

HALO MASS FUNCTION RESCALING AND CLUSTER COUNTS COSMOLOGY

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Action Dark Energy 2024

Théo Gayoux

LUTH  Observatoire
de Paris

Action Dark Energy 2024

PSL 



CONTEXT

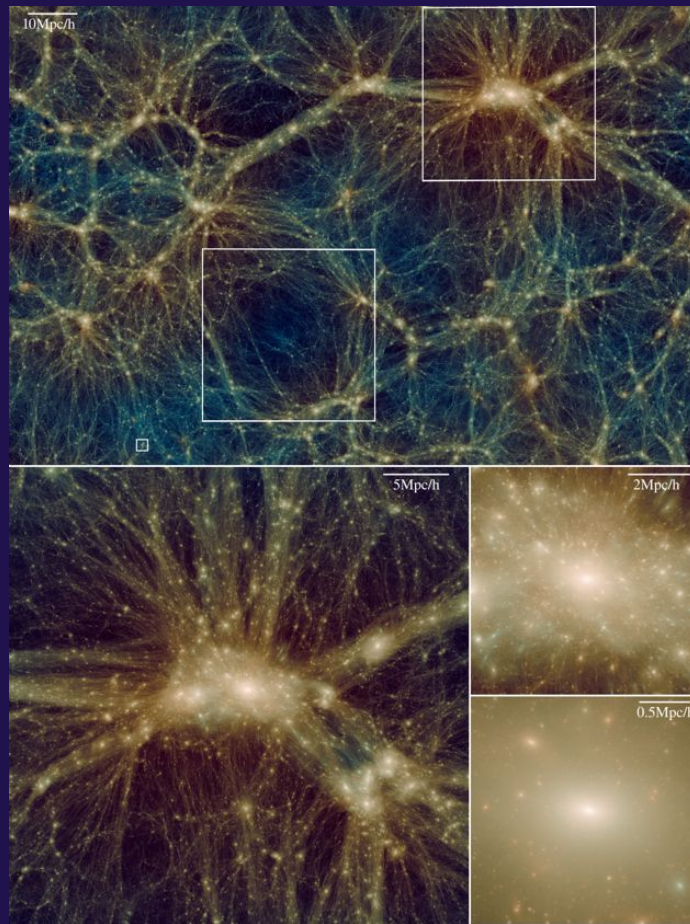
GALAXY CLUSTERS

Properties :

→ Largest structures gravitationally virialized in the Universe

→ Multicomponent systems : Dark Matter (~85%) & Baryons

→ Abundance very sensitive to cosmological parameters (e.g. σ_8 , Ω_m)



<https://skiesanduniverses.org/Simulations/Uchuu/Visualize/>

CONTEXT

GALAXY CLUSTERS

Observational side

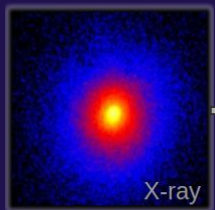


Richness
→ M_{200c}

Scaling relations



$$M_{\Delta c} = \frac{4}{3}\pi R_{\Delta}^3 \Delta\rho_c$$



X-ray
luminosity
→ M_{500c}

Images of Abell 1835 ($z = 0.25$)

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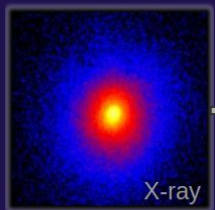


Optical

Richness

→ M_{200c}

Scaling relations

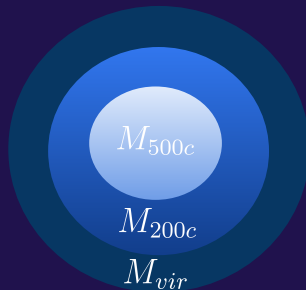


X-ray

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SO masses



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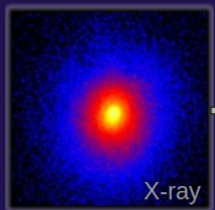
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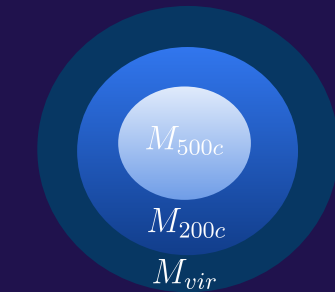
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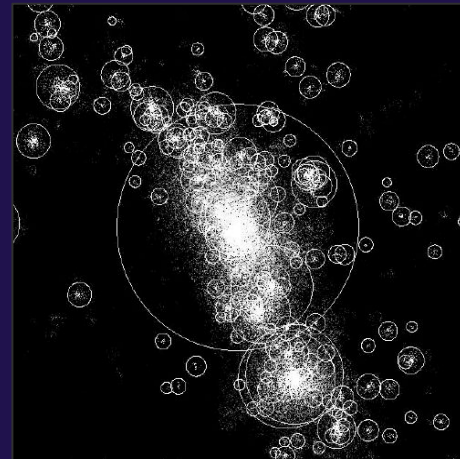


$$M_{\Delta c} = \frac{4}{3} \pi R_{\Delta}^3 \Delta \rho_c$$

Halo finders



Theoretical side



[Kravtsov et Al. 2003](#)

N-body simulations - Dark
matter halos

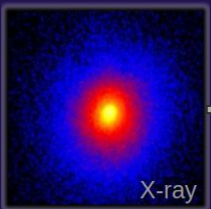
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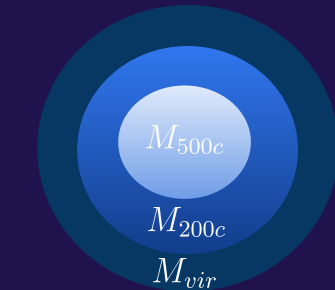


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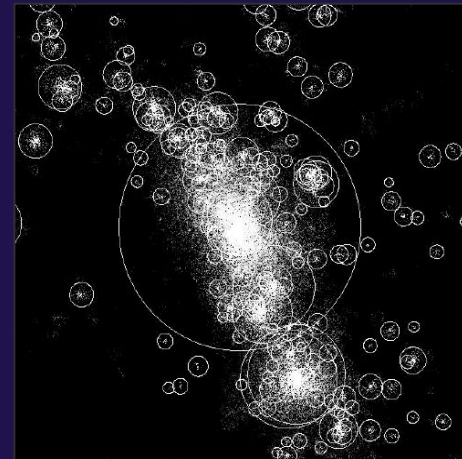


Fitting halo
catalogues



$$\frac{dn}{dM_{\Delta}} \text{ Halo mass function (HMF)}$$

Theoretical side



[Kravtsov et Al. 2003](#)

N-body simulations - Dark matter halos

Images of Abell 1835 ($z = 0.25$)

