Multi-tracer cosmology trough the lens of SHAM

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Motivations

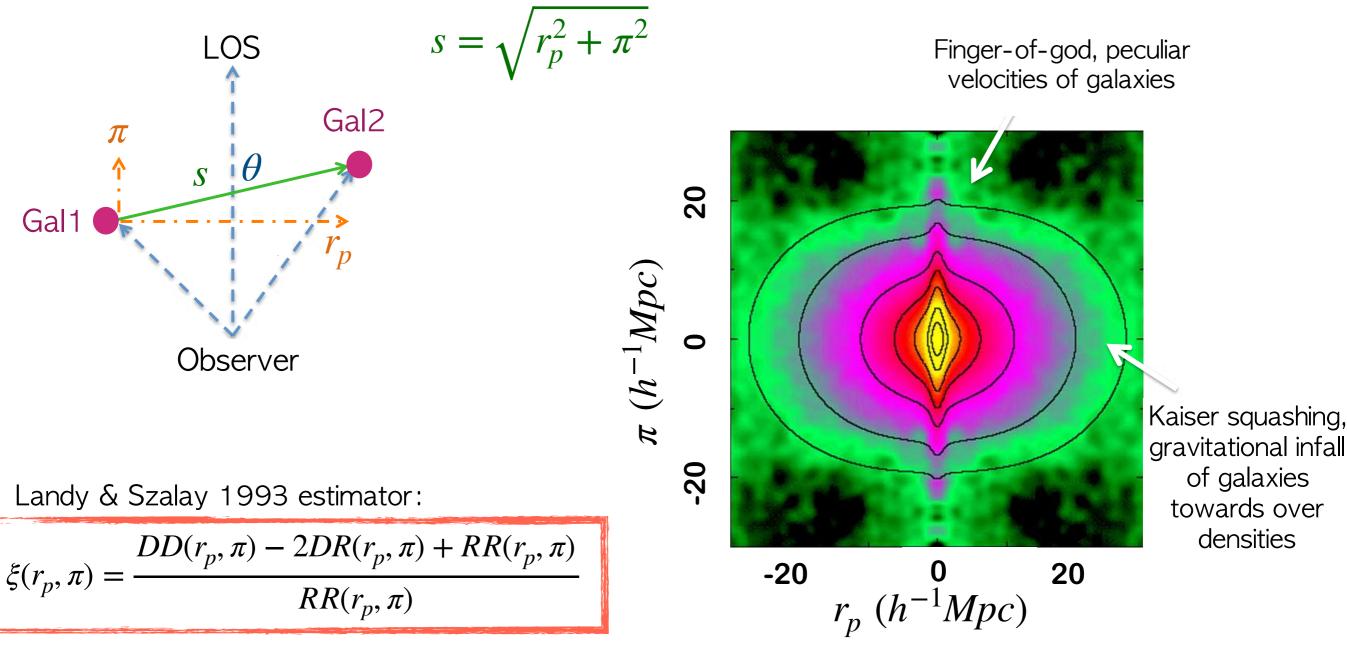
Improve the current galaxy-halo connection to be able to accurately predict the galaxy clustering signal of the multi-tracers (ELGs, LRGs, QSOs) from the upcoming surveys as a function of different baryonic properties.

In particular, we implement an extension of the Subhalo abundance matching (SHAM) model including different halo/galaxy secondary properties from the IllustrisTNG300 hydrosim to improve our knowledge of secondary halo bias and its connection with the environment (tidal tensor).

Galaxy two-point correlation function (2PCF)

$$dP = n^2 [1 + \xi(s)] dV_1 dV_2$$

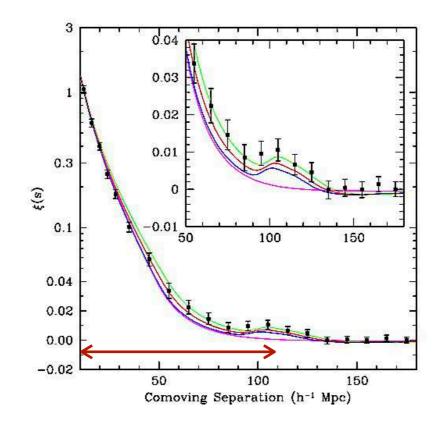
Excess probability over randoms to find 2 galaxies separated by a distance s in z-space:



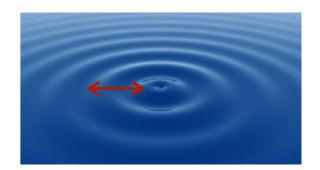
Expanding in Legendre polynomial we find the 2PCF multipoles:

$$\xi_l(s) = \frac{2l+1}{2} \int_{-1}^{+1} \xi(s,\mu) P_l(\mu) d\mu$$

I=0 monopole, spherical average I=2 quadrupole traces satellites



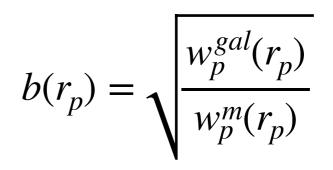
Eisenstein et al. 2005



BAO scale ~110 Mpc/h standard ruler for cosmological distances

The projected 2PCF mitigates the redshift-space distortions:

$$w_p(r_p) = 2 \int_0^\infty \xi(r_p, \pi) d\pi$$

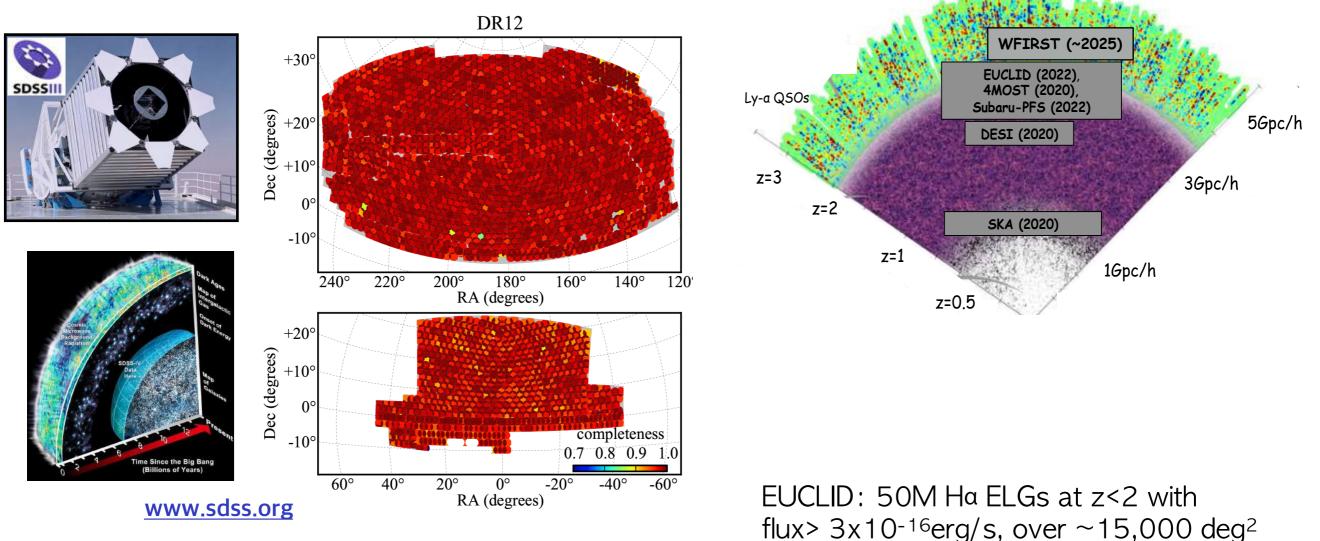


Large-scale structure surveys

Past & Ongoing

Future

SDSS-I/II (2005-09), SDSS-III/BOSS (2009-14) SDSS-IV/eBOSS (2014-20)



