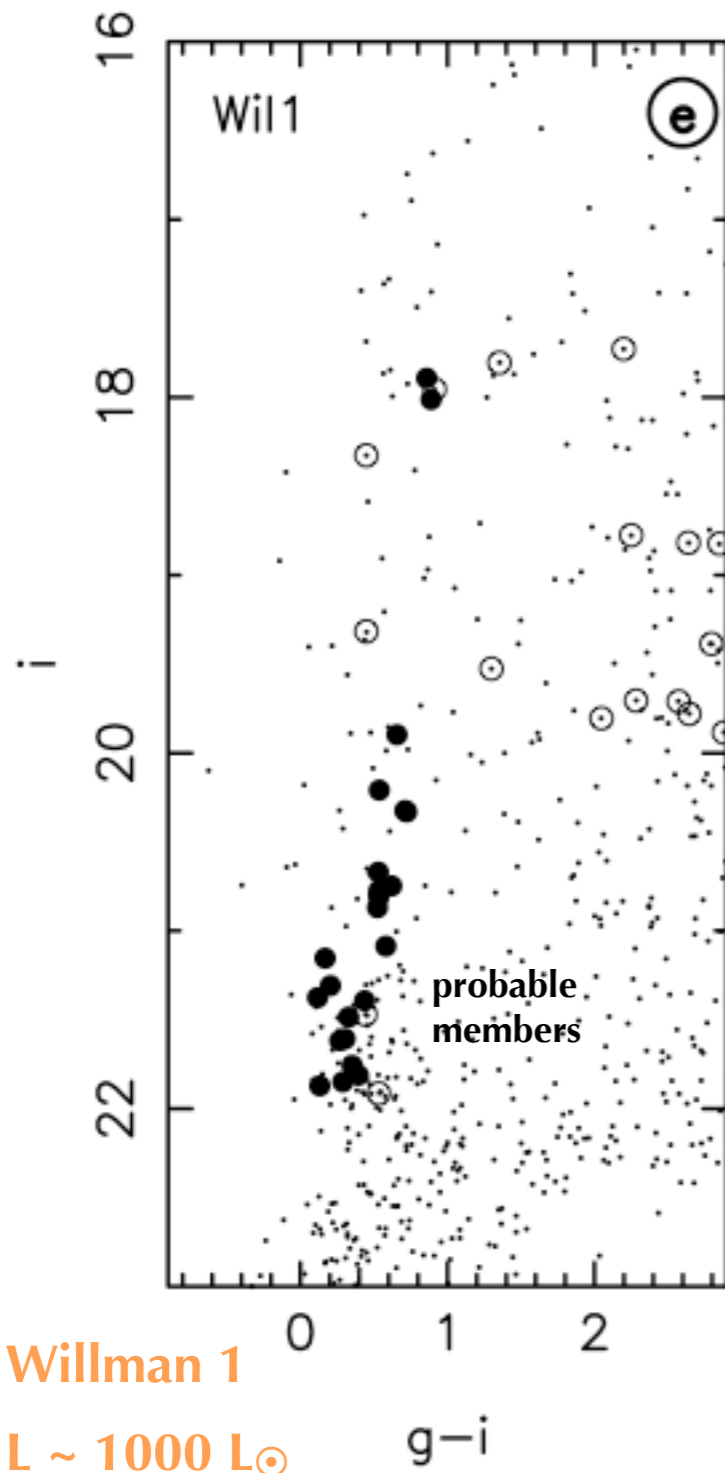
$L \sim 1000 L_{\odot}$

MAGNITUDES

tip RGB

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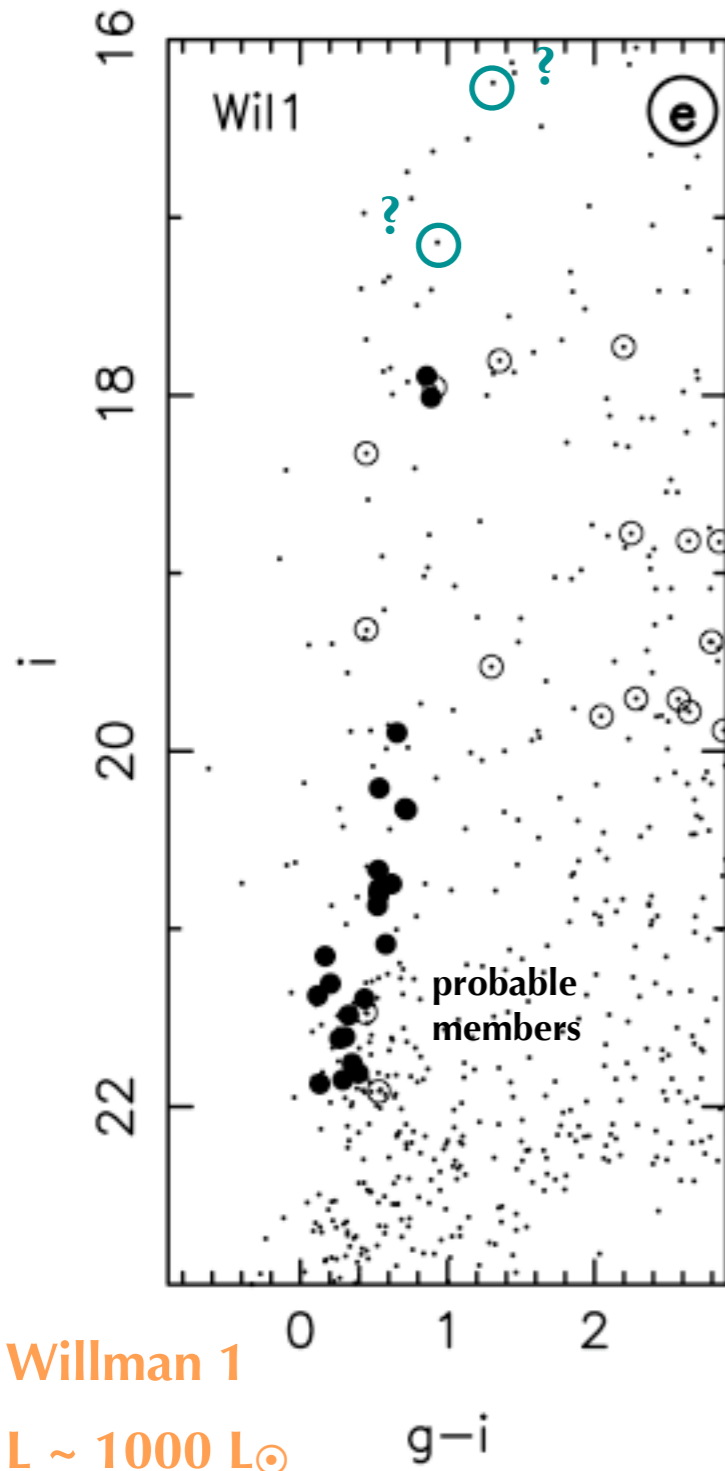
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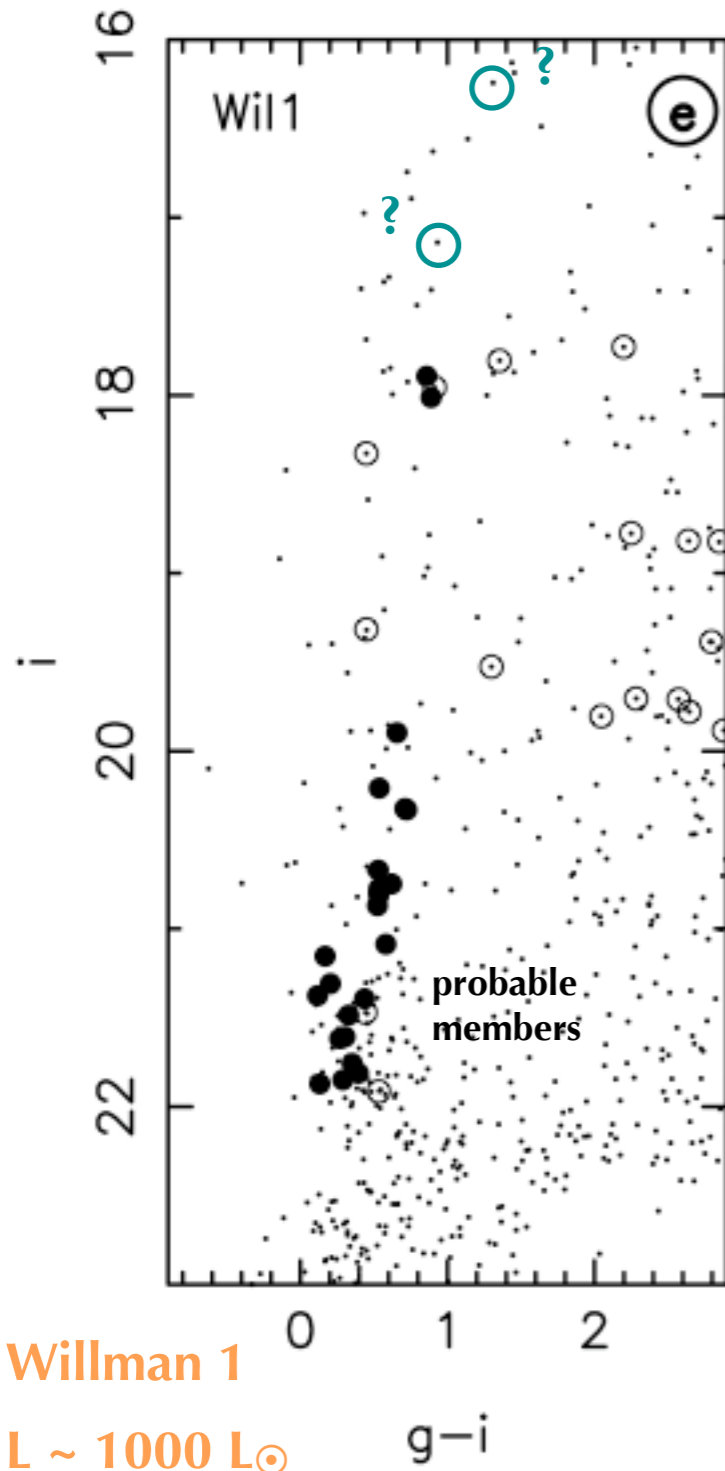
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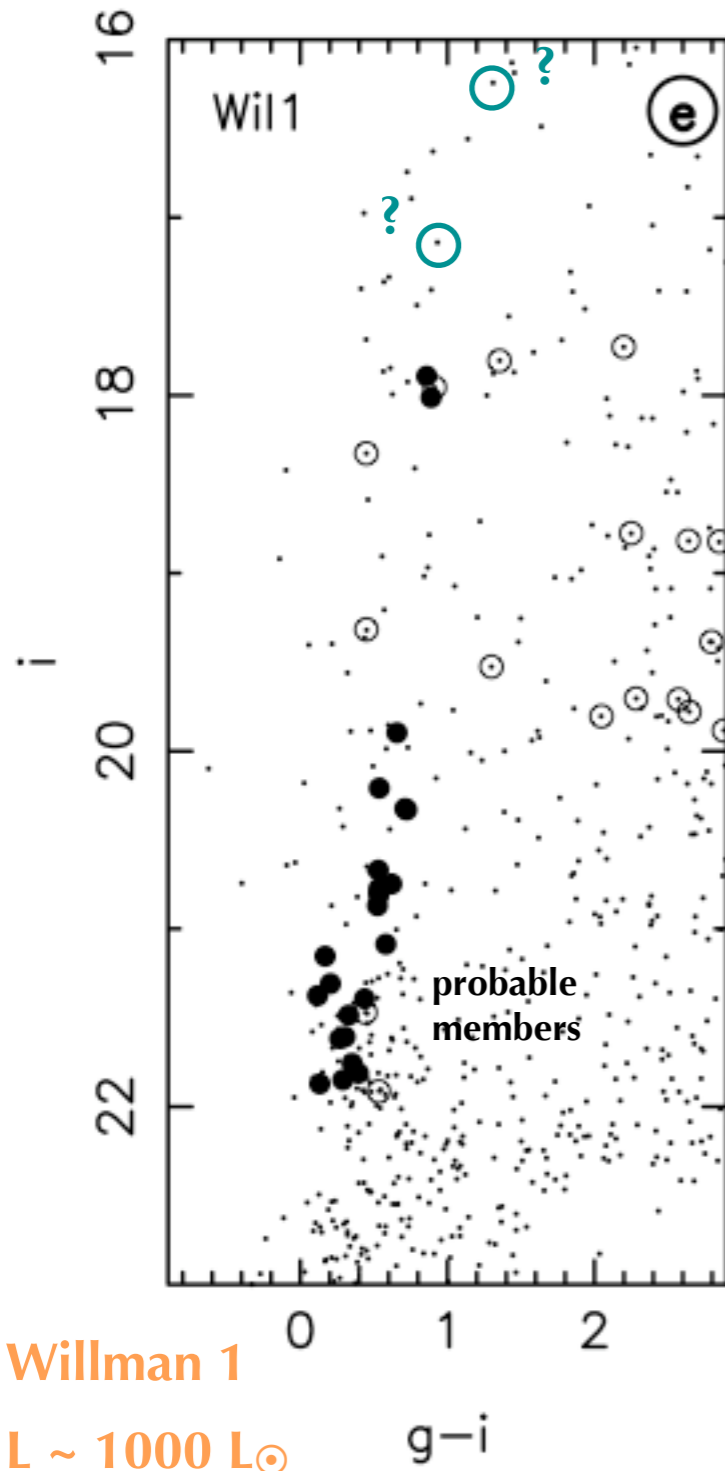
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10^3 times