CONTEXT

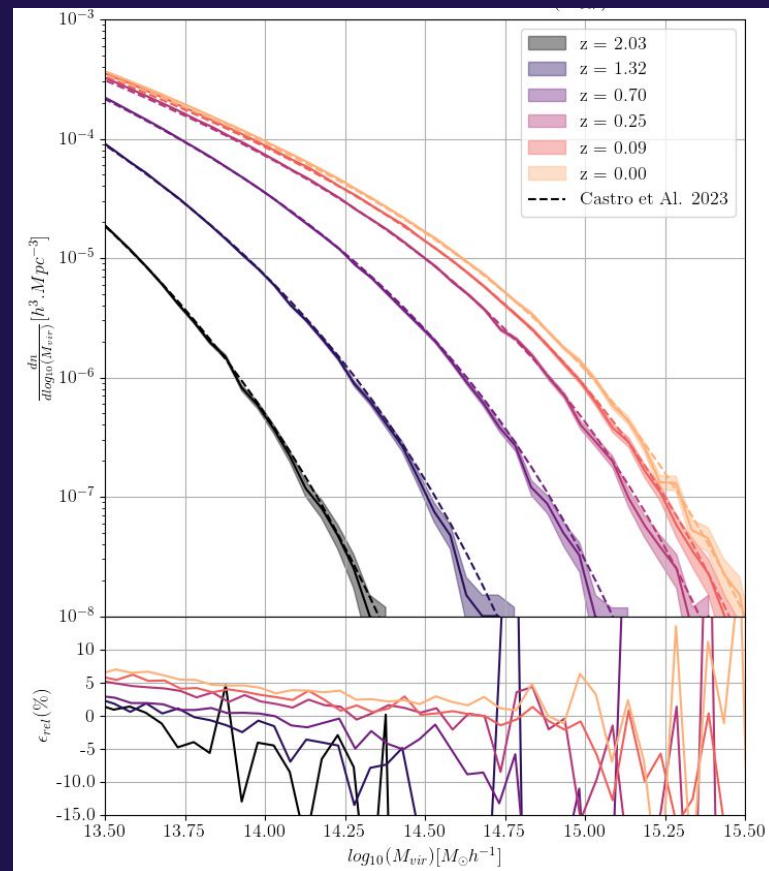
HALO MASS FUNCTION

$\frac{dn}{dM_{\Delta}}$ Number of haloes per unit volume
per mass bin

Euclid HMF (Castro et al. 2023)

→ Calibrated at M_{vir}

→ Captures deviation from
non-universality.



CONTEXT

NUMBER COUNT

Expected number count (NC): Number of clusters of a given observable X and z within the survey area

$$\frac{dN(X; z)}{dX dz} = \left(\frac{dV}{dz} \right) f(X, z) \int_0^\infty \frac{dn(M, z)}{dM_\Delta} \frac{dp(X|M, z)}{dX} dM$$

Survey's volume

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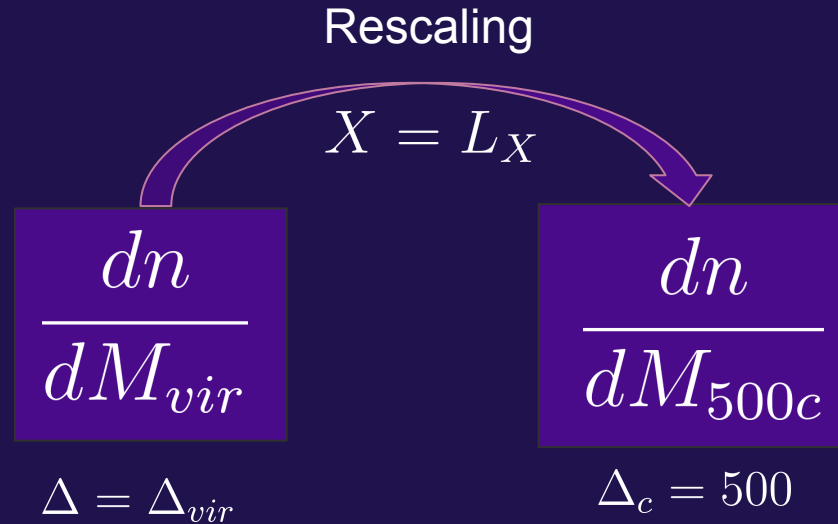
Survey's volume Selection function **HMF** Probability of X given M and z

e.g. we have $\frac{dn(M, z)}{dM_{vir}}$ (e.g. Euclid HMF) and we have $X = L_X$

OBJECTIF AND METHODS

RESCALING

Objective : Rescale the HMF (e.g. Euclid HMF Castro et al 23.) to match observables using accurate method to reduce the systematics on the NC



OBJECTIF AND METHODS

METHODS - DETERMINISTIC (C-M)

What we need

Chain rule: $\frac{dn}{dM_{\Delta}} = \frac{dn}{dM_{vir}} \frac{dM_{vir}}{dM_{\Delta}}$

$$M_{\Delta} = f(M_{vir}) ?$$

OBJECTIF AND METHODS

METHODS - DETERMINISTIC (C-M)

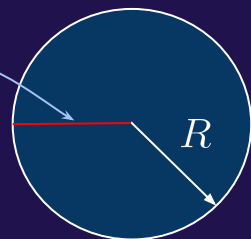
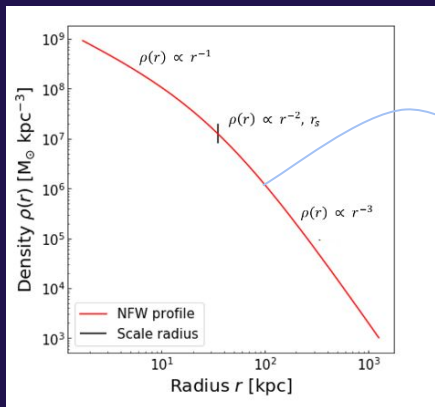
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$M_{\Delta} = f(M_{vir}) ?$

NFW profile (Navarro, Frenk & White 97)

$$\rho_{NFW}(r) = \frac{M_{vir}}{4\pi[\ln(1 + c_{vir}) - c_{vir}/(1 + c_{vir})]} \times \frac{1}{r \left(\frac{R_{vir}}{c_{vir}} + r \right)^2}$$



Spherical DM halo

OBJECTIF AND METHODS

METHODS - DETERMINISTIC (C-M)

What we need

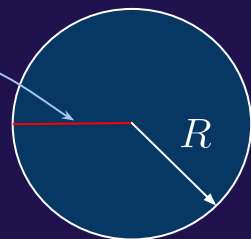
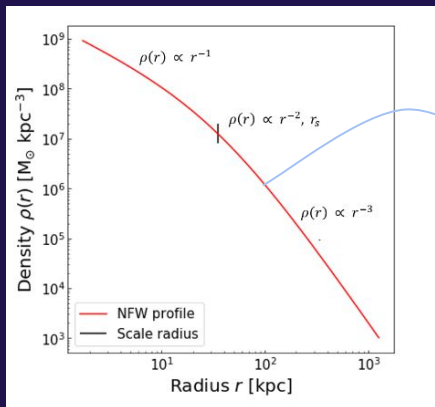
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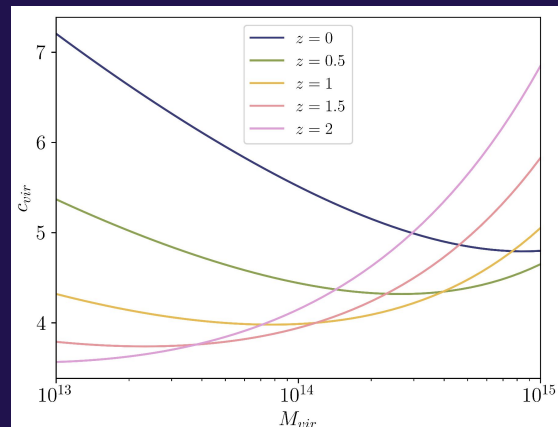
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c-M relation (Ishiyama et al. 21) $c_{vir} = f(M_{vir}, z, cosmo)$