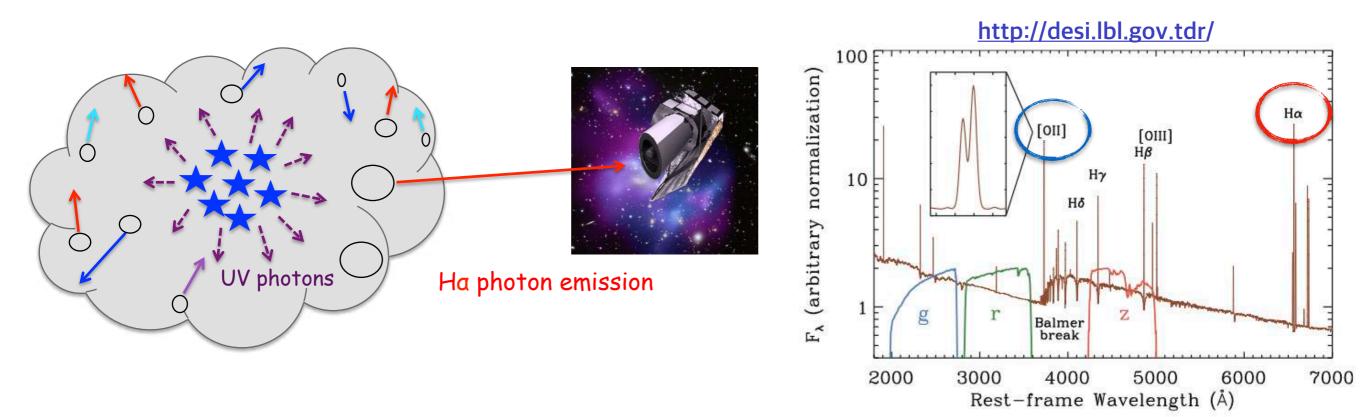
BOSS: 1.5M galaxies, mostly LRGs z<0.7, over ~10,000 deg²

eBOSS: 375k LRGs z<0.8, 260k [OII] ELGs z<1, 740k QSOs, over ~7500 deg²

multi-tracers

DESI: 10M [OII] ELGs z<1.7 over ~14,000 deg²

Emission line galaxies (ELGs)

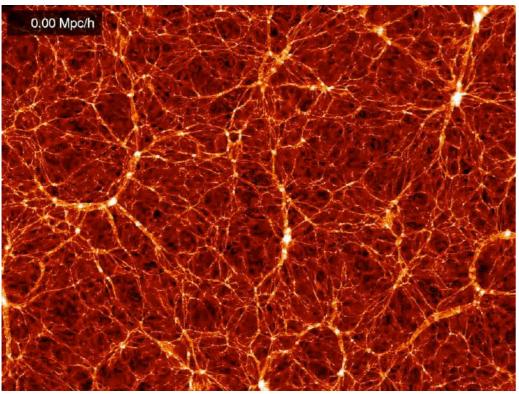


New-generation surveys will target ELGs out to $z\sim2$ to trace the BAO in their clustering, deliver 3D maps of the Universe with unprecedented accuracy, measure the growth of structure f(z), and the Universe expansion history H(z).

N-body cosmological simulations

Collisionless, cold DM particles are thrown in a cubic box with some initial conditions and let evolve under gravity. Halos are identified using halo finders: BDM, Rockstar, FOF

Expensive: solve equation of motion of N particle interacting gravitationally.