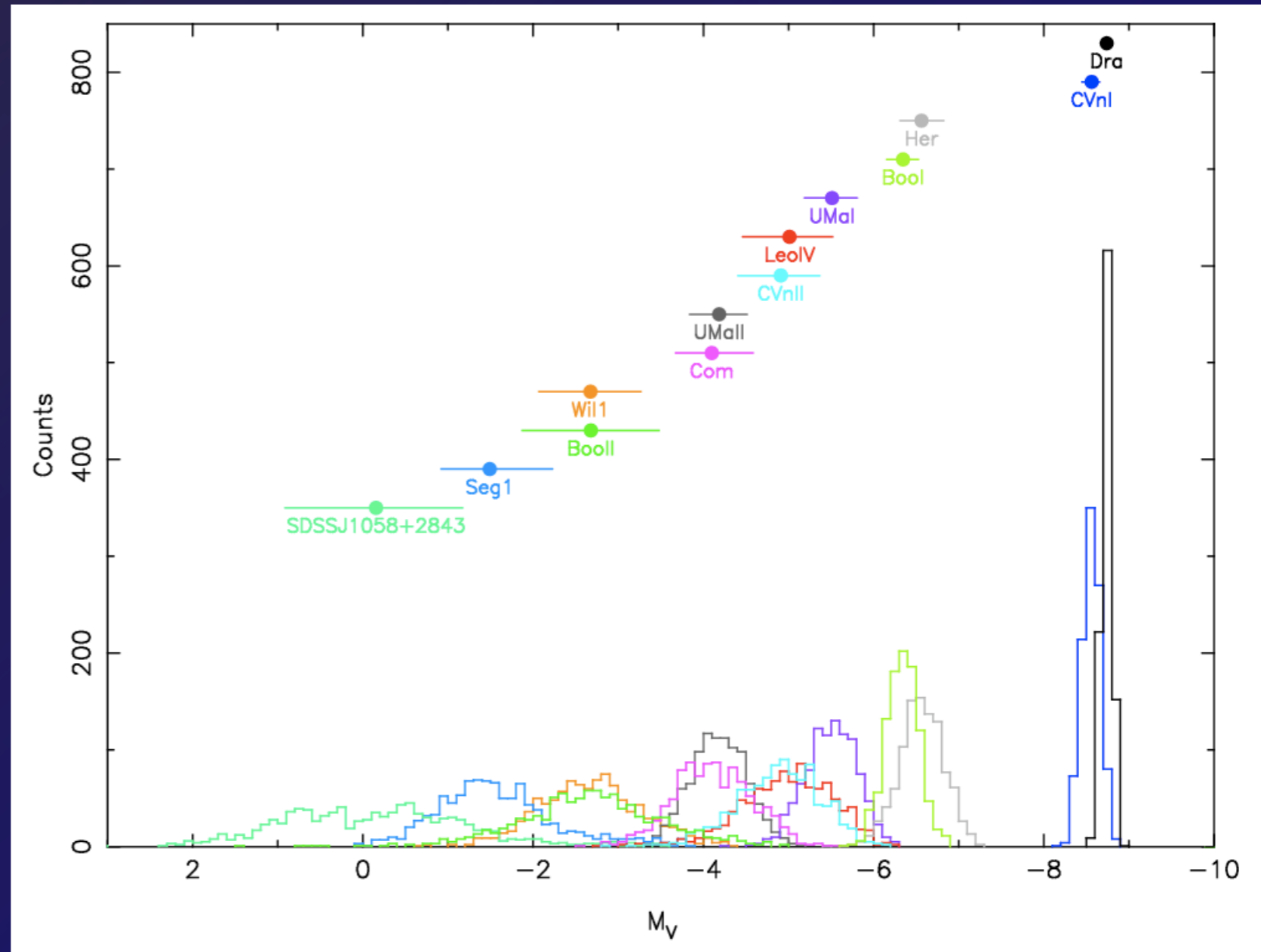
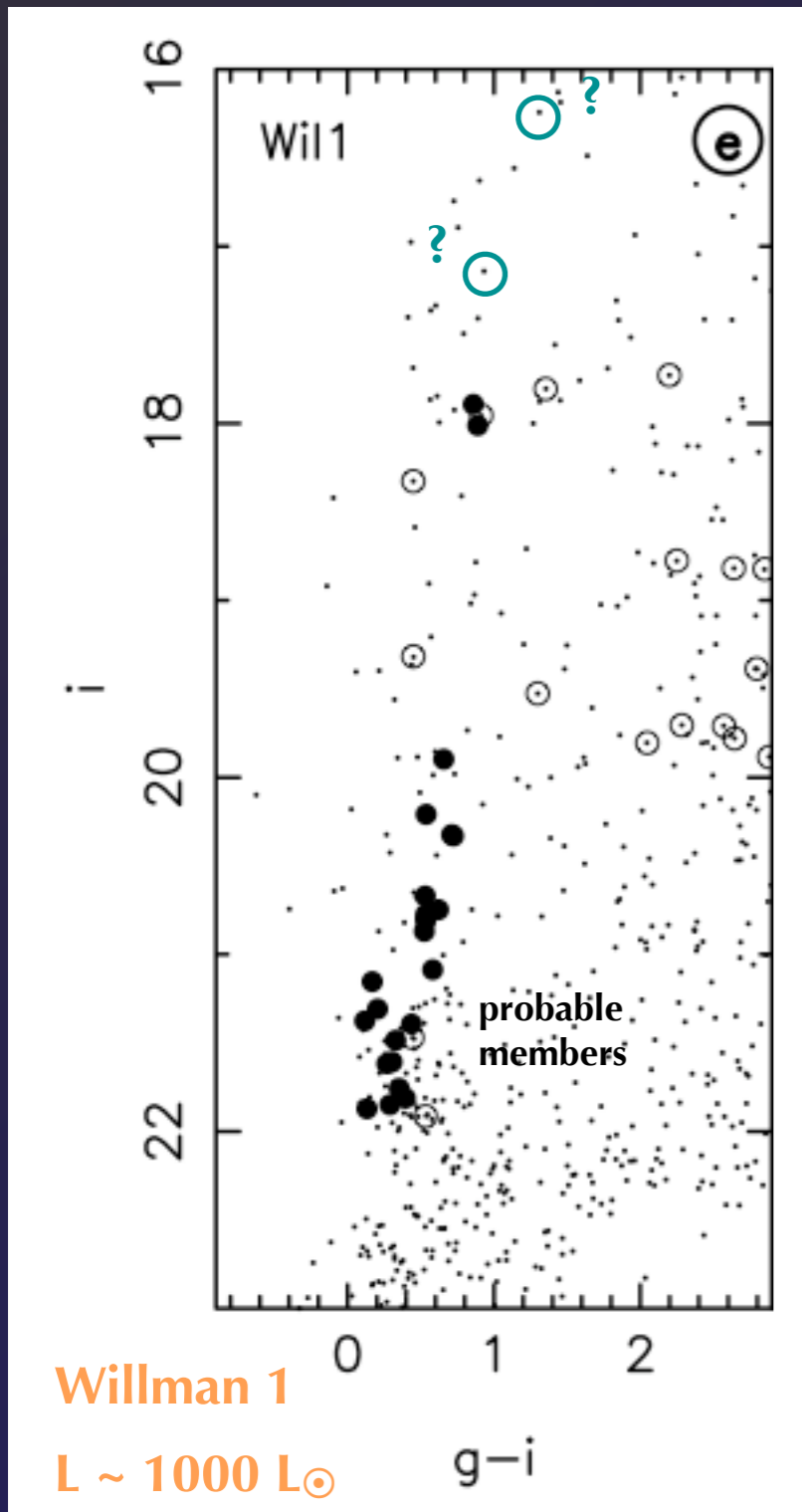
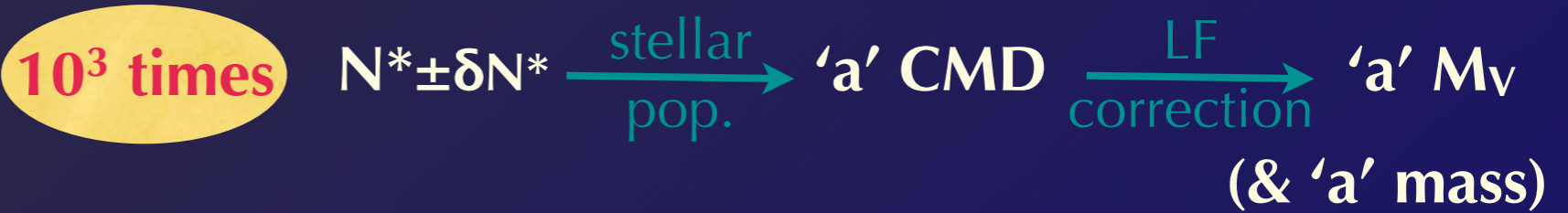
MAGNITUDES