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OBJECTIF AND METHODS

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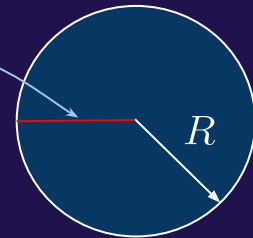
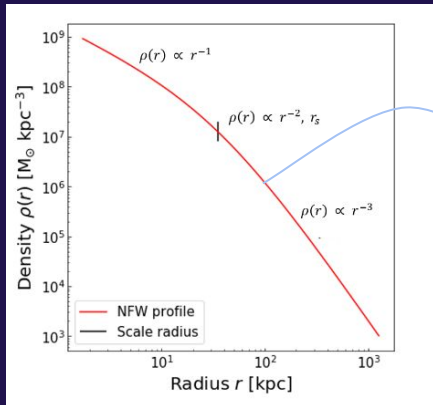
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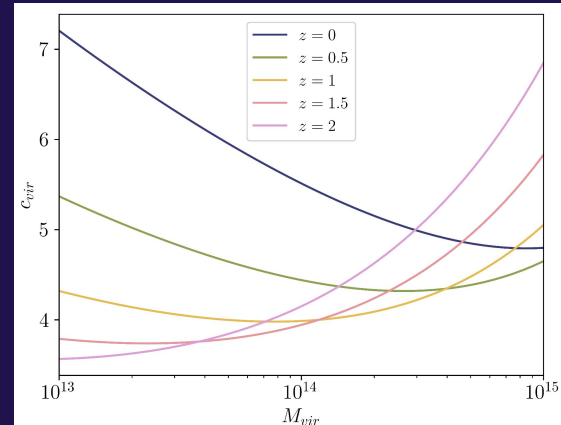
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Spherical DM halo



We can deduce M_{Δ} knowing ρ_{NFW} (Hu & Kratsov 18)

OBJECTIF AND METHODS

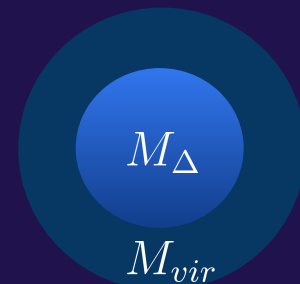
METHODS - SEMI-DETERMINISTIC (C-M + PDF)

M : random variable drawn from the HMF

$$\frac{dn_{\Delta}}{dM_{\Delta}} = \int \frac{dn_{vir}}{dM_{vir}} \rho(M_{\Delta} | M_{vir}) dM_{vir}$$

Despali et al. 2016

$$M_{\Delta} \sim \frac{dn}{dM_{\Delta}}$$

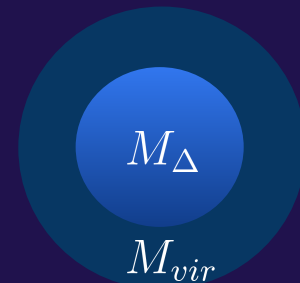


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Internal structure

Despali et al. 2016

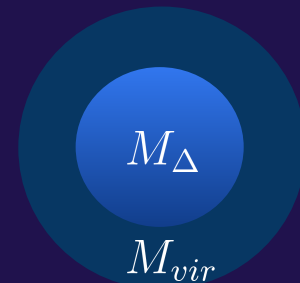
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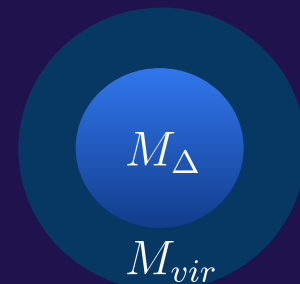
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Internal structure

Despali et al. 2016

What we need

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$$\rho(M_{\Delta} | M_{vir}) dM_{vir} = \rho(c_{vir} | M_{vir}) \left| \frac{dM_{vir}}{dc_{vir}} \right| dc_{vir}$$

log-normal distribution c-M relation

OBJECTIF AND METHODS

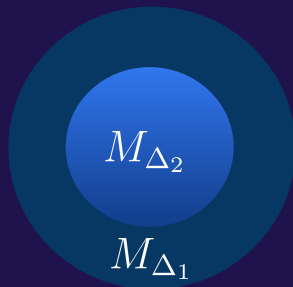
METHODS - STOCHASTIC (SPARSITIES)

→ **Non-parametric approach** using sparsities (Richardson & Corasaniti 2023)

Sparsities

$$s_{\Delta_1, \Delta_2} = \frac{M_{\Delta_1}}{M_{\Delta_2}}$$

$$\Delta_1 < \Delta_2$$



OBJECTIF AND METHODS

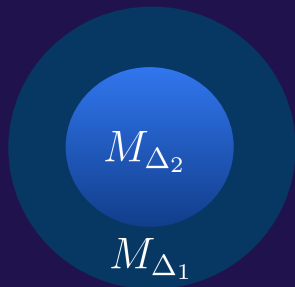
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s : random variable

$$s \sim \rho_s(s | M_{vir})$$

OBJECTIF AND METHODS

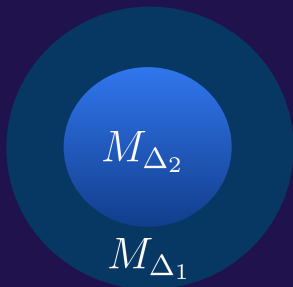
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Internal structure: Sparsities p.d.f.s

$$\rho(M_{\Delta} | M_{vir}) dM_{vir} = s \rho(s | s M_{\Delta}) ds$$

Calibrated with
N-body sim

OBJECTIF AND METHODS

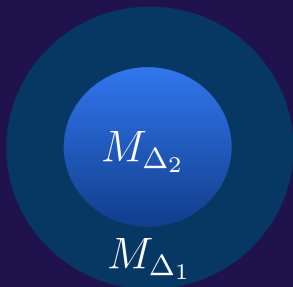
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Calibrated with
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Inward transformation

$$\frac{dn}{dM_{\Delta}}(M_{\Delta}) = \int_1^{\infty} s \rho_s(s | s M_{\Delta}) \frac{dn}{dM_{\Delta_{vir}}}(s M_{\Delta}) ds$$

OBJECTIF AND METHODS

METHODS - STOCHASTIC PDFS

$\rho_s(s | sM_{\Delta_2})$ is calibrated using N-body simulation

→ **Cosmology dependent**

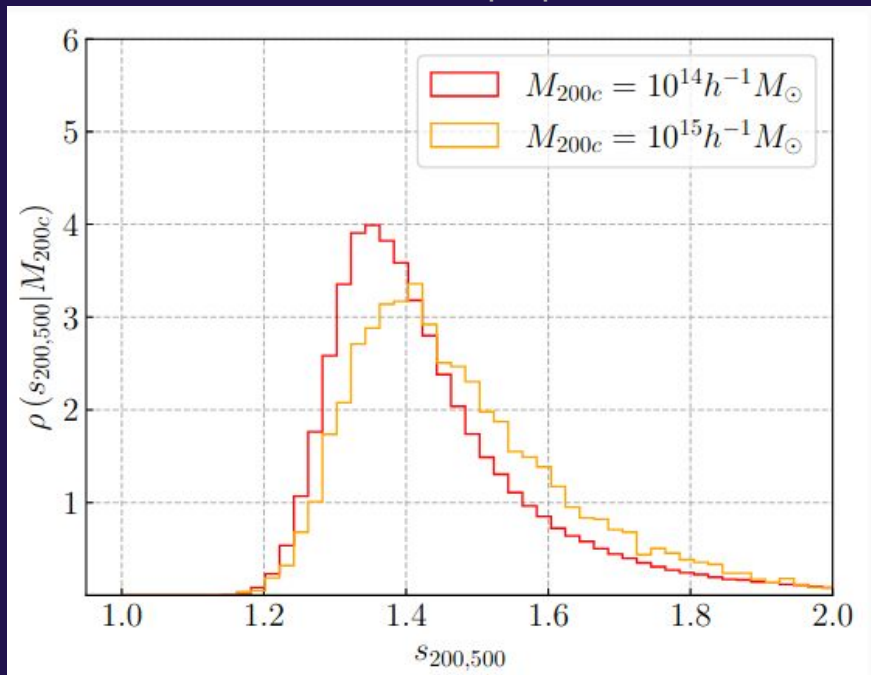
RayGal LCDM (Rasera et al., 2021)

- **Nbody DM only simulation**
- WMAP-7 best-fit cosmological model
- $N_p = 4096^3$ particles
- Halos detected with SO halo finder
- $L = 2625 \text{ Mpc}/h^3$