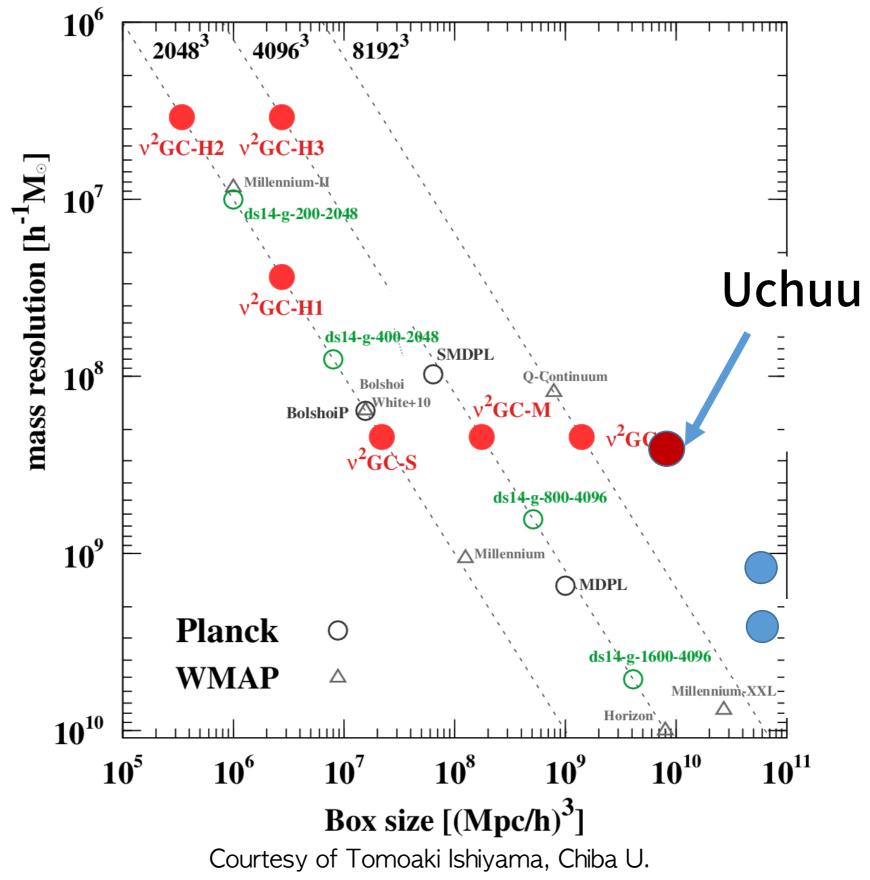


MultiDark



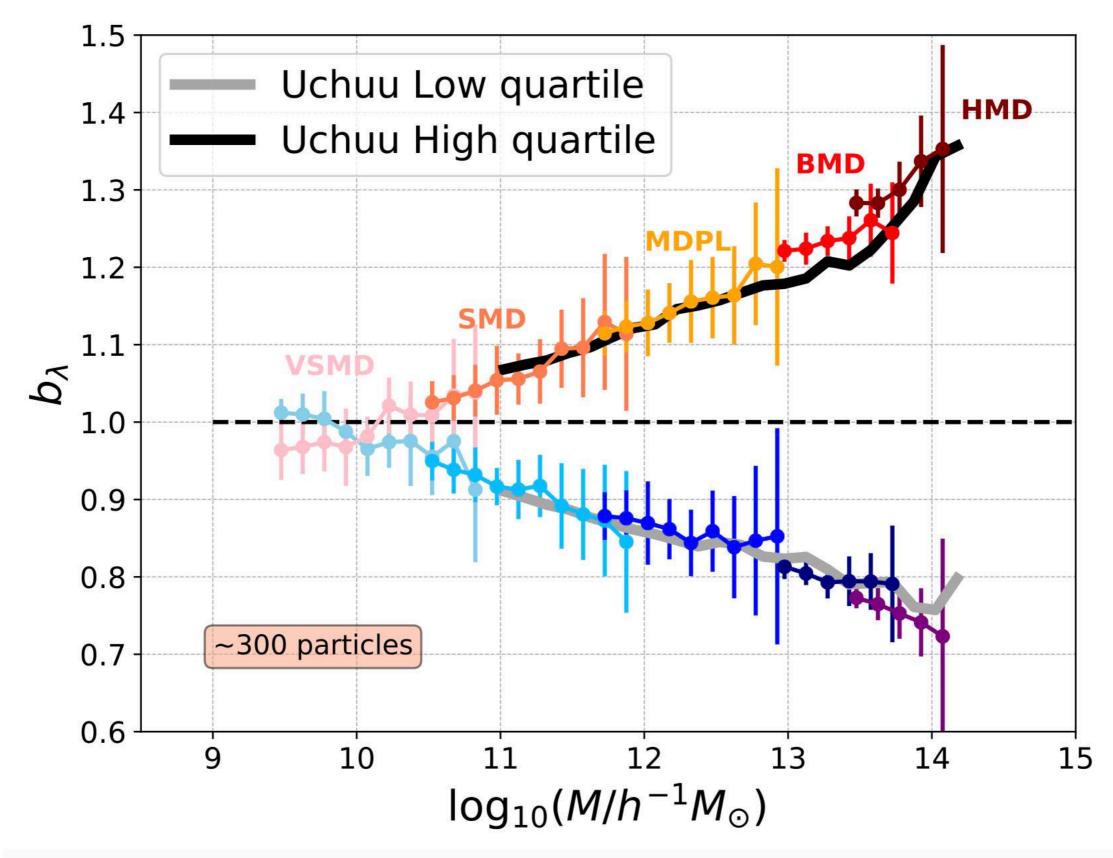
	N particles	Lbox (Mpc/h)	mass resolution (M _{sun} /h)			
BigMD MDPL2 SMD	3840 ³ 3840 ³ 3840 ³	2500 1000 400	2.36x10 ¹⁰ 1.5x10 ⁹ 9.63x10 ⁷			
cosmosim.org						

N particles 12800 ³	Lbox (Mpc/h) 2000	mass resolution (M _{sun} /h) 3.27x10 ⁸				
skiesanduniverses.org						



Uchuu high resolution will be key for resolving the small halo masses of ELGs:

Uchuu will allow us to investigate the secondary bias parameter over 4 orders of magnitude:



Courtesy of Antonio D. Montero-Dorta

The galaxy-halo connection

Galaxies are biased tracers of the underlying dark matter distribution. We populate DM haloes with galaxies using 2 methods:

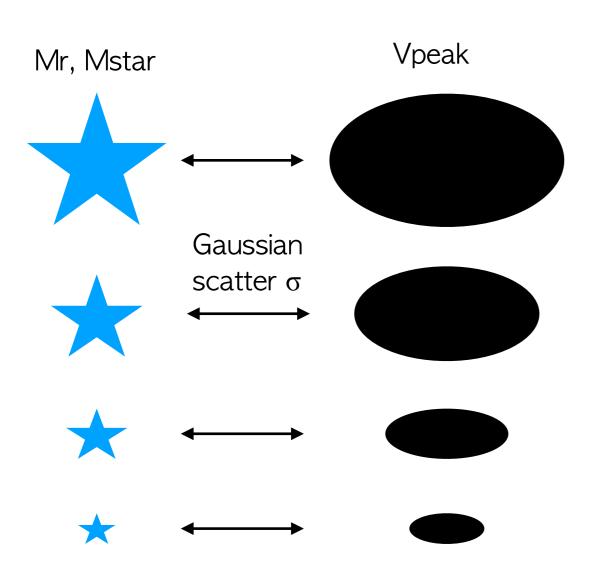
I. Halo Occupation Distribution (HOD)

Cooray+02; Berlin & Weinberg+02; Kravtsov+04; Zheng+05,07

II. SubHalo Abundance Matching (SHAM) Conroy+06, 09; Behroozi+10; Trujillo-Gomez+11

SubHalo Abundance Matching (SHAM)

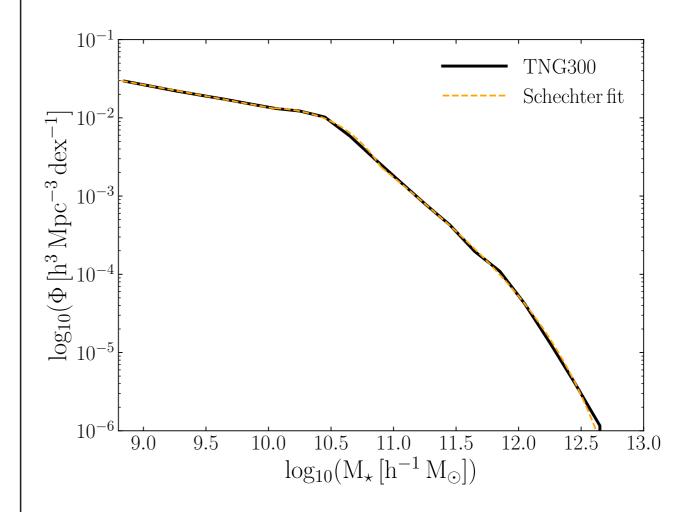
Rank order haloes and galaxies using their proxies:



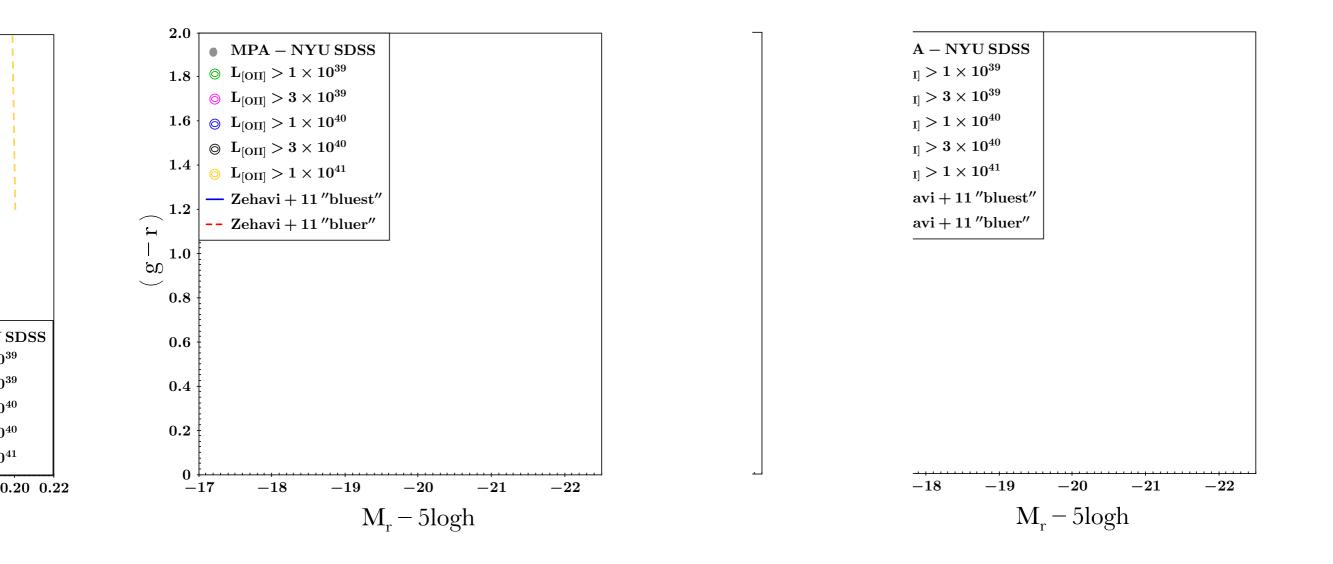
Basic assumption: more massive/luminous galaxies occupy more massive haloes