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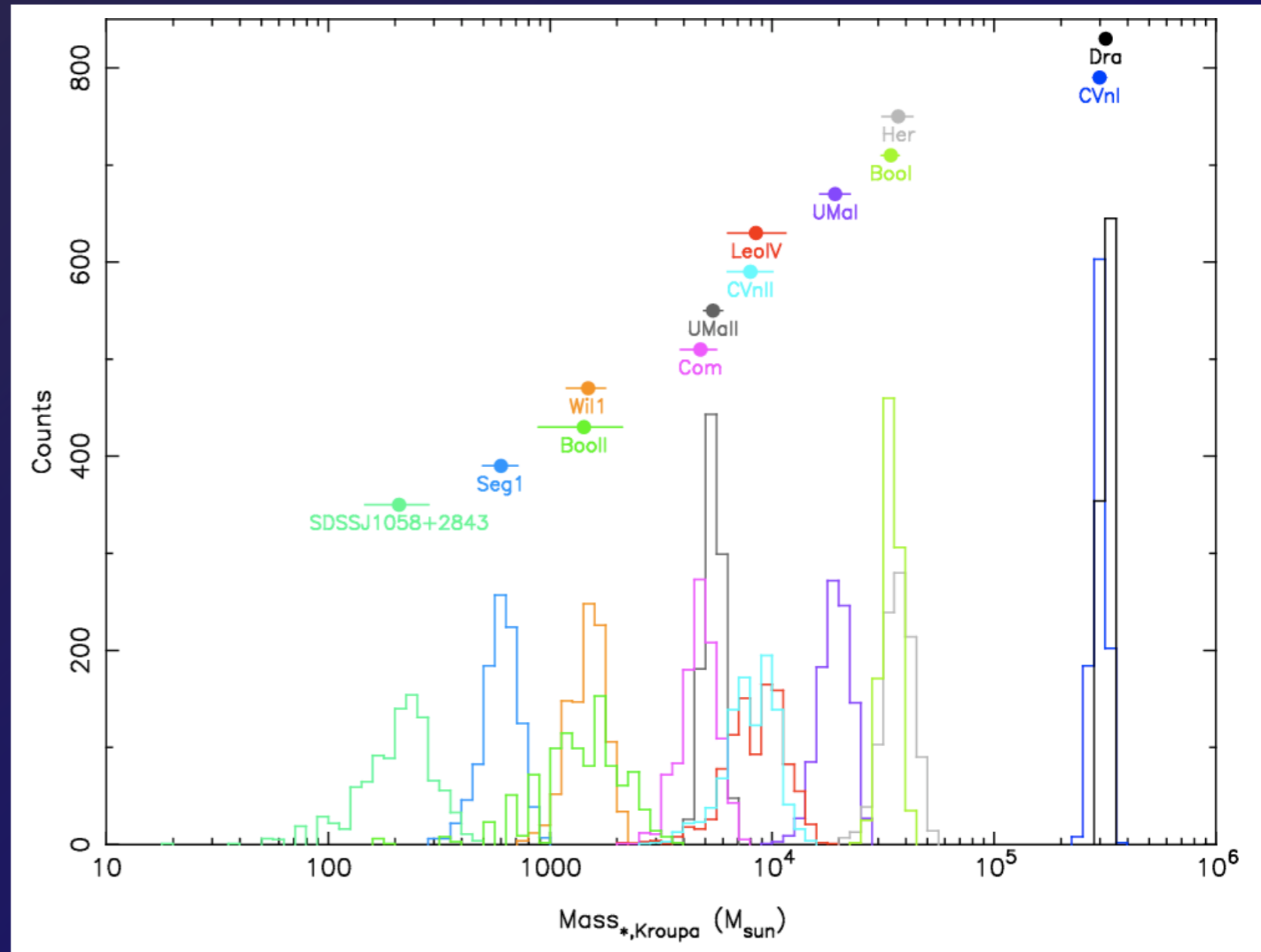
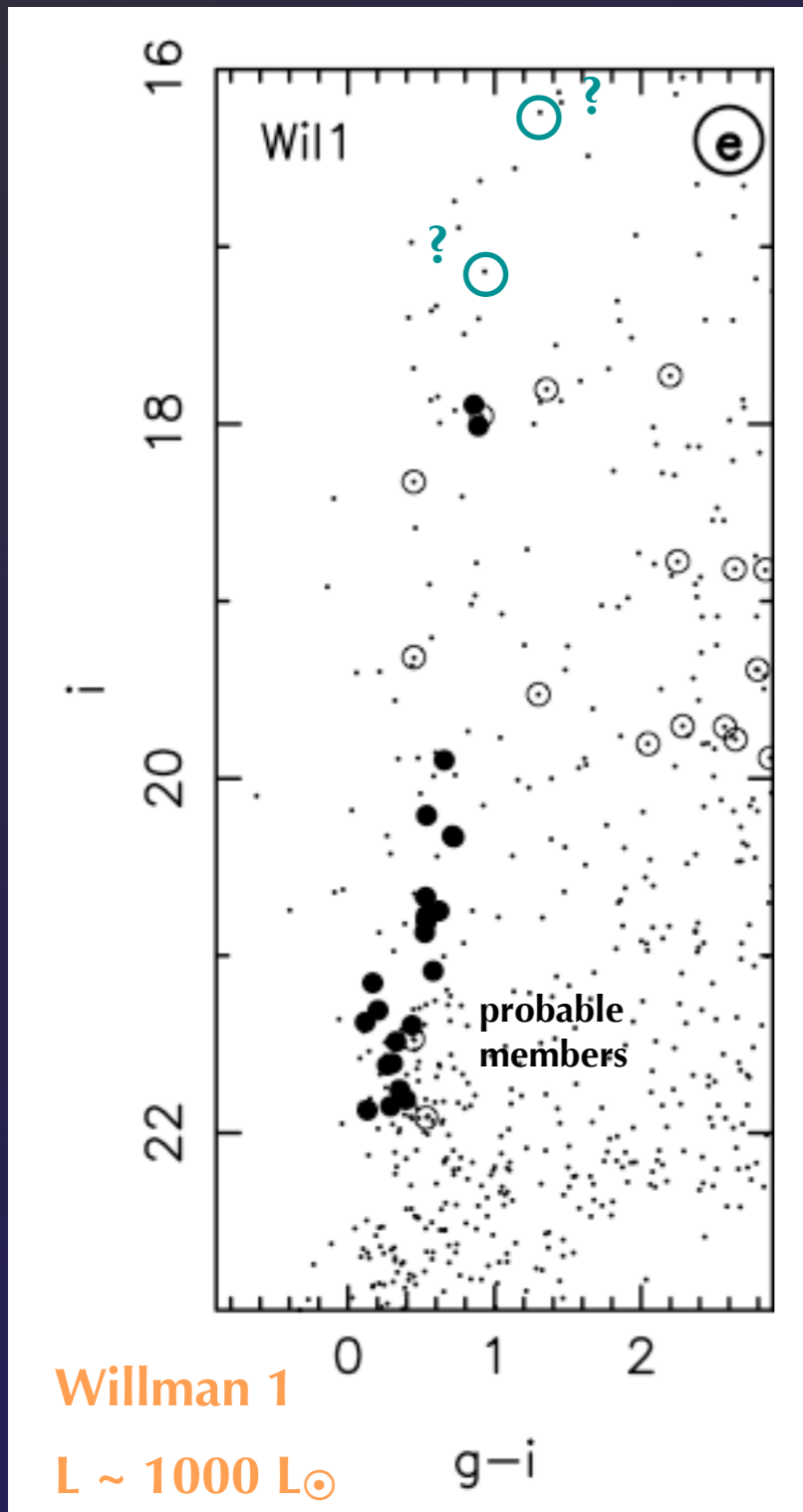
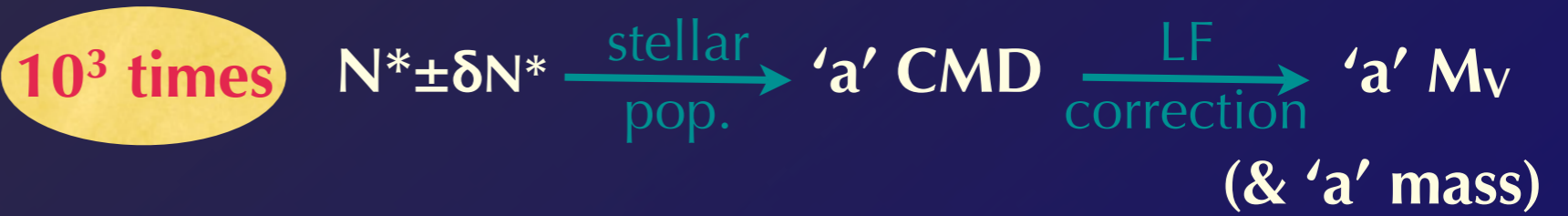
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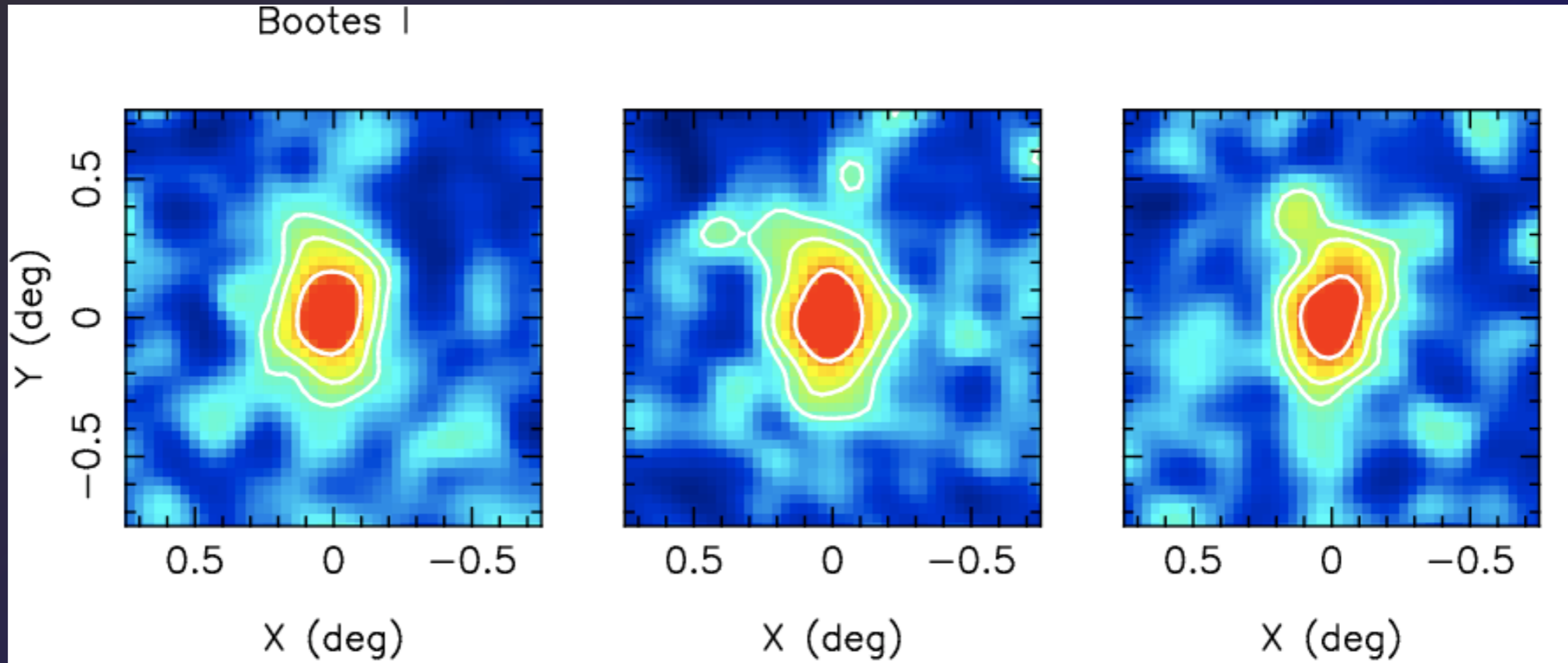