OBJECTIF AND METHODS

METHODS - STOCHASTIC PDFS

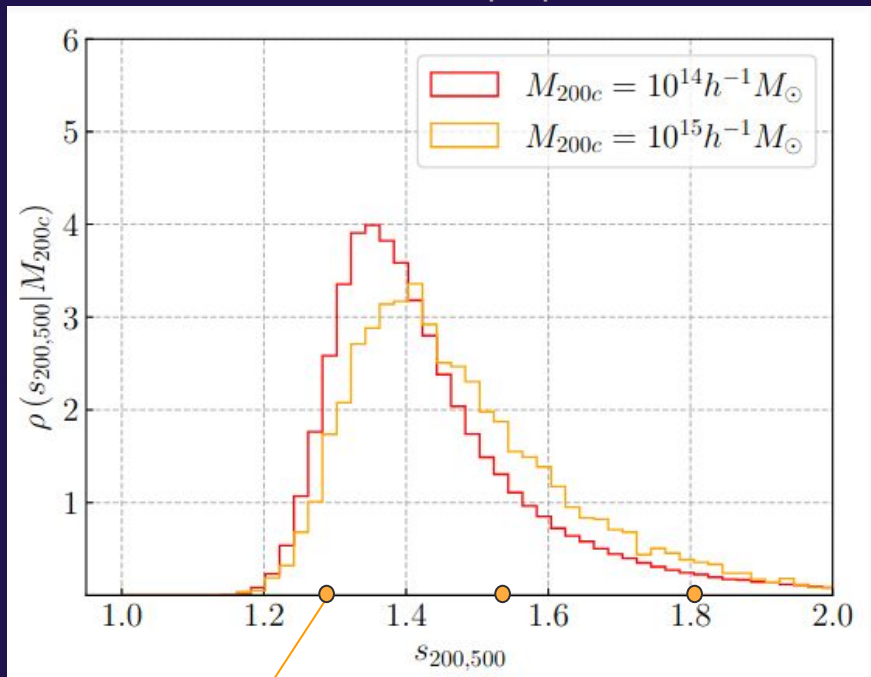
Casarez et al. in prep



OBJECTIF AND METHODS

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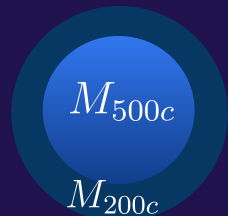
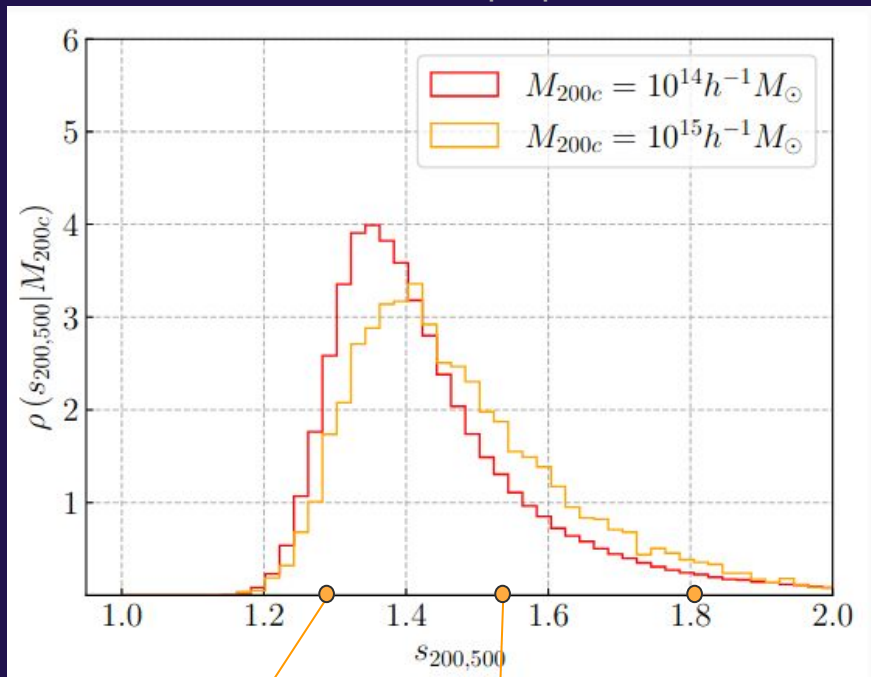
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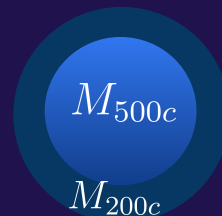
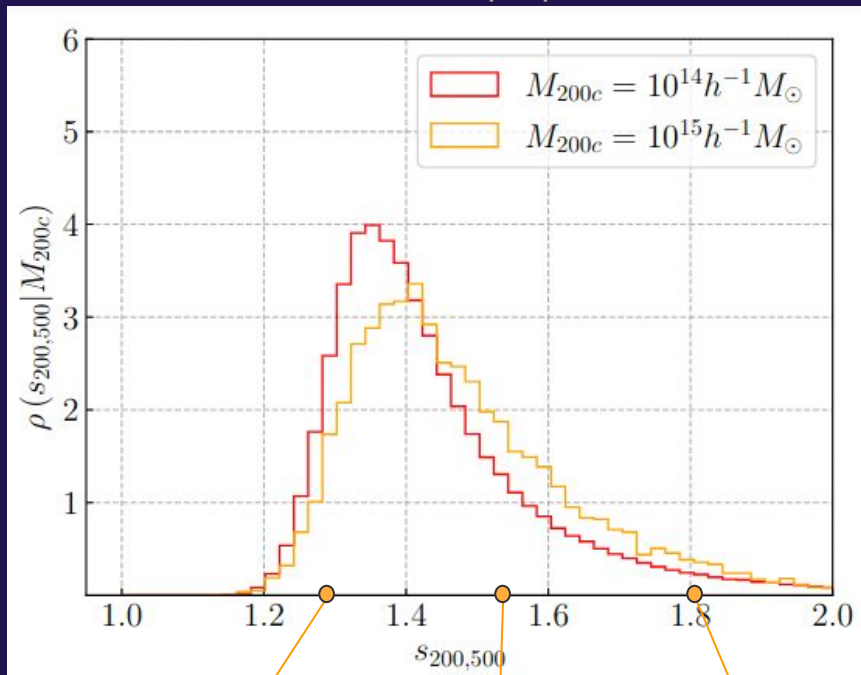
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OBJECTIF AND METHODS

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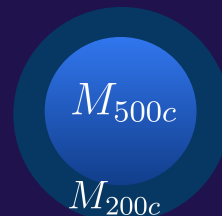
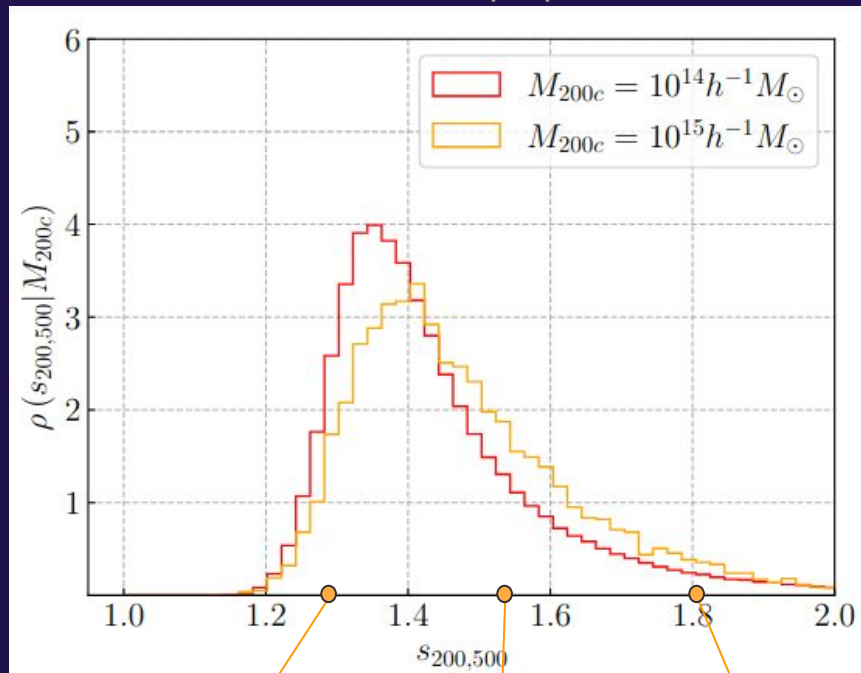