Sample Mr/Mstar from the observed cumulative stellar mass/luminosity function until the observed number density n(z) is reached.

$$n_{\text{gal}}(>M_{\star}) = n_{\text{h}}(>V_{\text{peak}})$$



Standard SHAM works well only for complete galaxy samples

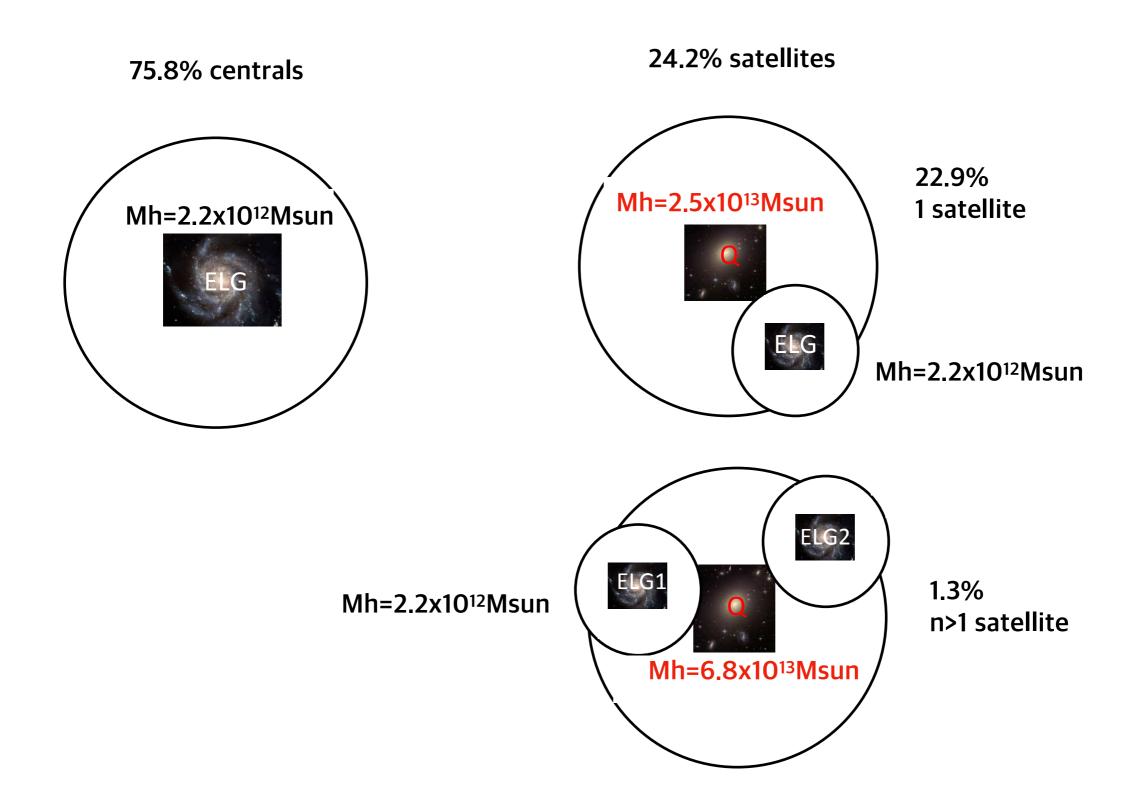


			typical V_{peak} [km s ⁻¹]	typical M_h $[h^{-1}M_{\odot}]$
			127 ± 58 177 ± 53 201 ± 86	$(3.17\pm0.19)\times10^{11}$ $(6.64\pm0.41)\times10^{11}$ $(1.54\pm0.09)\times10^{12}$
0.17	$\begin{array}{c} 3 \times 10 \\ 1 \times 10^{41} \end{array}$	19.4 ± 0.4 18.0 ± 0.5	$283 \pm 117 \\ 341 \pm 140$	$(2.92\pm0.18)\times10^{12}$ $(5.49\pm0.34)\times10^{12}$

On average:

$$M_{halo} = (2.2 \pm 0.1) \times 10^{12} \, h^{-1} M_{\odot}$$
$$f_{sat} = (24.2 \pm 0.4) \,\%$$

[OII] ELG halo occupation distribution at $z\sim0.1$:



SHAM + Age Matching in IllustrisTNG300

Age-Matching (Hearin et al 2013): SHAM + secondary matching between galaxy color (g-r) and halo Zstarve.

We implement AM using different halo/galaxy secondary properties from IllustrisTNG300 to improve our knowledge of secondary halo bias and its connection with the environment (tidal tensor).



$L_{box} = 205 \text{ Mpc/h}$

 $M_{DM} = 5.9 \times 10^7 M_{sun}$

 $N_{DM} = 2500^3$

 $M_{gas} = 1.7 \times 10^7 M_{sun}$

TNG300 GALAXIES -> Hydrosim

	l: Mstar				
11:	color	sSFR	size		

OUR DATA

TNGDMO300 HALOES -> DM-only

	I:VI	peak		+ SHAM - AM
11:	Z _{starve}	Cinfall	δ ^{env}	
• • •	Tidal dens	ity & anis	sotropy	

Halo tidal properties

Compute the halo tidal tensor on a cubic lattice using the SPIDER code (Martizzi et al. 2019; https://github.com/ dmartizzi/spider-public):

 $T_{ij}(\vec{x}) = \partial_i \partial_j \psi_R(\vec{x})$ gravitational potential smoothed at R

Poisson: $\nabla^2 \psi_R(\vec{x}) = \delta_R(\vec{x})$ smoothed density contrast

In Fourier space: $\delta_R(\vec{k}) = \delta(\vec{k})e^{-k^2R^2/2}$ Gaussian smoothing

We interpolate the tensor at each halo location (x, y, z) and the smoothing scale R at the halo radius to create a halo-by-halo catalogue of tidal estimates.