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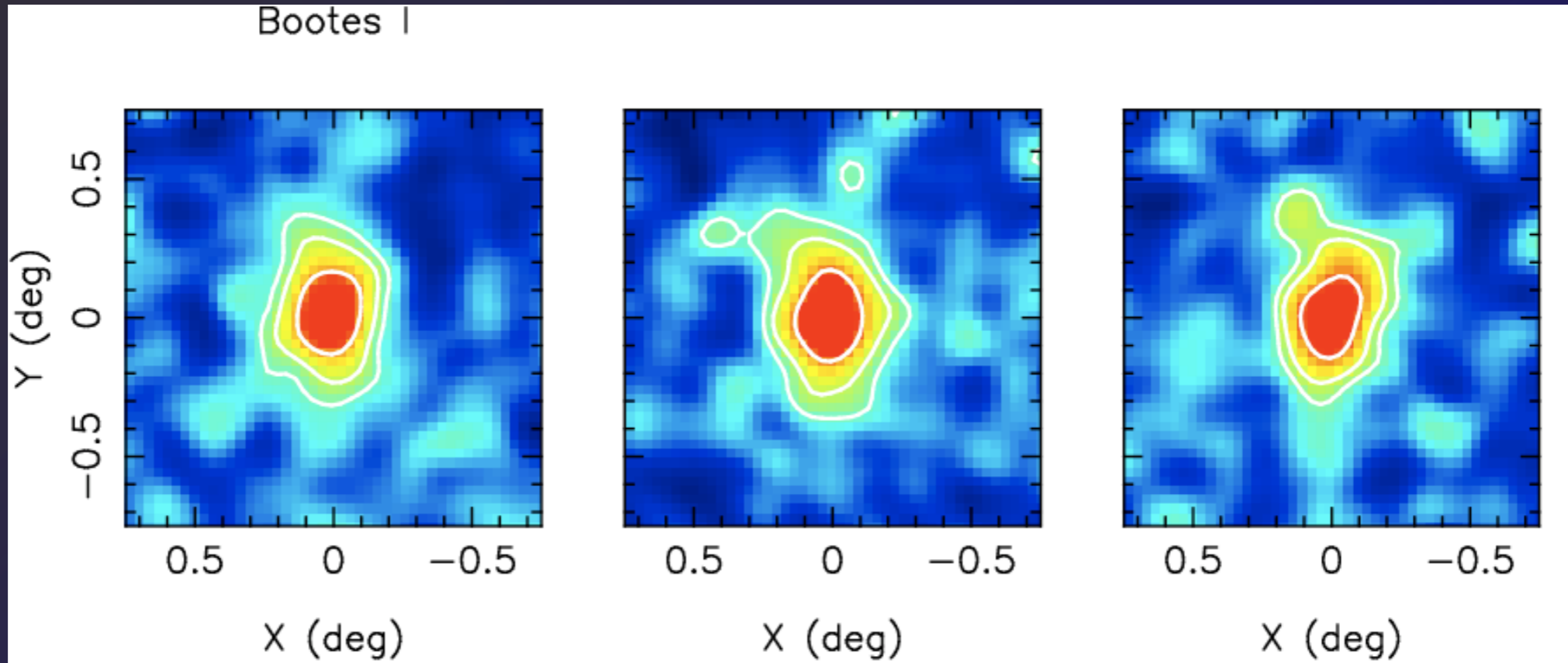
SIGNS OF DISTORTION?



Amount of scatter in pixel counts, accounting for Poisson noise with 4 pixel grids:

$$\left(\frac{\sigma_{\text{sc}}}{\text{total}}\right)^2 = \left(\frac{1}{N} \sum_{i=1}^N (D_i - M_i)^2 - \frac{1}{N} \sum_{i=1}^N (P_i - M_i)^2\right) \cdot \left(\frac{1}{N} \sum_{i=1}^N D_i\right)^{-2}$$

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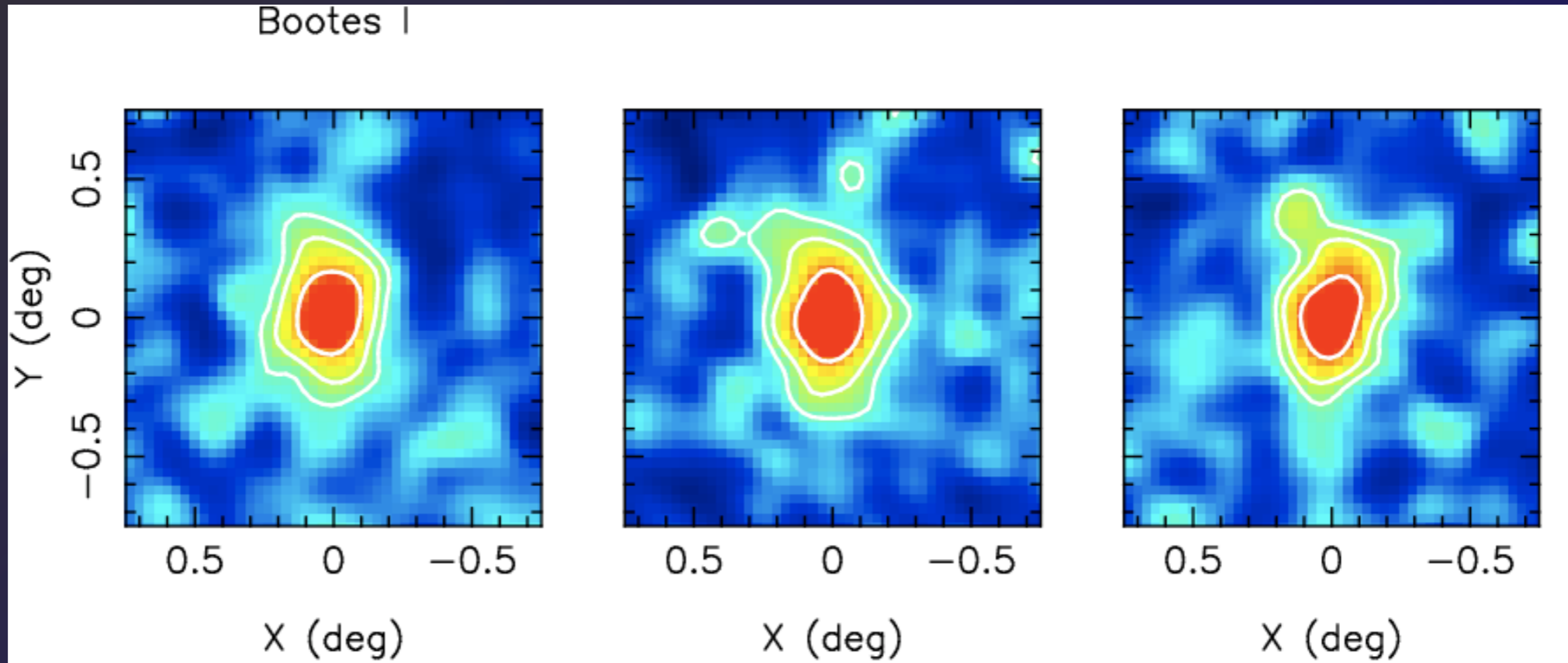