OBJECTIF AND METHODS

METHODS - STOCHASTIC PDFS

→ Give information on matter distribution within haloes with a statistical approach

Casarez et al. in prep



OBJECTIF AND METHODS

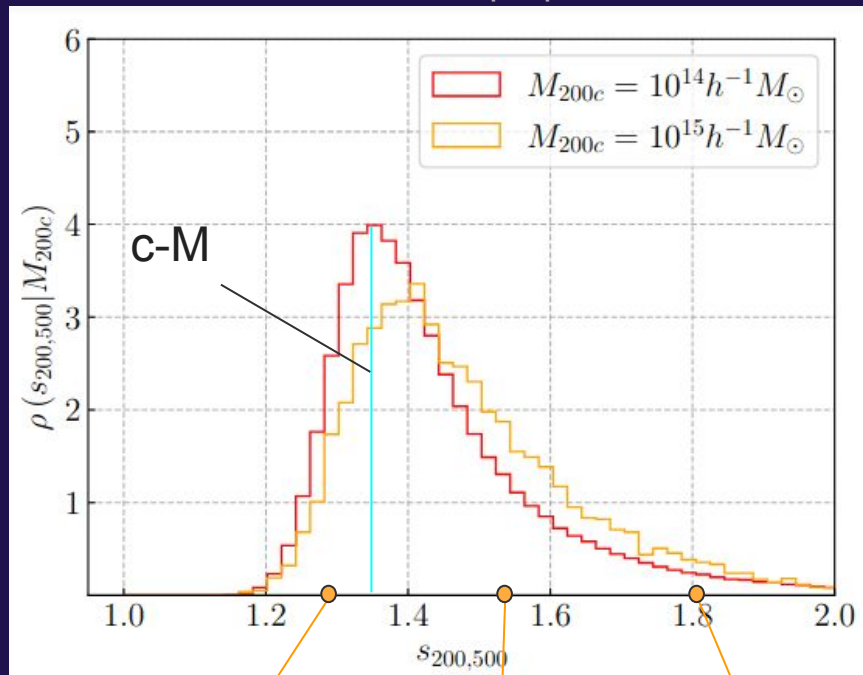
METHODS - STOCHASTIC PDFS

→ Give information on matter distribution within haloes with a statistical approach

→ In the case of c-M :

$$\rho_s(s|sM_\Delta) = \delta(s - s^{NFW}(c))$$

Casarez et al. in prep



OBJECTIF AND METHODS

SUMMARY - 3 METHODS

$$\frac{dn}{dM_{\Delta}} = \frac{dn}{dM_{vir}} \frac{dM_{vir}}{dM_{\Delta}} \quad \mathbf{C-M}$$

$$\frac{dn_{\Delta}}{dM_{\Delta}} = \int \frac{dn_{vir}}{dM_{vir}} \rho(c_{vir} | M_{vir}) \left| \frac{dM_{vir}}{dc_{vir}} \right| dc_{vir} \quad \mathbf{C-M + PDF}$$

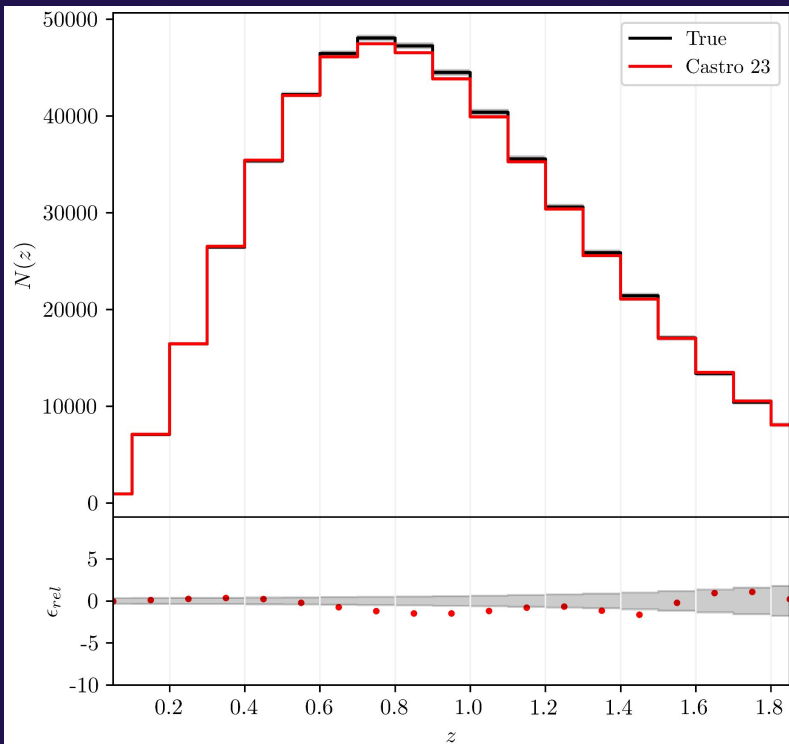
$$\frac{dn}{dM_{\Delta}}(M_{\Delta}) = \int_1^{\infty} s \rho_s(s | sM_{\Delta}) \frac{dn}{dM_{\Delta_{vir}}}(sM_{\Delta}) ds \quad \mathbf{Sparsities}$$

RESULTS (PRELIMINARY)

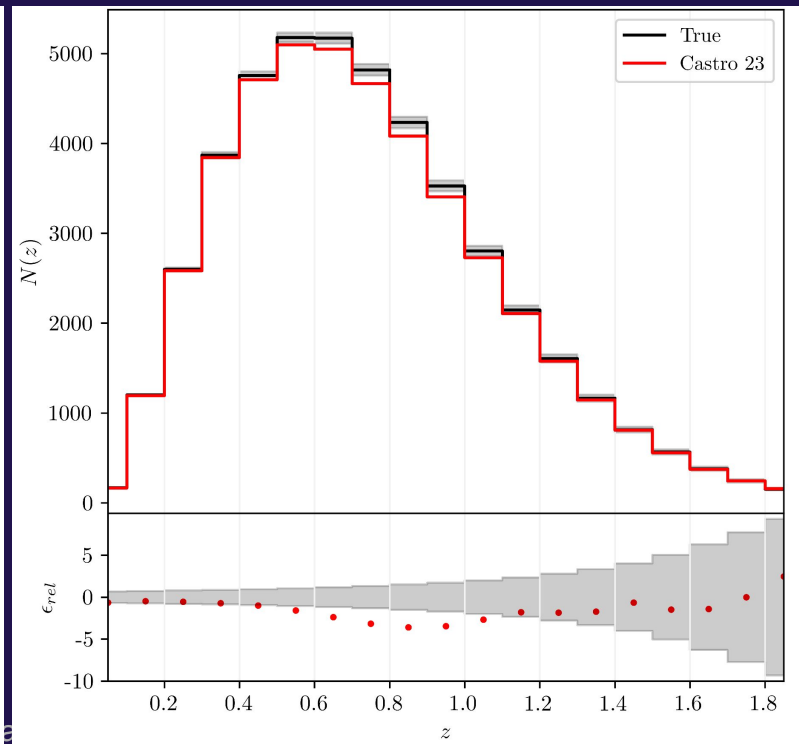
CLUSTER COUNTS - MVIR

$$N_{vir}(z) = \Delta\Omega\Delta z \frac{d^2V}{d\Omega dz} \int_{M_{min}}^{M_{max}} dM \frac{dn}{dM_{vir}}(M, z)$$

