

### Voronoi Volume Function

A new probe of cosmology & galaxy evolution

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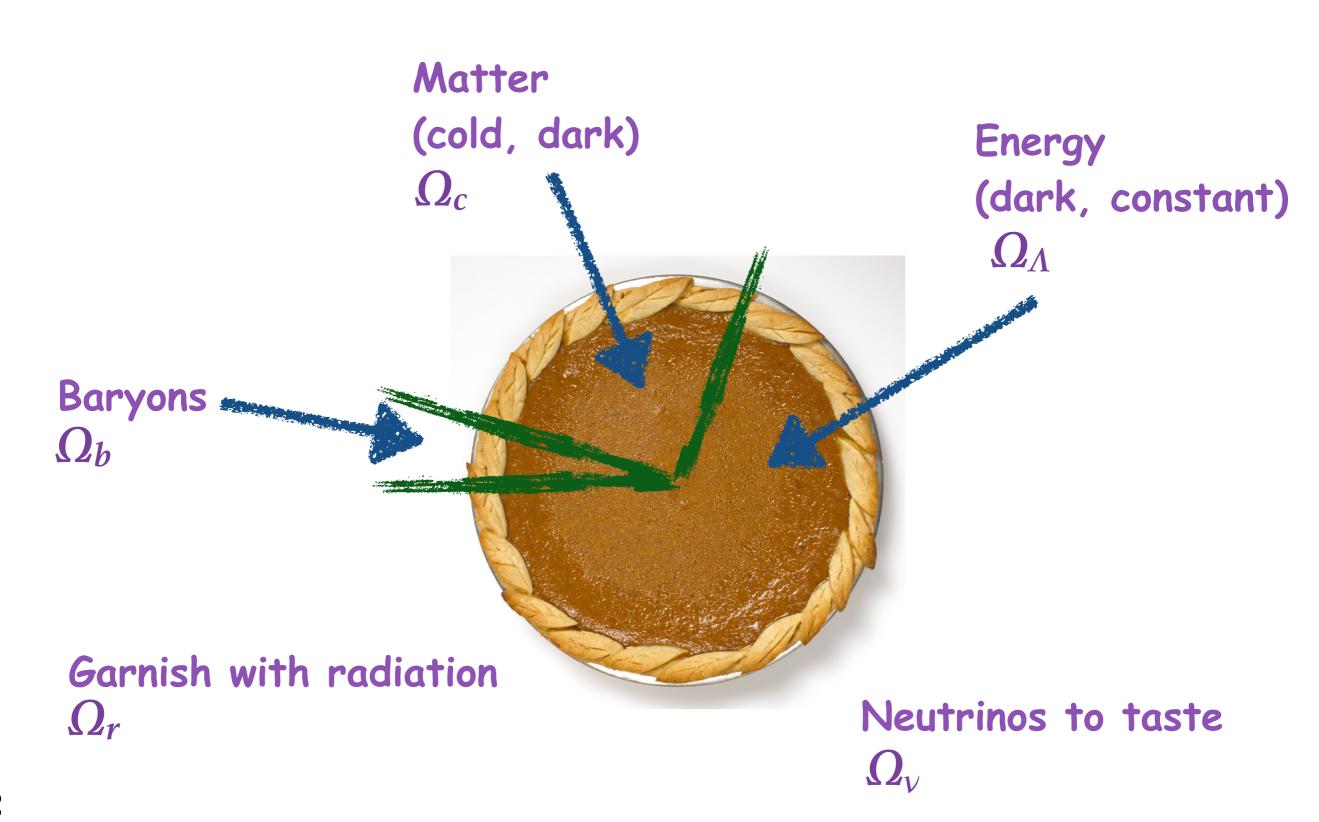


### **Outline**

- ◆ Introduction
  - Development of the Cosmic Web (non-Gaussian spatial distribution of matter)
- ◆ Information beyond 2-pt statistics
  - Voronoi tessellation
  - Voronoi volume function (VVF): definition and analytical expectations
  - Simulations: VVF dependence on halo mass, large-scale clustering, redshift, nature of dark matter, RSD, substructure
- ◆ Preliminary comparisons with observations