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Scatter in the data wrt model

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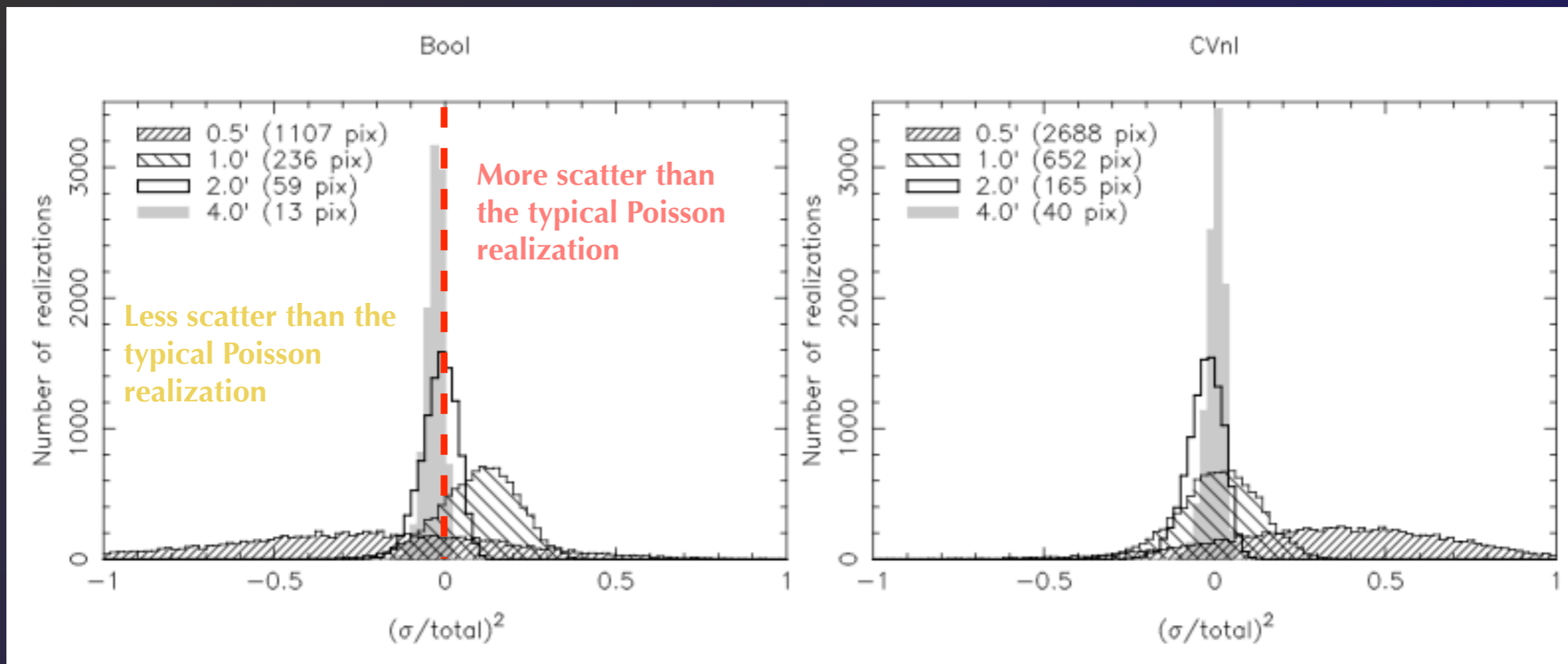


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Scatter in the data wrt model

Scatter in a Poisson realization of best model

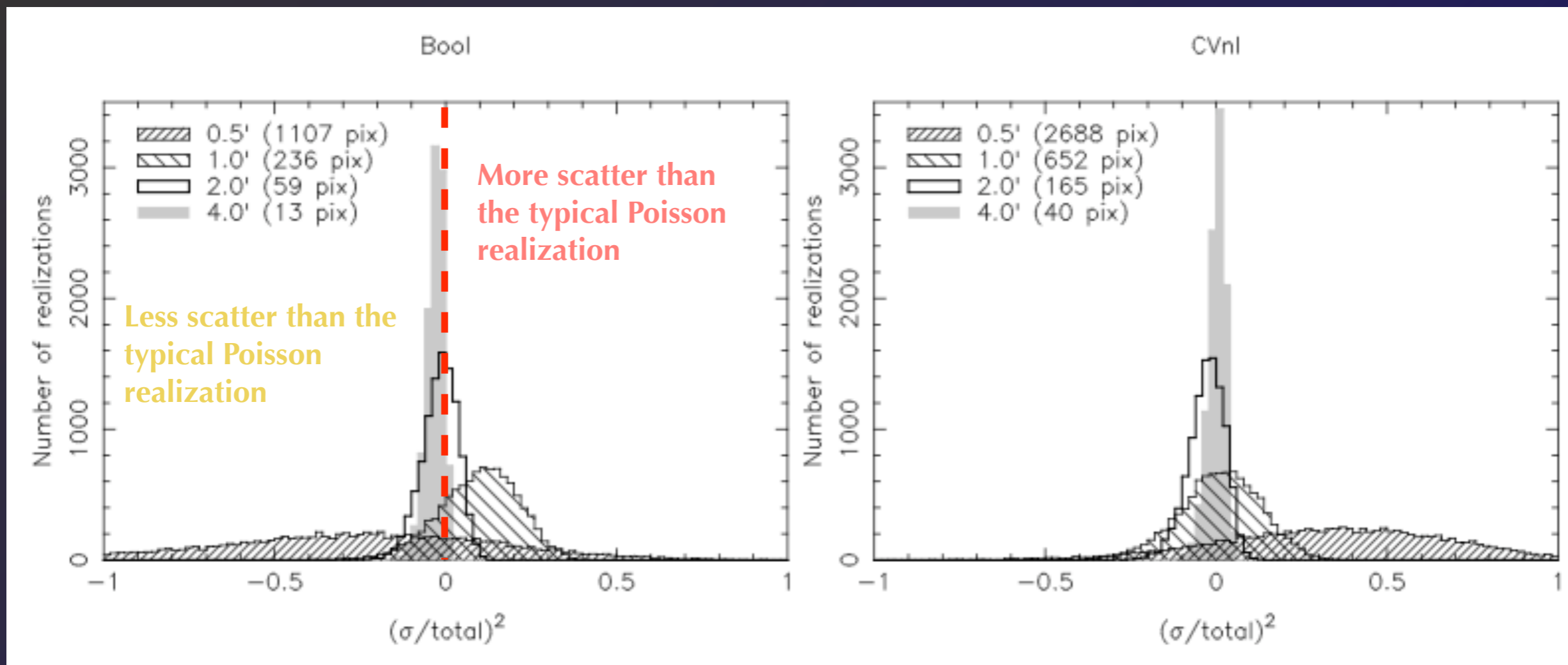


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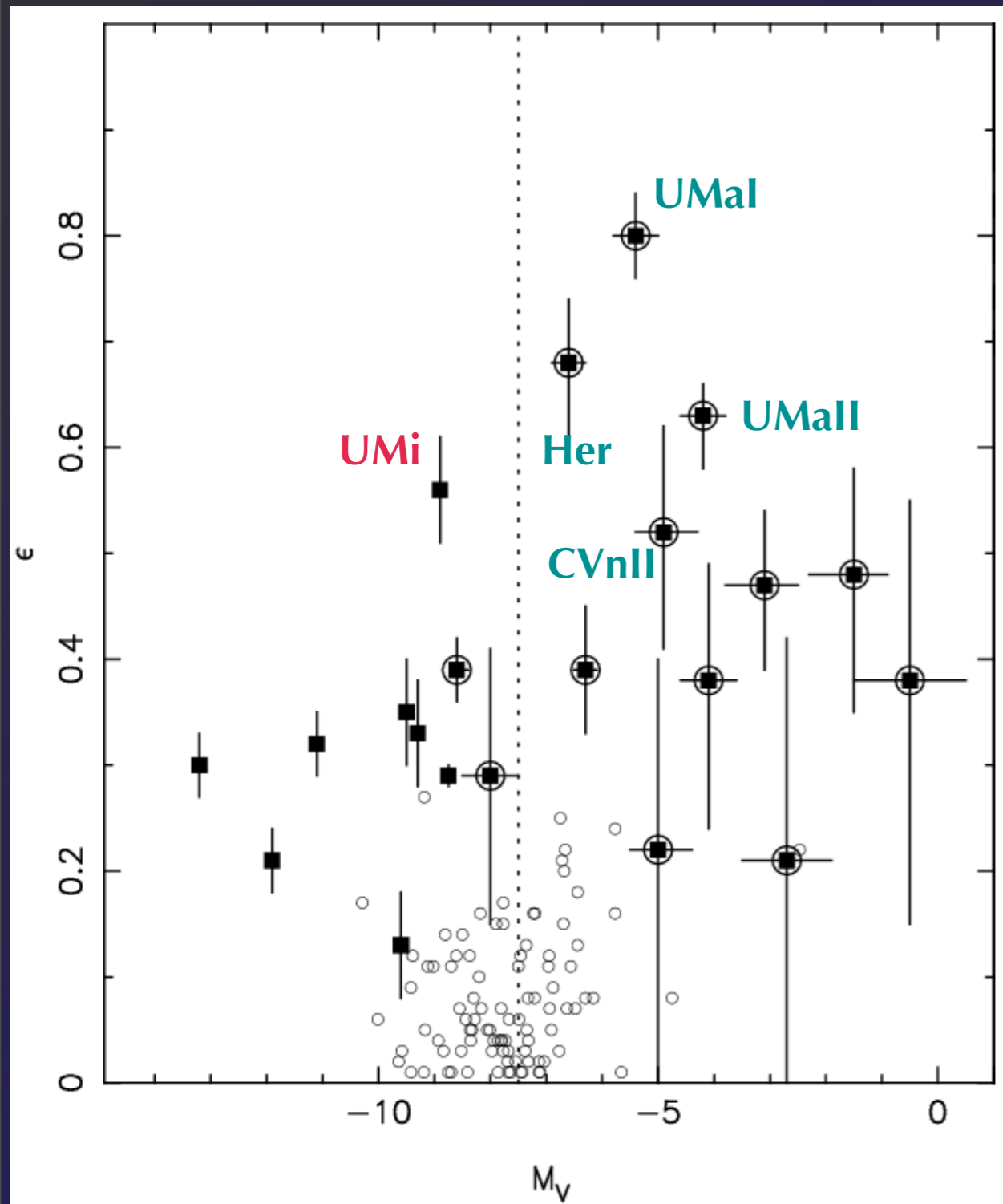
Marginal distortions ($1-2\sigma$) detected in CVnl & UMaII but none otherwise