$$M_{min} = 3.10^{13} M_{\odot}/h$$



$$M_{min} = 10^{14} M_{\odot}/h$$



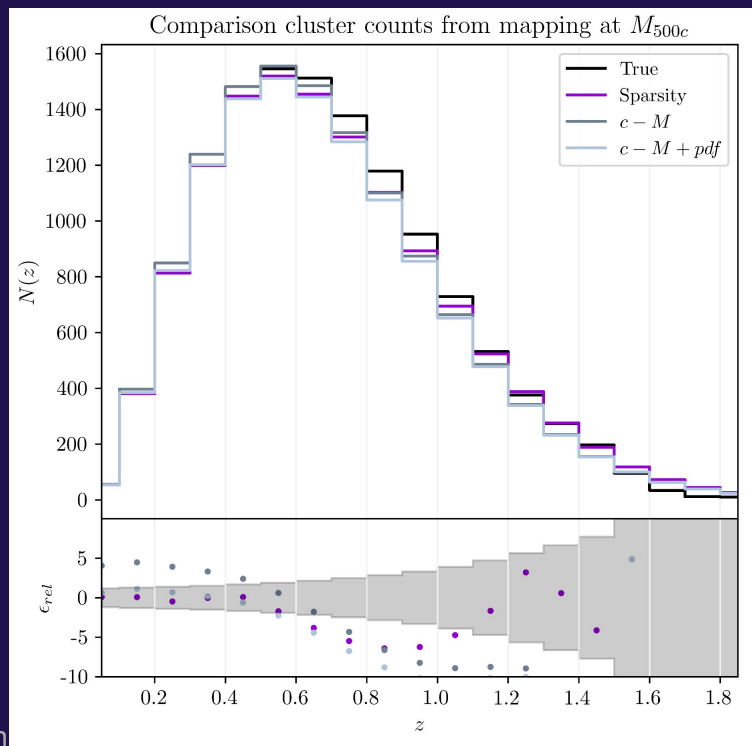
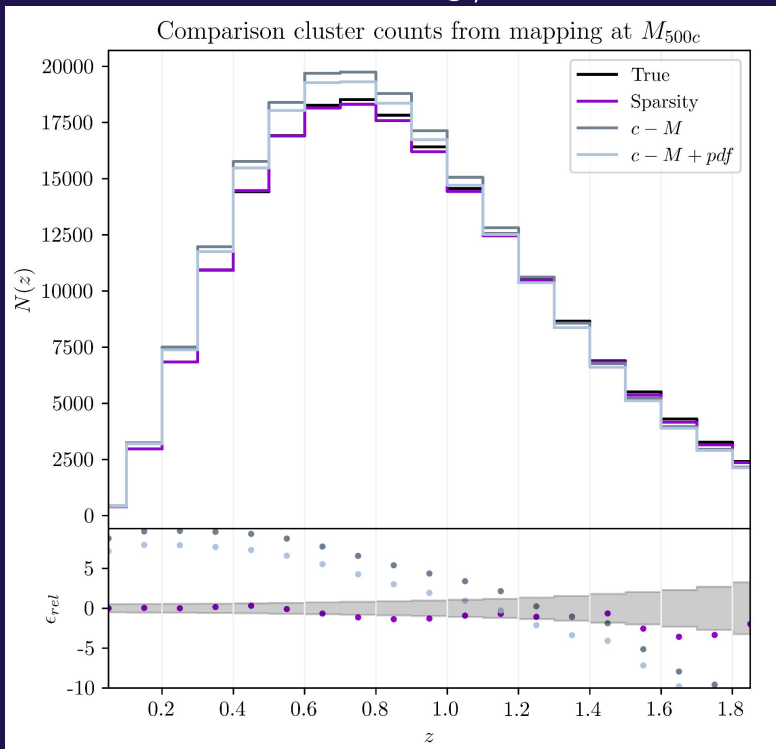
RESULTS (PRELIMINARY)

CLUSTER COUNTS - M500C

$$N_{500c}(z) = \Delta\Omega\Delta z \frac{d^2V}{d\Omega dz} \int_{M_{min}}^{M_{max}} dM \frac{dn}{dM_{500c}}(M, z)$$

$$M_{min} = 3.10^{13} M_{\odot}/h$$

$$M_{min} = 10^{14} M_{\odot}/h$$



Stochastic approach:

Maintain a good level of accuracy

$c-M$ & pdf relations:

Gives rise to a ~7-8% discrepancy

RESULTS (PRELIMINARY)

LIKELIHOOD

$$\mathcal{L}(N_{\text{obs}}|N_{\text{theory}}) = \prod_{i=1}^{N_{\text{bins}}} \frac{1}{\sqrt{2\pi\sigma_i^2}} \exp\left(-\frac{1}{2} \frac{(N_{\text{obs},i} - N_{\text{theory},i})^2}{\sigma_i^2}\right)$$

Gaussian Likelihood

RayGal data

Shot noise

Priors:

$$\sigma_8 \sim \text{Uniform}(c, d)$$

$$\Omega_m \sim \text{Uniform}(a, b)$$

$$H_0 \sim \mathcal{N}(\mu_{H_0}, \sigma_{H_0}^2)$$

$$\Omega_b \sim \mathcal{N}(\mu_{\Omega_b}, \sigma_{\Omega_b}^2)$$

N.B.

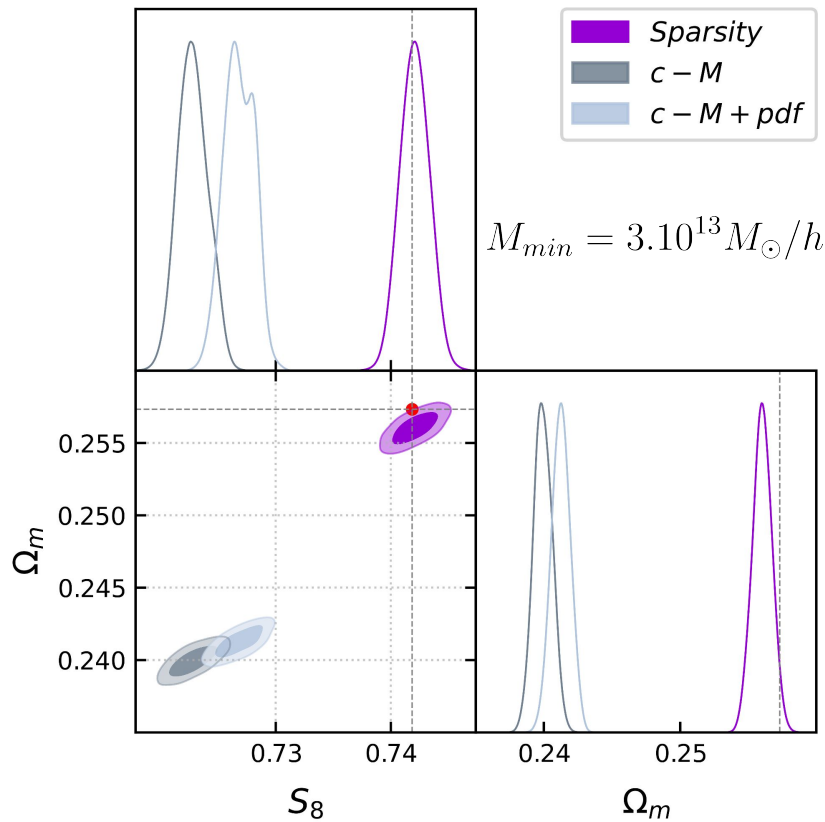
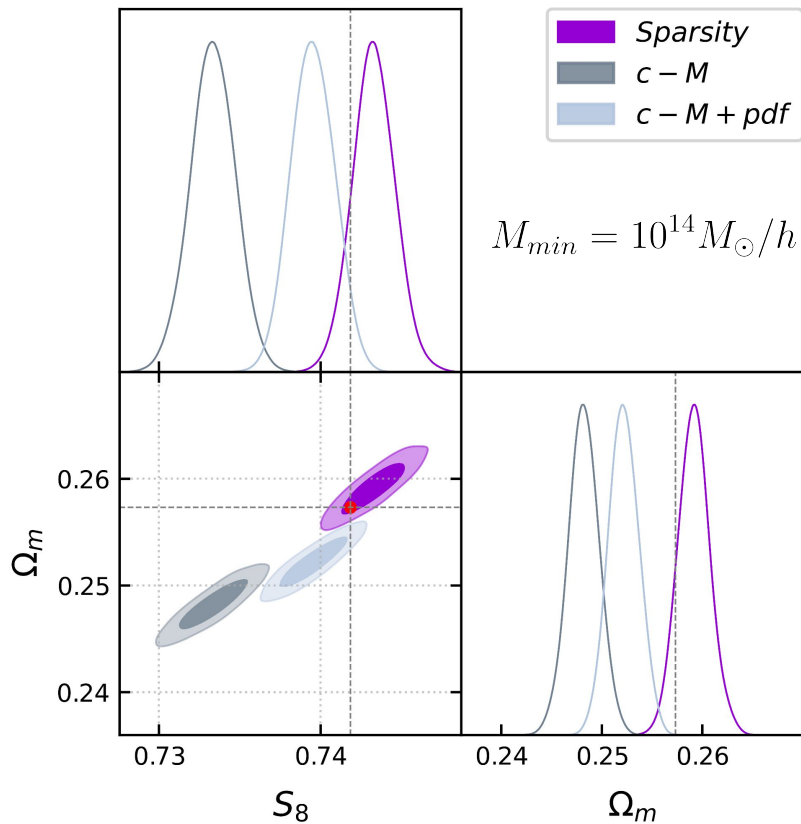
Cosmology for c-M