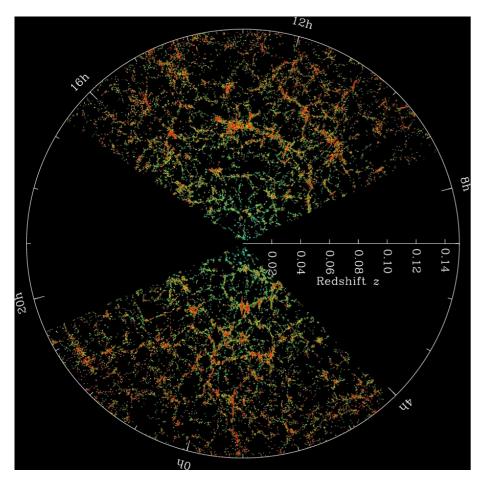


# **Standard Model of Cosmology**

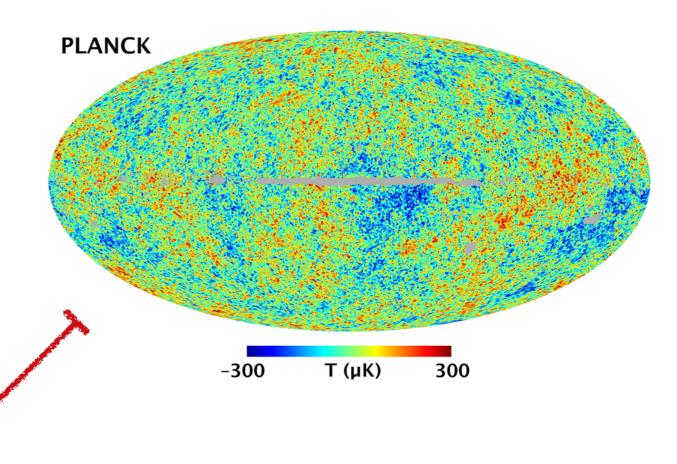


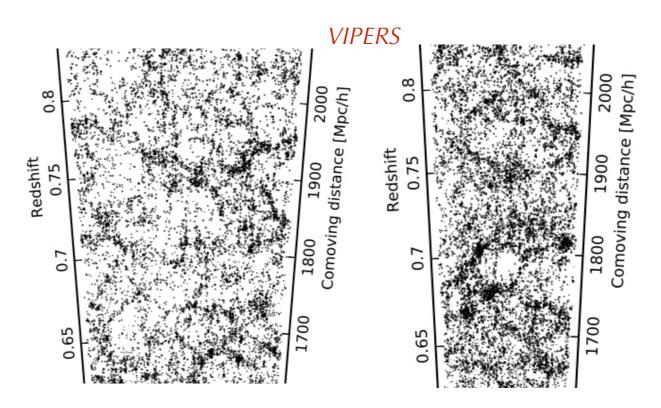


## **Growth of Structure**



Sloan Digital Sky Survey





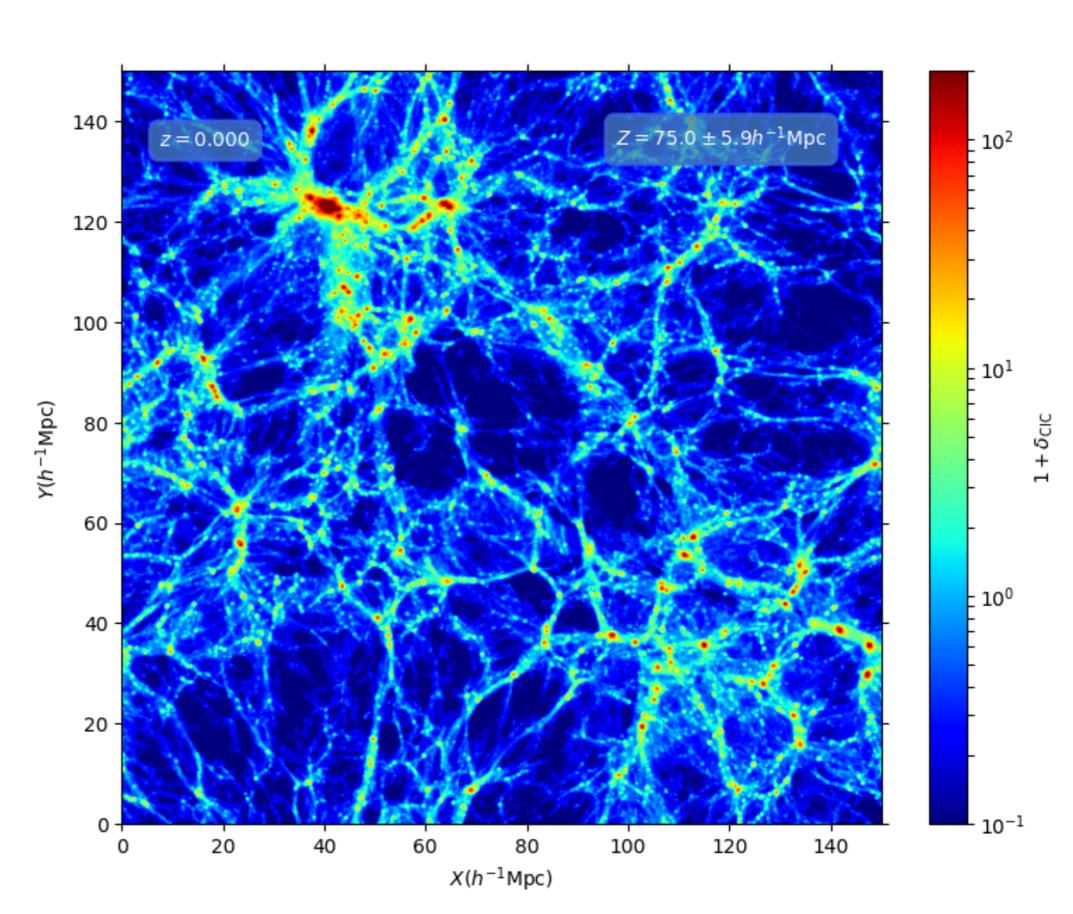


### **Growth of Structure**

Standard  $\Lambda$ CDM cosmology.

Collisionless cold dark matter.

Newtonian self-gravity.





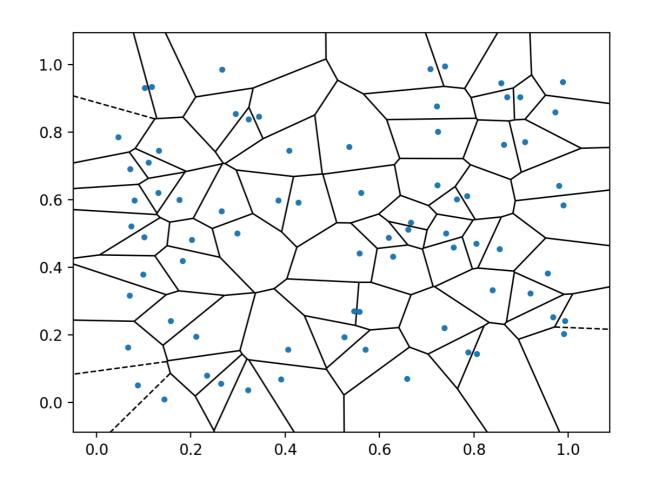
#### Voronoi tessellation

Given  $N_{\text{trc}}$  tracers at positions  $\{\mathbf{x}_t\}$  with  $1 \le t \le N_{\text{trc}}$ , the Voronoi tessellation is a partition of space into  $N_{\text{trc}}$  cells  $\{\mathscr{C}_t\}$  such that, for a given tracer t,  $\mathscr{C}_t$  is the set of points closer to t than to any other t'.

- Used in various fields such as meteorology, epidemiology, geophysics, computational fluid dynamics (e.g. AREPO) etc.
- Several applications in cosmology too. E.g., cosmic web classification, void identification, etc.

#### This talk:

Volume function of Voronoi cells of 3d clustered tracers



SciPy implementation of Voronoi tessellation of uniformly distributed 2d tracers.