➔ No distortions or SDSS not deep enough to find them

Faint galaxies appear **flatter** than bright galaxies

ELLIPTICITY

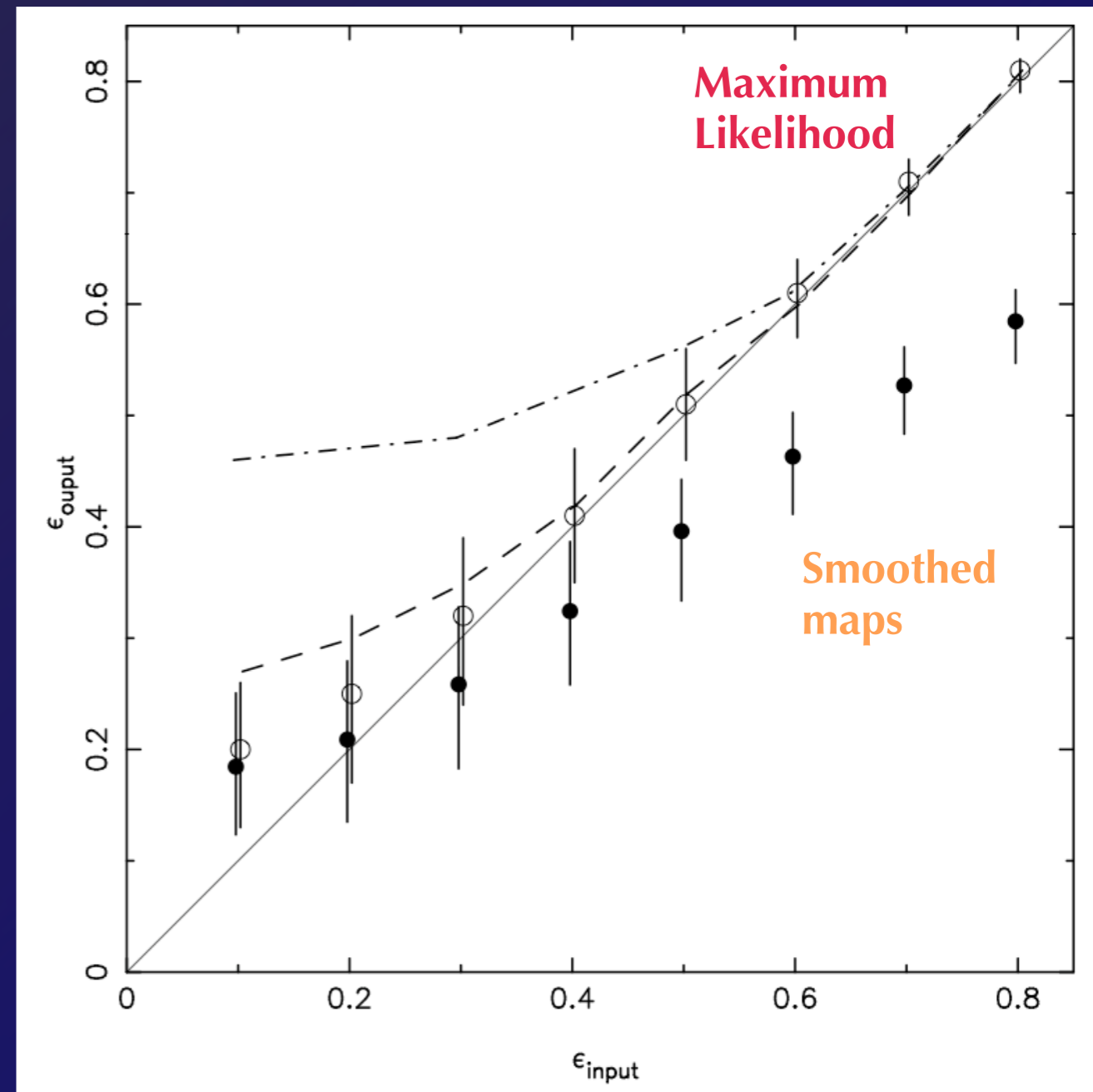
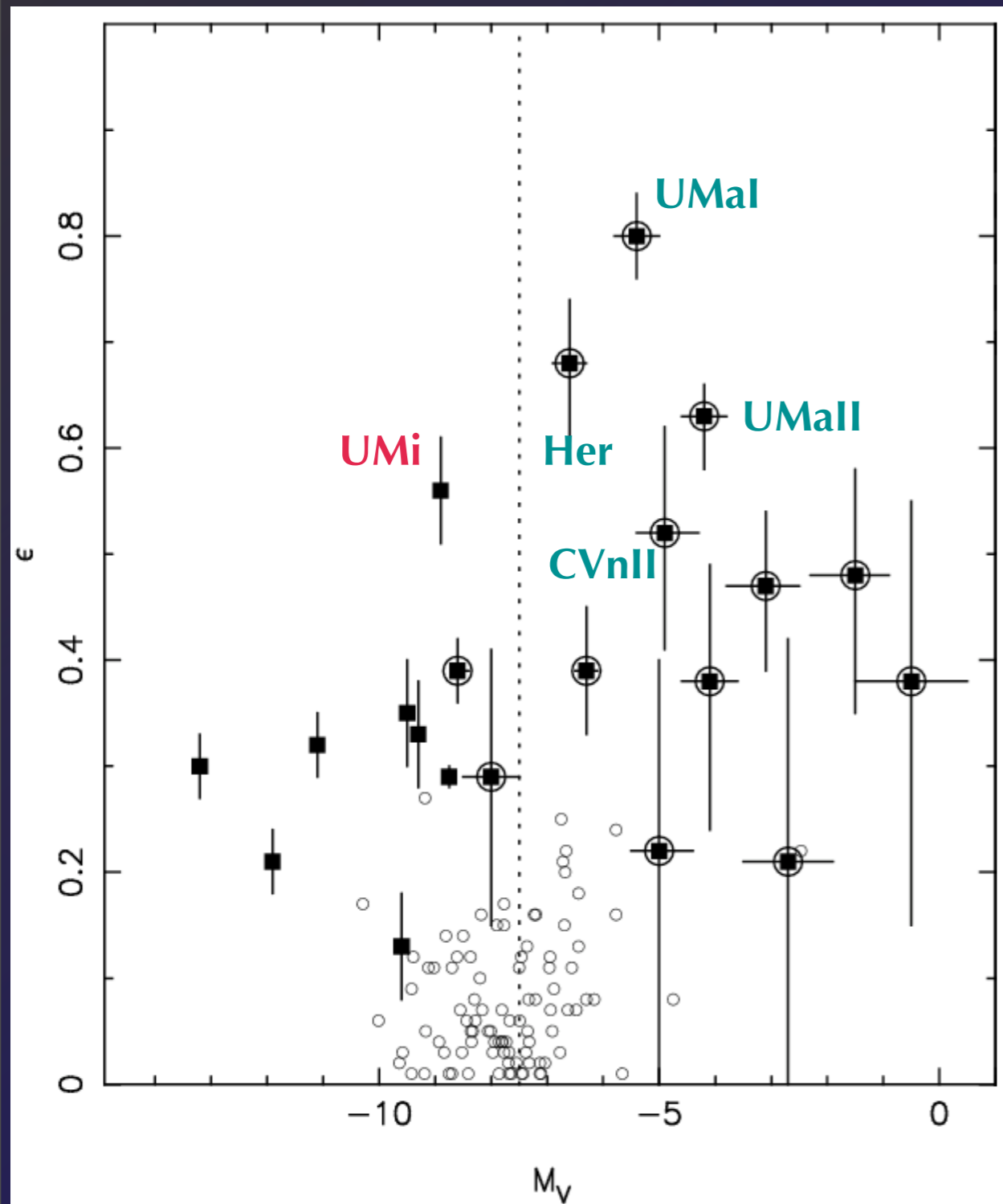
- mean $\epsilon = 0.32 \pm 0.02$ ($M_V > -8.0$) vs. $\epsilon = 0.47 \pm 0.03$ ($M_V > -8.0$)
- KS test: **99.6%** proba that **different** subsamples
- **3 most flattened** systems are faint