and p.d.fs of \mathbf{s} are fixed to RayGal cosmology

RESULTS (PRELIMINARY)

COSMOLOGICAL PARAMS

$$\Delta_c = 500$$



CONCLUSION

Deterministic (c-M)

Strong disagreement with the fiducial cosmology (for both M_{\min})

Systematic bias toward lower values of $S8$ and Ω_m

Semi-deterministic (c-M+pdf)

2σ disagreement with RayGal cosmology but a slight improvement

Stochastic (sparsities)

Very good agreement for both M_{\min} ($< 1\sigma$)

Non-parametric approach gives robust cosmological predictions in this case

PERSPECTIVES

- Use more **realistic synthetic data** set (e.g. proper light cone) → **Flagship** octant of the sky (WIDE survey)
- Add more complexity in the cluster counts (e.g. **selection function**, **mass-observable relation** etc.)
- Handle the cosmology dependence of sparsities → ***e-mantis*** emulator (Saez Casares et al. in prep)

THANK YOU FOR YOUR ATTENTION :)

c-M relation:

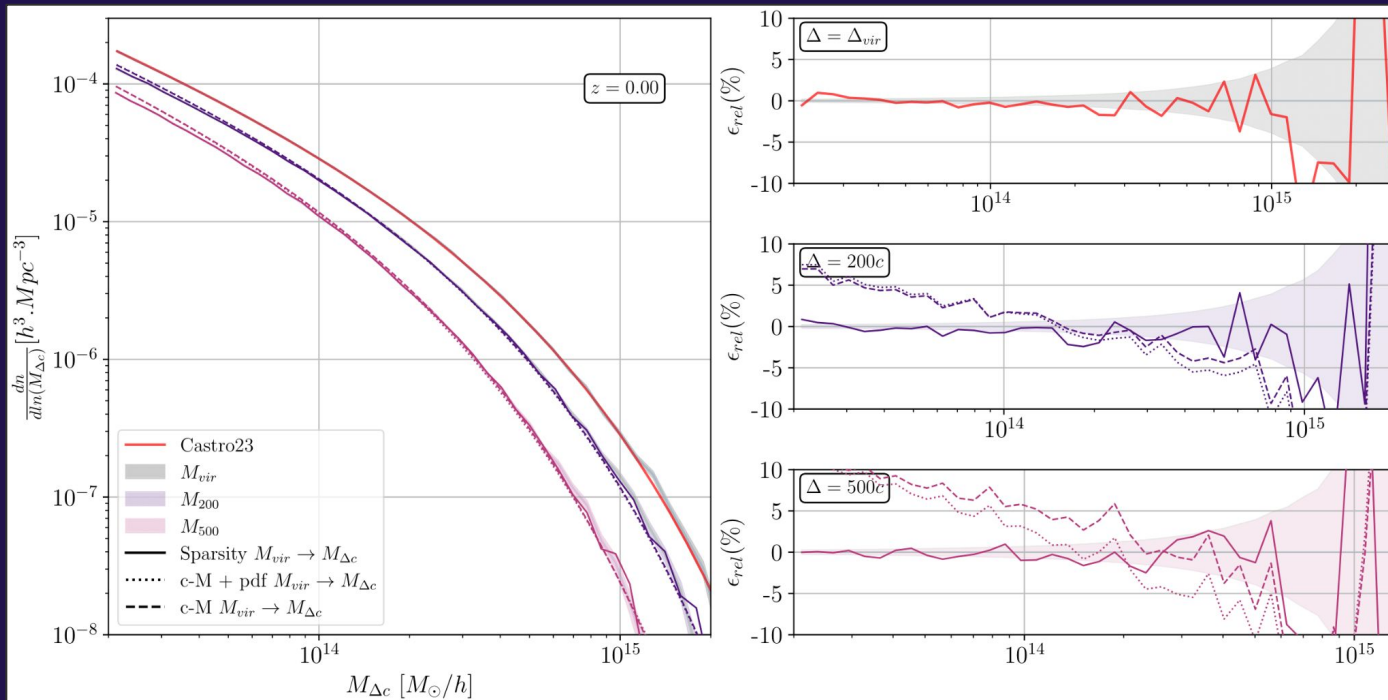
$$c = C(\alpha_{\text{eff}}) \times \tilde{G} \left(\frac{A(\alpha_{\text{eff}})}{\nu} \left[1 + \frac{\nu^2}{B(\alpha_{\text{eff}})} \right] \right) \quad \text{6 free parameters}$$

HMF:

$$\frac{dn}{d\ln(M)} = \frac{\bar{\rho}_m(0)}{M} \nu f(\nu) d\ln(\nu)$$

$$\nu f(\nu) = A(p, q) \sqrt{\frac{2a\nu^2}{\pi}} e^{-a\nu^2/2} \left(1 + \frac{1}{(a\nu^2)^p} \right) (\nu\sqrt{a})^{q-1}$$

RESCALED HMF COMPARISON WITH RAYGAL Z=0



Stochastic approach:

Maintain same level of agreement at Δ_{vir} for 200c and 500c.