Diagonalize it and extract the eigenvalues λ_1 , λ_2 , λ_3 that give us a LSS classification:

knots:	$\lambda_{i,j,k} \ge 0.3$
filaments:	$\lambda_{i,j} \ge 0.3$
sheets:	$\lambda_i \ge 0.3$
voids:	$\lambda_{i,j,k} < 0.3$

Halo tidal overdensity:

$$\delta_{\rm R} = \lambda_1 + \lambda_2 + \lambda_3$$

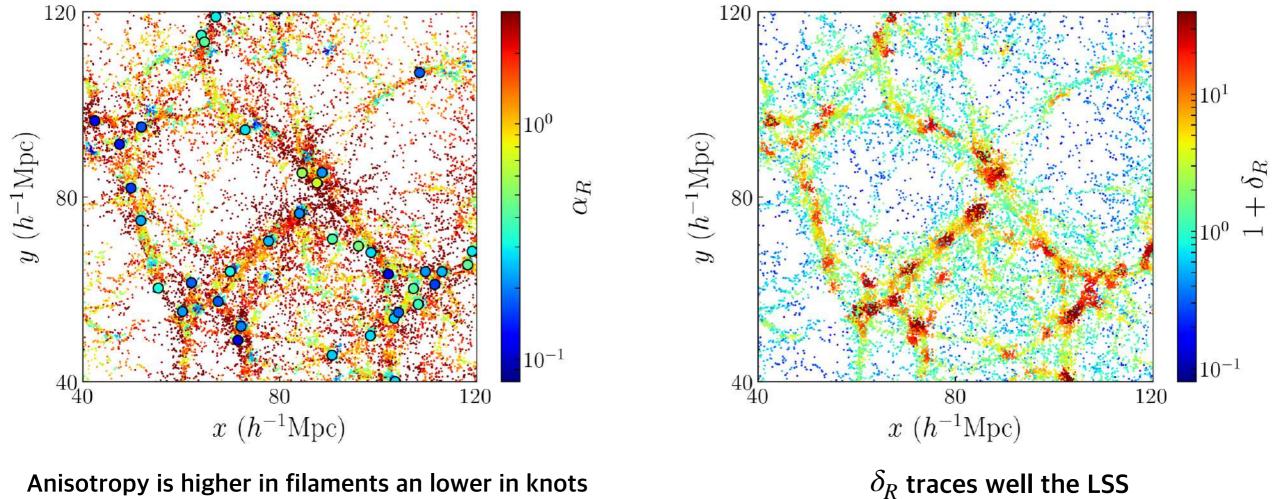
Tidal anisotropy parameter:

$$\alpha_{\rm R} \equiv (1 + \delta_{\rm R})^{-1} \sqrt{q^2}$$

$$q^{2} = \frac{1}{2} \left[(\lambda_{2} - \lambda_{1})^{2} + (\lambda_{3} - \lambda_{1})^{2} + (\lambda_{3} - \lambda_{2})^{2} \right]$$

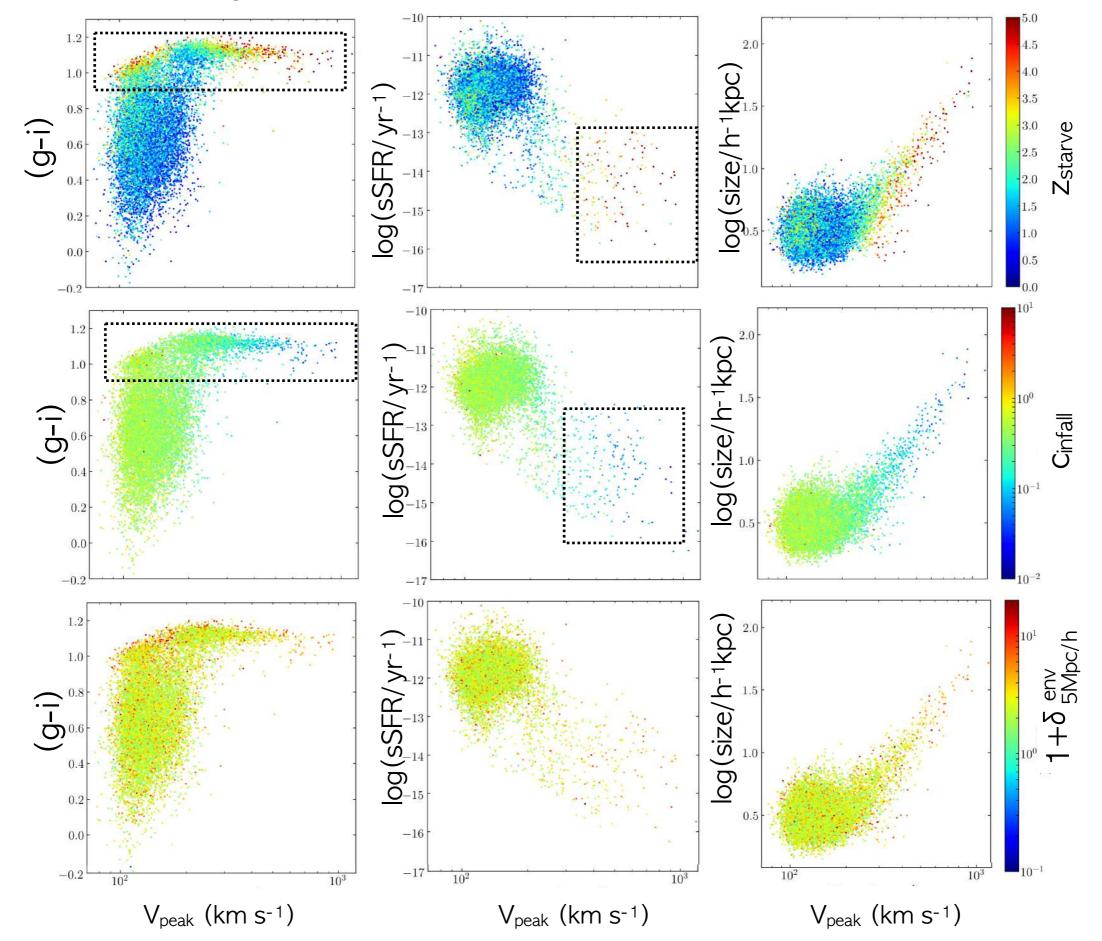
Paranjape et al. 2018; Ramakrishnan et al. 2019

The anisotropy parameter is a mediator between the internal and large-scale properties of haloes

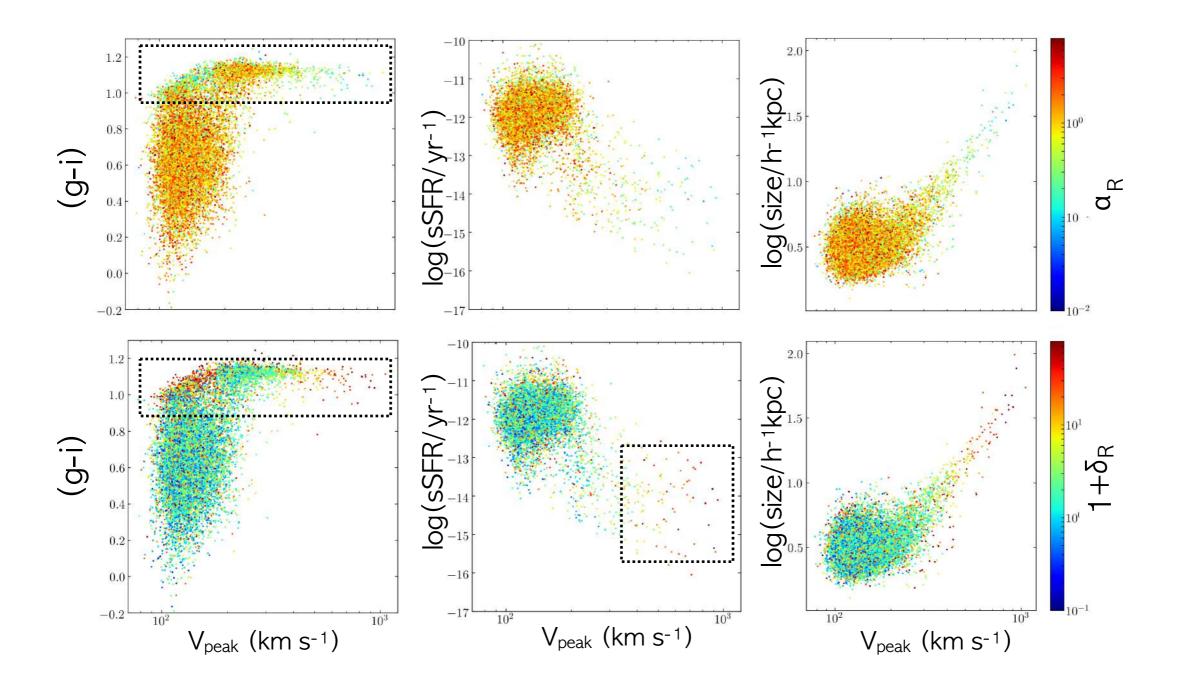


Anisotropy is higher in filaments an lower in knots

Correlations between halo/galaxy secondary properties:

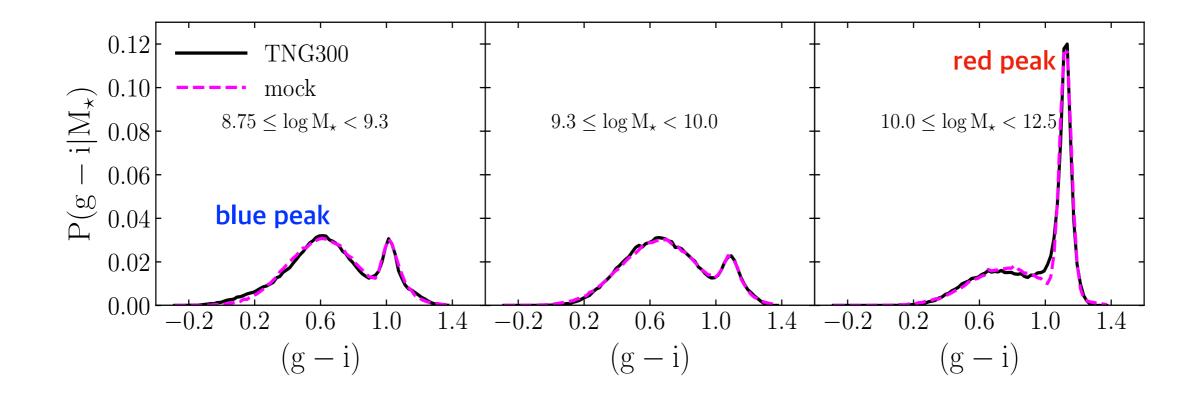


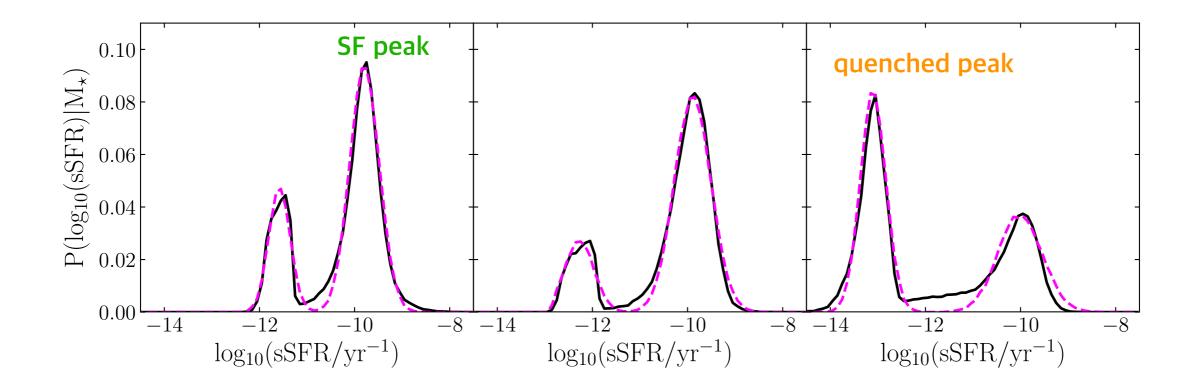
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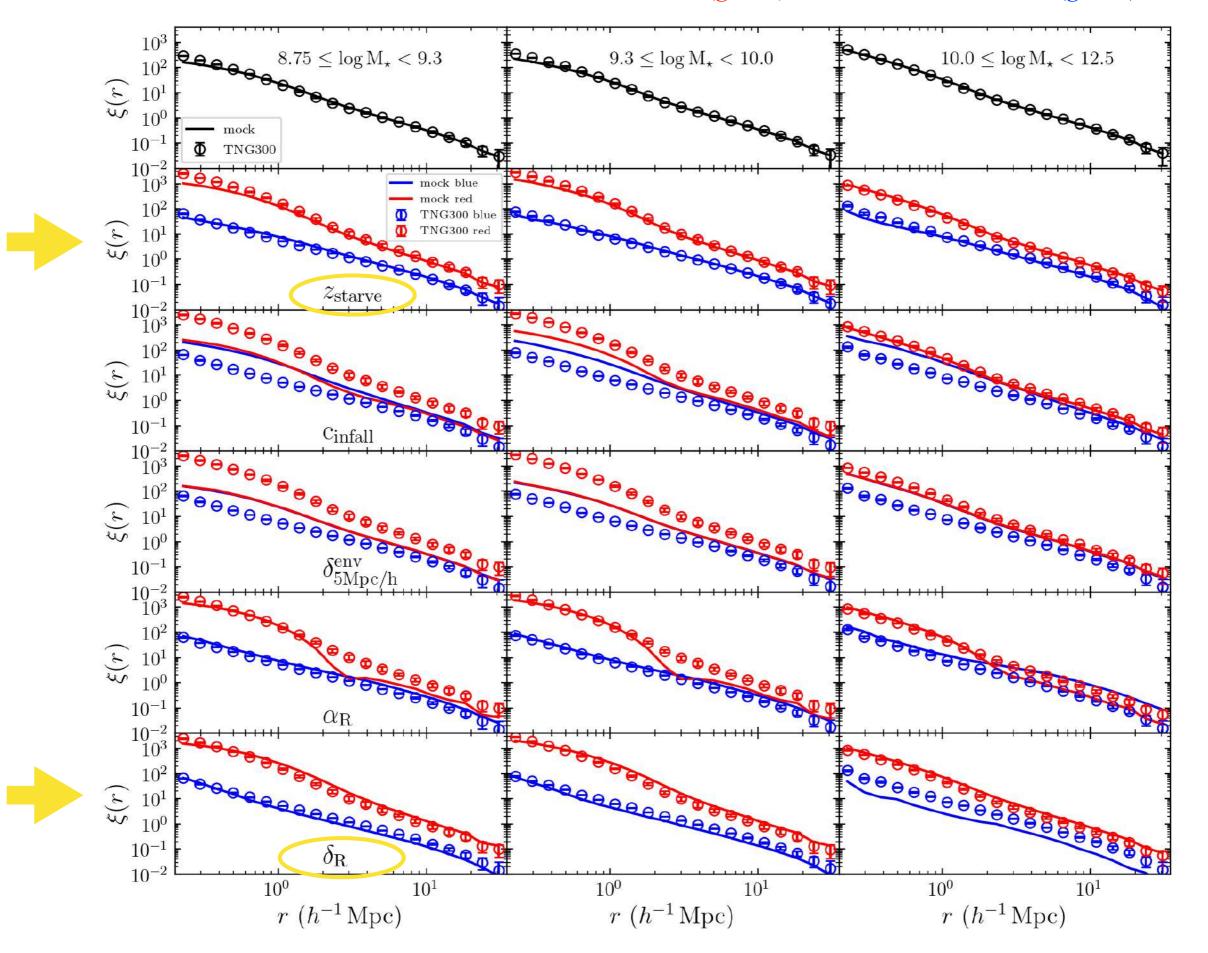