Faint galaxies appear **flatter** than bright galaxies

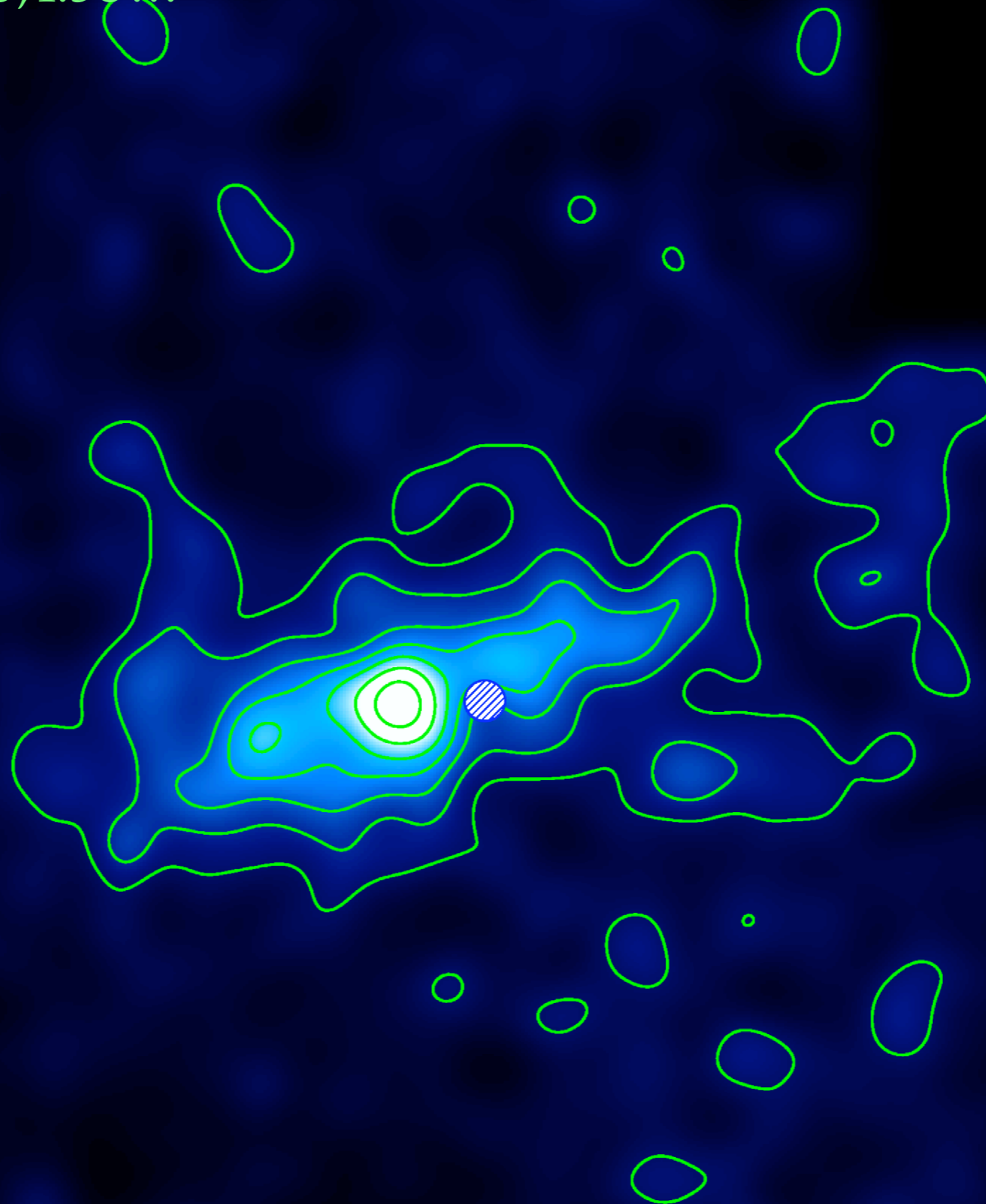
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HERCULES

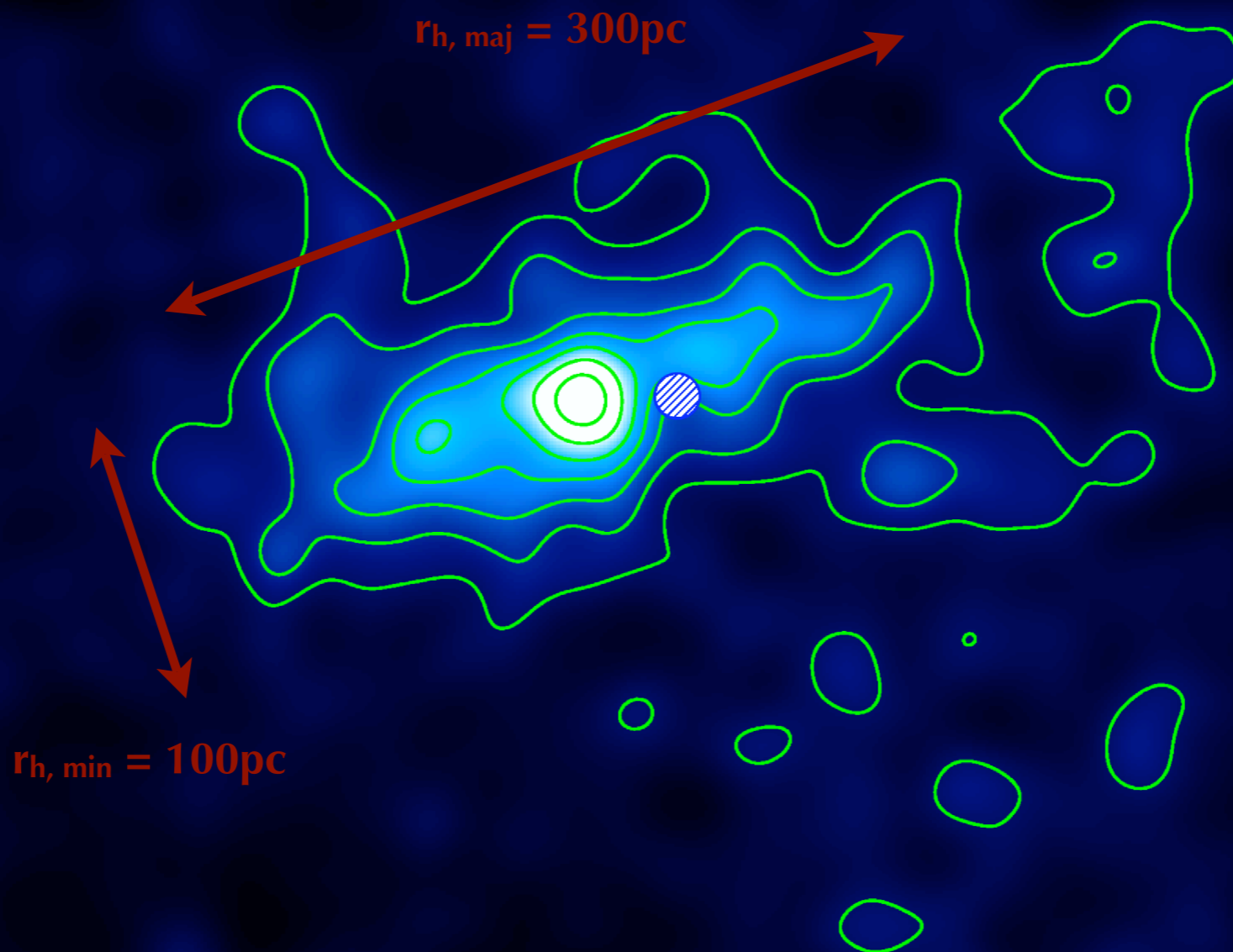
1.5 σ , 3 σ , 4.5 σ ...



Coleman et al. (2007)

HERCULES

$1.5\sigma, 3\sigma, 4.5\sigma \dots$



Coleman et al. (2007)

A TIDALLY SHAPED HERCULES?

Sky

$$\epsilon = 0.68$$

Observational
“fact”

A TIDALLY SHAPED HERCULES?

Deprojected
ellipticity

Sky

$\epsilon=0.68$

Observational
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A TIDALLY SHAPED HERCULES?

Deprojected ellipticity



Sky



Observational
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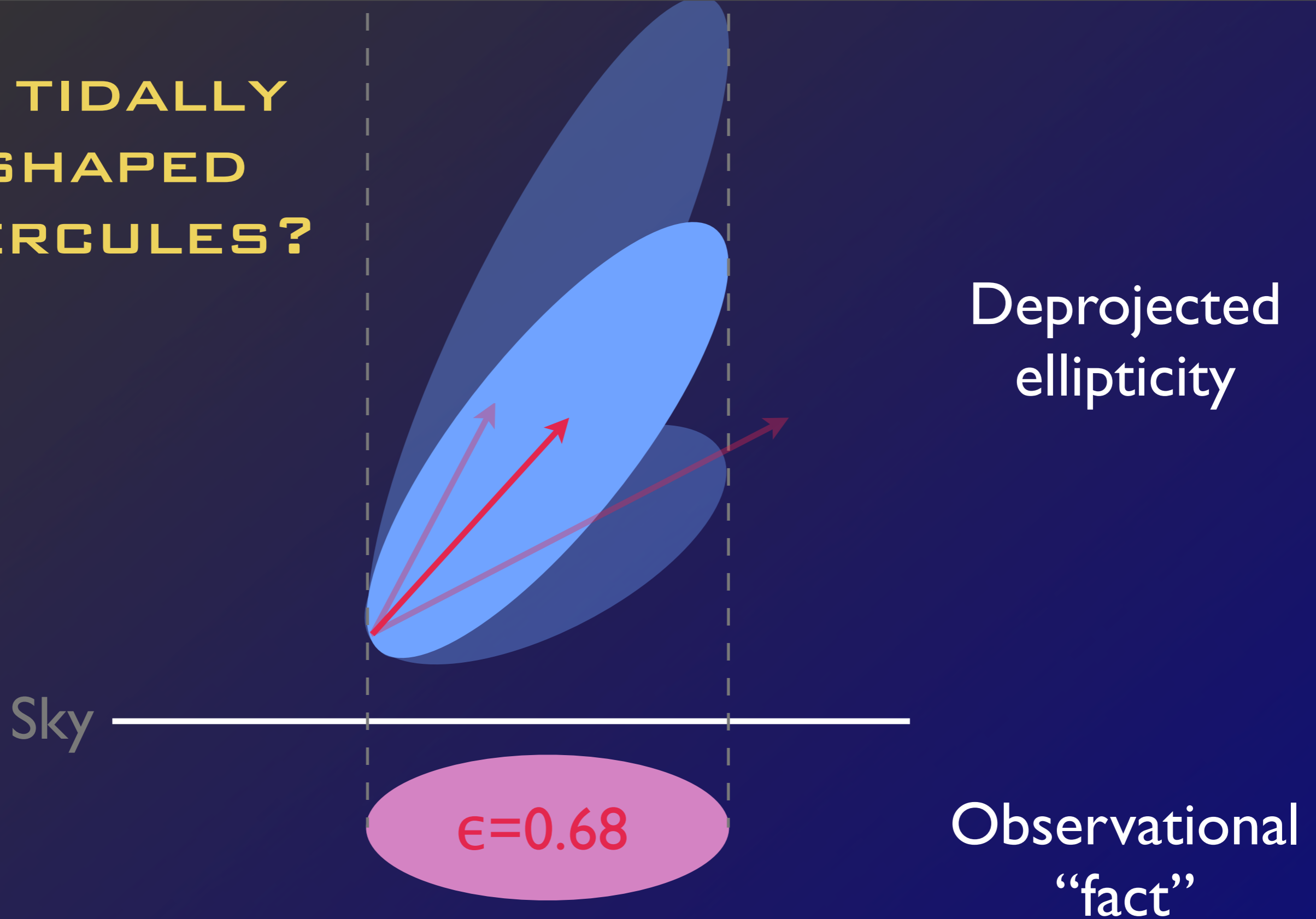
Sky