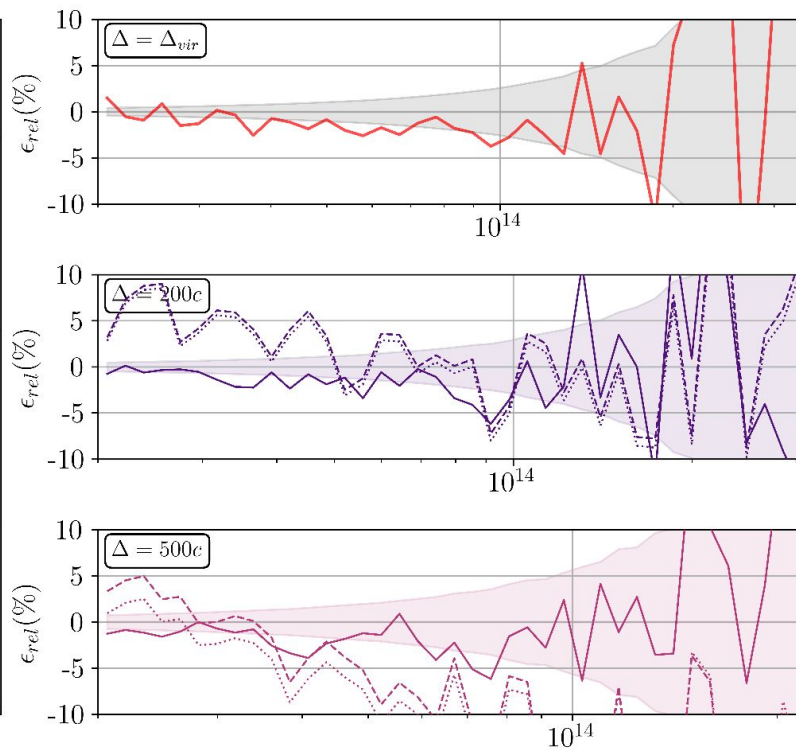
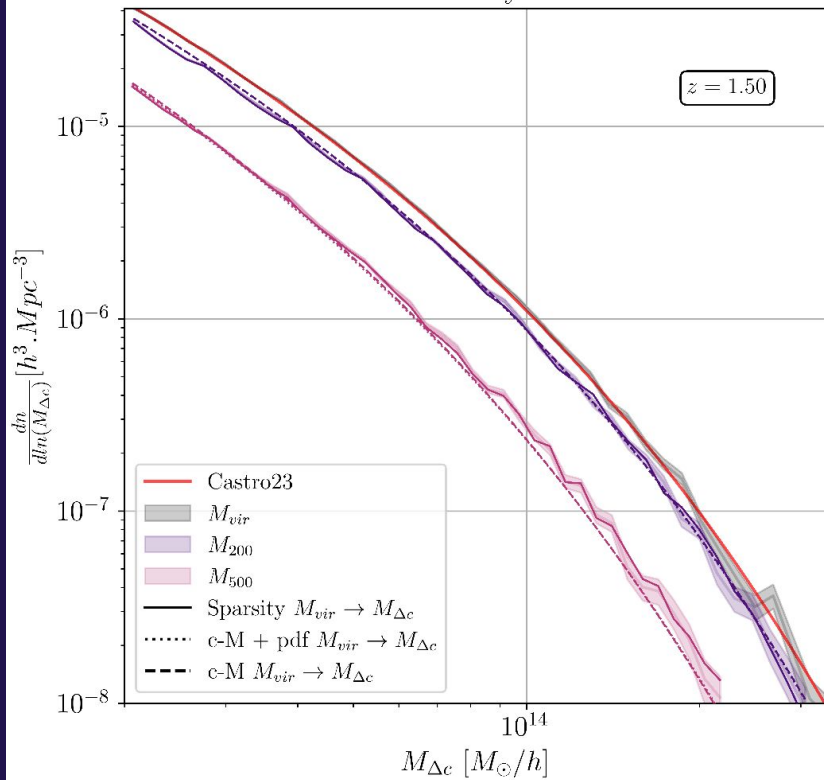
c-M & pdf relations:

Introduce discrepancy above 5% at high masses and $\sim 8\%$ at low masses even accounting for the p.d.f of c

RESCALED HMF

COMPARISON WITH RAYGAL Z=1.5

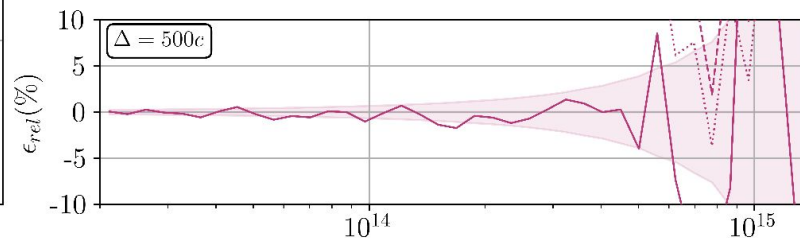
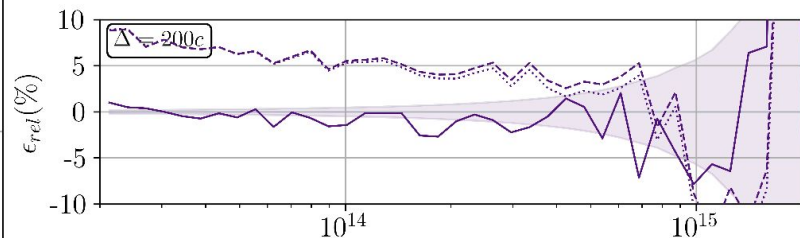
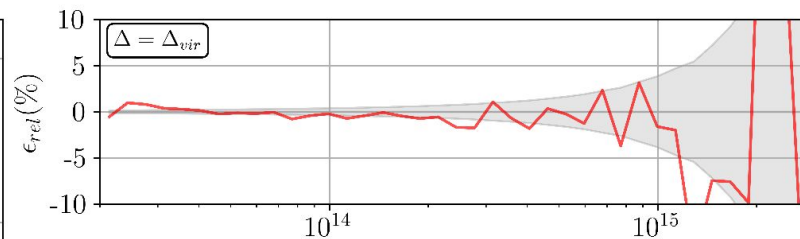
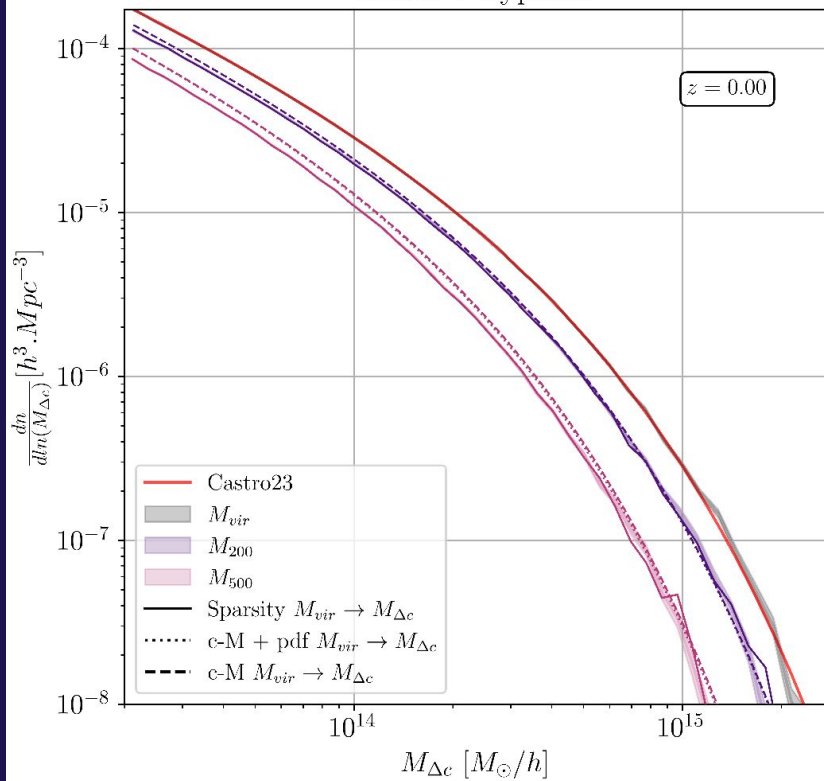
Model : ishiyama21



RESCALED HMF

KLYPIN 11

Model : klypin11



BACKUP SLIDES

TINKER 08 + DESPALI 16

$$\Delta_c = 500$$

$$M_{min} = 10^{14} M_{\odot} / h$$

M_{200}