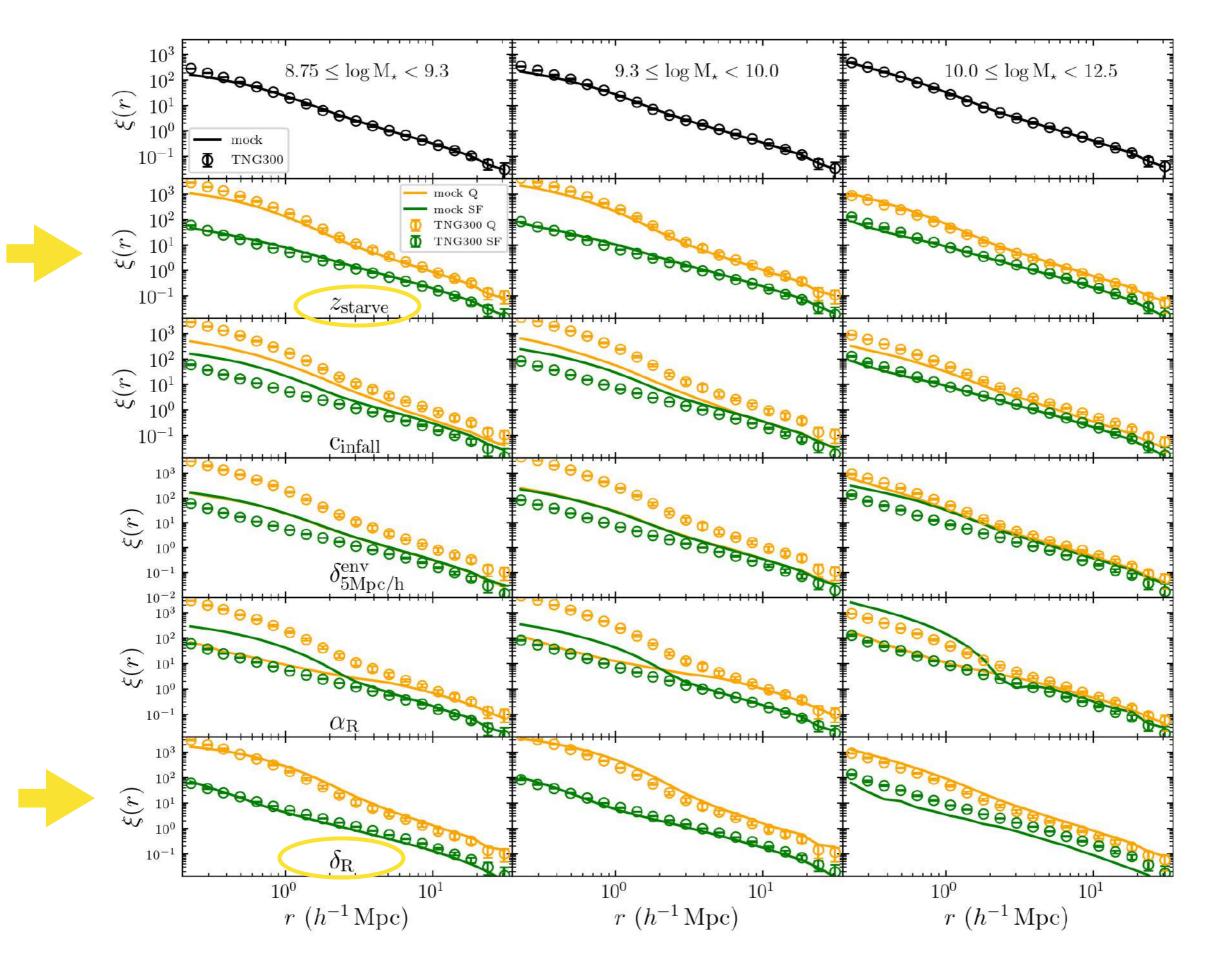
Even if apparently mild, these correlations are the drivers to obtain a good secondary matching

The secondary matching is performed through conditional PDFs in bins of stellar mass:







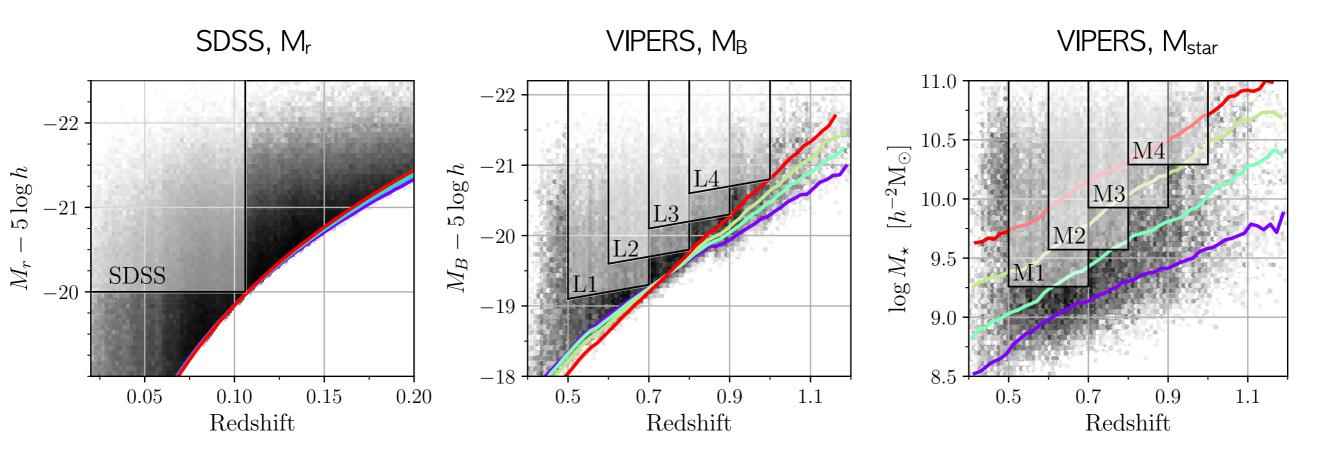


Summary Part I

- SHAM is a simple, yet powerful, prescription able to place galaxies into their host DM haloes reproducing the clustering of ~ complete samples.
- DESI, Euclid, LSST, etc ... will target millions of galaxy multi-tracers, most of them incomplete in many parameter spaces. Therefore, a sophisticated modification of SHAM is needed.