Deprojected
ellipticity

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A TIDALLY SHAPED HERCULES?



A TIDALLY SHAPED HERCULES?

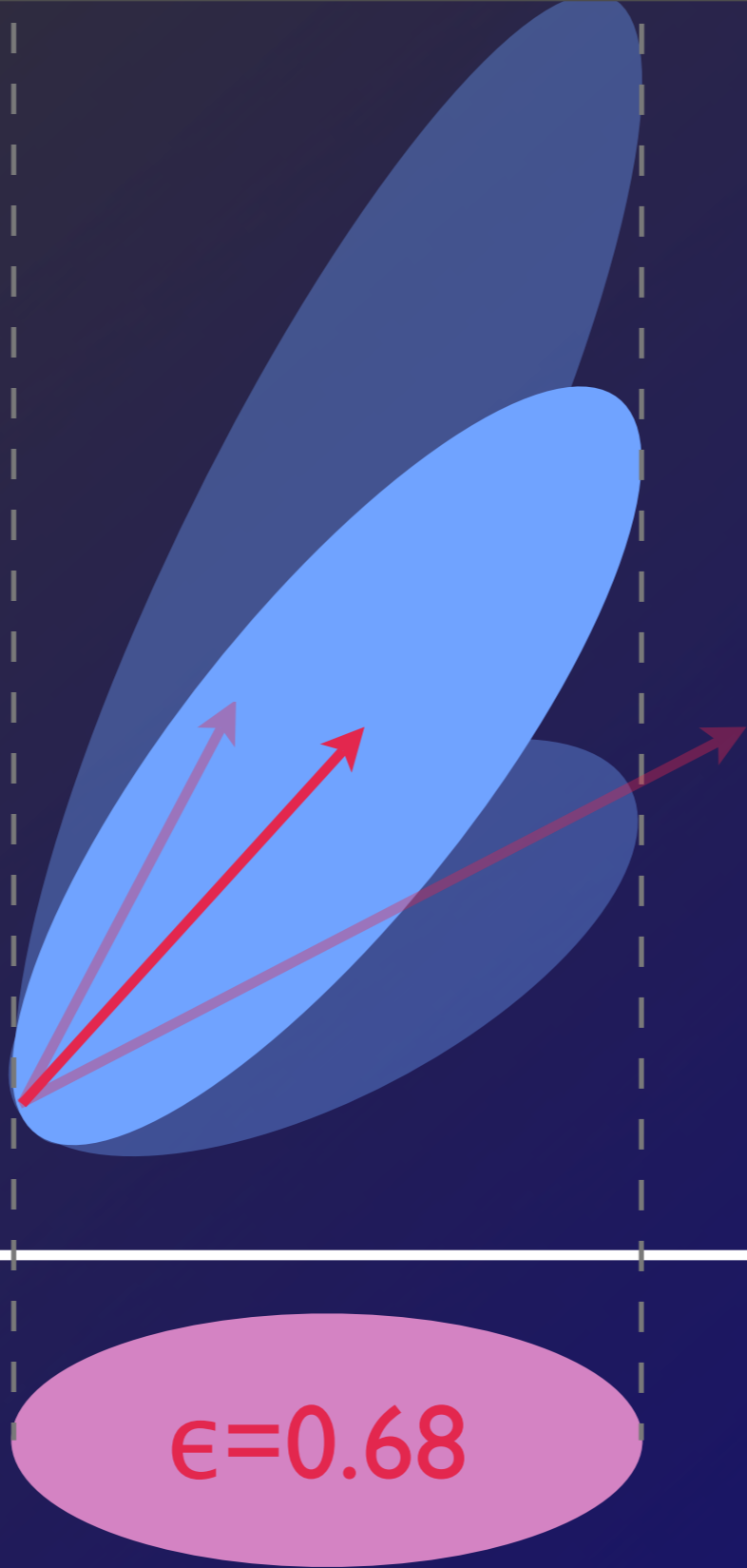
$$v_r = 145 \text{ km/s}$$

Deprojected ellipticity

Sky

$$\epsilon = 0.68$$

Observational
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A TIDALLY SHAPED HERCULES?

$$v_r = 145 \text{ km/s}$$

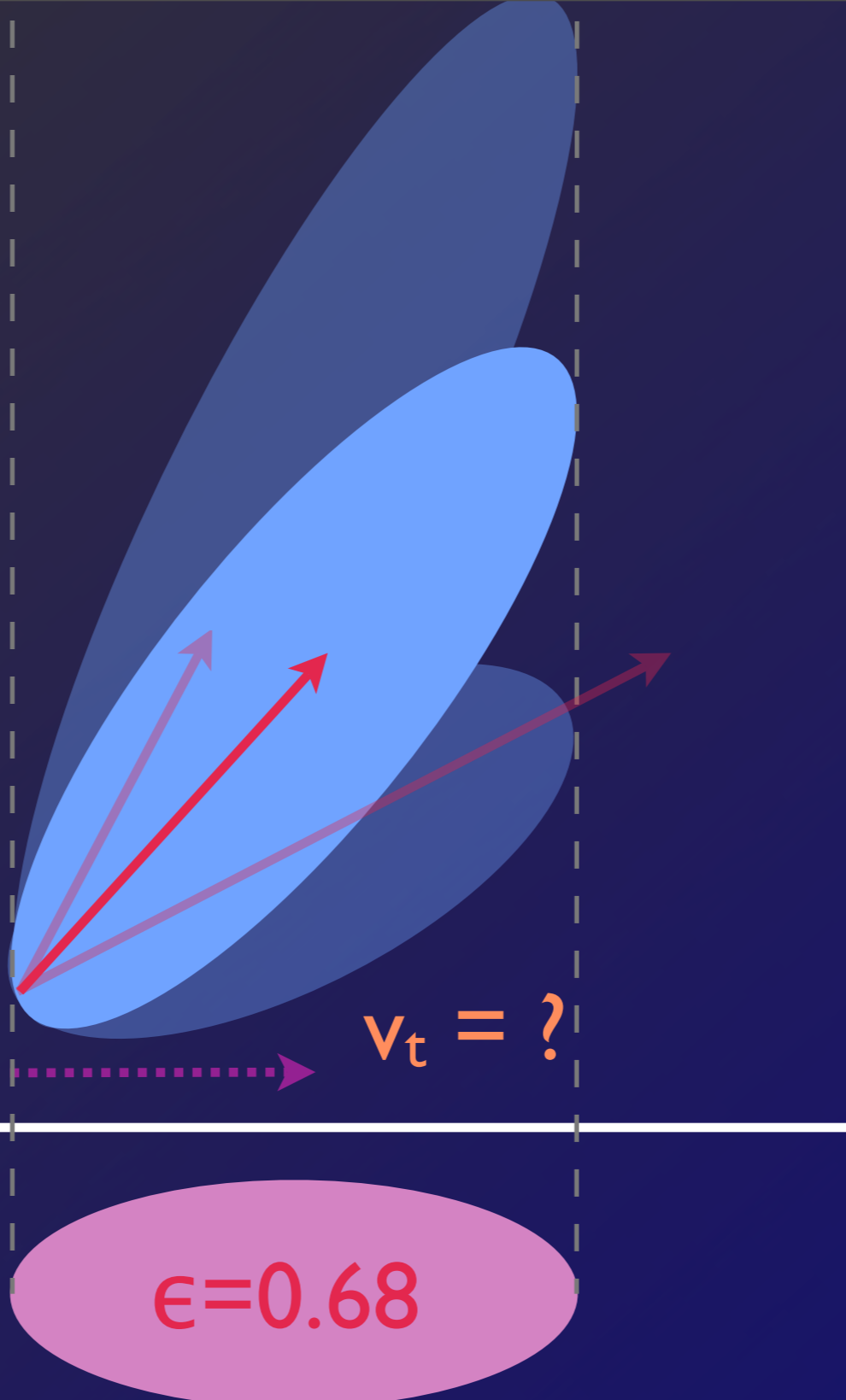
$$v_t = ?$$

Deprojected ellipticity

Sky

$$\epsilon = 0.68$$

Observational
“fact”



A TIDALLY SHAPED HERCULES?

$v_r = 145 \text{ km/s}$

$v_t = ?$

Deprojected ellipticity

Sky

$\epsilon = 0.68$

Observational
“fact”

$$\epsilon_{\text{true}} = 1 - \left(\frac{b}{a}\right)_{\text{true}} = 1 - \left(\frac{b}{a}\right)_{\text{obs}} \left(1 + \frac{v_r^2}{v_t^2}\right)^{-1/2}$$

A TIDALLY SHAPED HERCULES?