- SHAM+Age Matching links the internal and large-scale galaxy/halo properties allowing us to model the clustering of multi-tracers on all scales, properly including the secondary halo bias and the physics of galaxy formation/evolution.
- The accuracy of this method strongly depends on the halo/galaxy secondary proxies chosen. The more they correlate, the better is the secondary matching, the higher the accuracy in the 2PCF modelling. Among the proxies we have tested the best are: galaxy color, sSFR and halo z_{starve} , δ_R

Constraining the growth history $\sigma_{s}(z)$ using basic SHAM

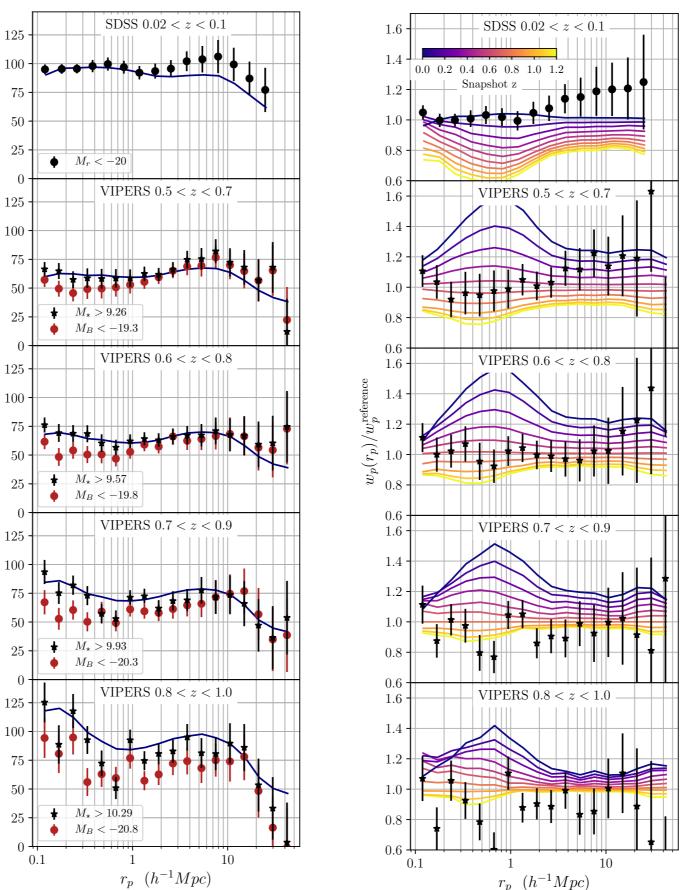


Granett, Favole, Montero-Dorta et al. 2019

Sample	Redshift	Mean z	Threshold	Count	Volume $10^6 h^{-3} \text{Mpc}^3$	Density $10^{-3}h^3$ Mpc ⁻³
SDSS	0.020 < z < 0.106	0.063	$M_r < -20.0$	117959	21.90	5.85
L1 M1	0.5 < z < 0.7 0.5 < z < 0.7	$\begin{array}{c} 0.61\\ 0.61\end{array}$	$\begin{split} M_B &< -19.3 + (0.7 - z) \\ \log M_\star &> 9.26 h^{-2} \mathrm{M}_\odot \end{split}$	$23352 \\ 22508$	$\begin{array}{c} 4.93\\ 4.93\end{array}$	11.8 11.8
L2 M2	0.6 < z < 0.8 0.6 < z < 0.8	$0.70 \\ 0.70$	$M_B < -19.8 + (0.8 - z)$ $\log M_{\star} > 9.57 h^{-2} M_{\odot}$	$20579 \\ 19577$	$5.98 \\ 5.98$	8.57 8.57
L3 M3	0.7 < z < 0.9 0.7 < z < 0.9	$0.80 \\ 0.80$	$M_B < -20.3 + (0.9 - z)$ $\log M_{\star} > 9.93 h^{-2} M_{\odot}$	$13046 \\ 12270$	$6.96 \\ 6.96$	4.79 4.79
L4 M4	0.8 < z < 1.0 0.8 < z < 1.0	$0.90 \\ 0.89$	$M_B < -20.8 + (1.0 - z)$ log $M_{\star} > 10.29 h^{-2} M_{\odot}$	$\begin{array}{c} 6305\\ 5881 \end{array}$	7.86 7.86	2.13 2.13

We use basic SHAM with no parameters to avoid limiting the cosmological interpretation Small MultiDark (L=400 Mpc/h) simulation with Planck15 fiducial cosmology

125100755025ſ $(h^{-1}Mpc)^{1.8}$ 125M_B selected 2PCFs 100 are lower since the bluer 75 $r_p^{0.8} w_p(r_p)$ **VIPERS** rest-frame is 50more sensitive to 25recent SF activity and 125less to total mass 100 75

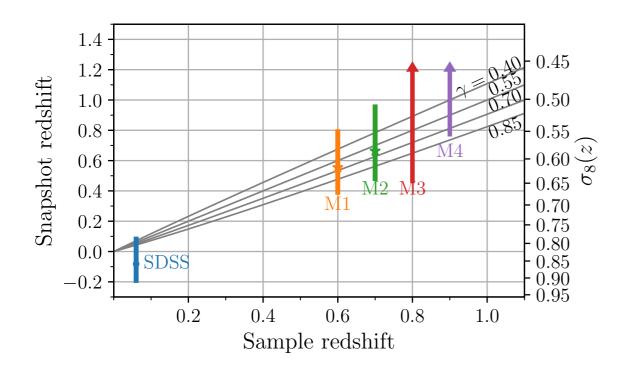


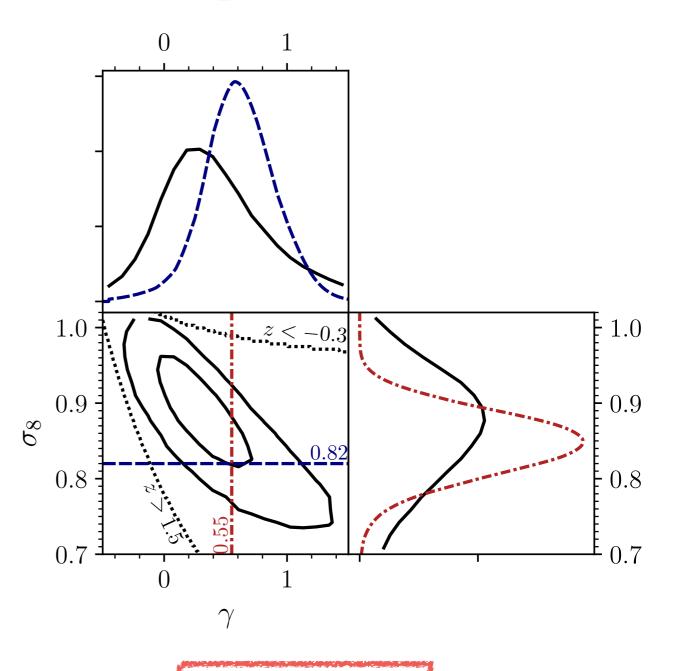
Granett, Favole, Montero-Dorta et al. 2019

The growth history $\sigma_8(z)$ can be parametrised in terms of the growth index γ as:

$$\sigma_8(z) = \sigma_8(0) \exp\left[-\int_0^z \Omega_m(z)^{\gamma} dln(1+z')\right]$$

 $\sigma_8(z)$ in the fiducial cosmology is fixed at the simulation snapshot. So $w_p(r_p)$ measured in the simulation implicitly gives us $w_p(\sigma_8)$ and, by minimising the likelihood, we infer $\sigma_8(w_p^{obs})$.





 $\gamma = 0.6 \pm 0.3$

 $\sigma_8 = 0.85 \pm 0.04$

We fix $\sigma_8(0) = 0.82$ from the fiducial cosmology and get:

We fix $\gamma = 0.55$ from the fiducial cosmology and get:

Summary Part II

• Our results can be considered model dependent as they rely on SHAM

But they are not calibrated on CMB, which tightens the errors

Our errors are better than VIPERS estimates from RSD only

- With future surveys we will measure the growth history at low z with no need of CMB
- Next step is to constrain $\sigma_8(z)$ using SHAM + Age Matching
- Future: comprehensive clustering + weak lensing model able to accurately predict both observables for galaxy multi-tracers