### Voronoi tessellation



WMAP7  $\Lambda$ CDM

#### tracers:

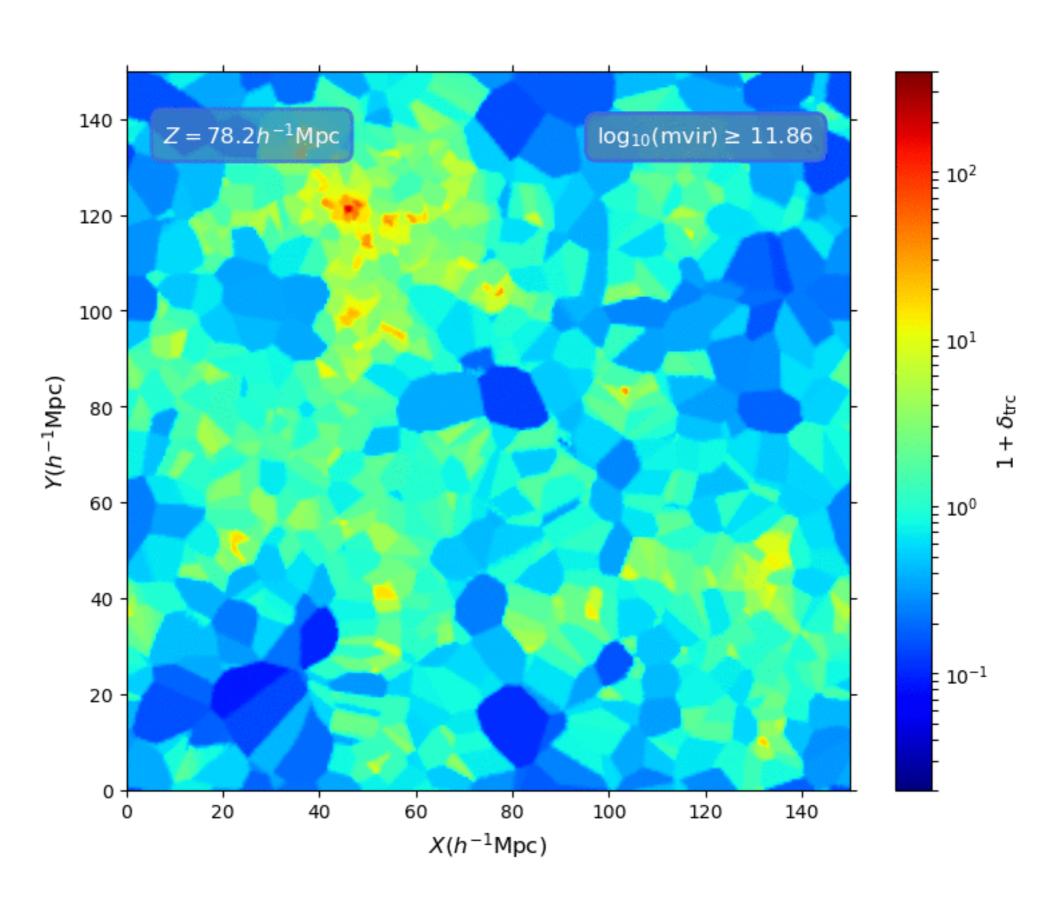
mass-thresholded haloes at z = 0

#### tessellation:

Monte Carlo algorithm

#### colour:

$$1 + \delta_{\text{trc}} = (n_{\text{trc}}V)^{-1}$$





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### Voronoi volume function

#### **Definitions**

If V(t) is volume of cell  $\mathcal{C}_t$  containing tracer t then

$$\langle V \rangle = \frac{1}{N_{\text{trc}}} \sum_{t} V(t) = \frac{V_{\text{tot}}}{N_{\text{trc}}} = n_{\text{trc}}^{-1}$$

Define

$$y \equiv V/\langle V \rangle = n_{\rm trc} V$$

We will denote the probability distribution p(y) as the Voronoi volume function (VVF).

Clearly 
$$\langle y \rangle = \int dy \, p(y) \, y = 1.$$



### Voronoi volume function

Uniformly distributed (Poisson) tracers

For uniformly distributed (Poisson) tracers,  $\langle y^2 \rangle$  is known analytically [Gilbert 1962]

$$\langle y^2 \rangle_{\text{Poisson}} = \frac{8\pi^2}{3} \int_0^\infty dz \, z^2 \int_{-1}^1 d\mu \frac{1}{v(z,\mu)^2} \simeq 1.179$$

where 
$$v(z, \mu) = \frac{\pi}{3} \left[ 2z^3 + 3\mu z(z^2 + 1) - (3\mu^2 z^2 + 1)z + 3(1 - \mu z)T + 2T^{3/2} \right]$$
  
with  $T = \left| z^2 + 1 - 2\mu z \right|$ .

Although p(y) is not known analytically, accurate fitting functions exist, e.g.:

$$p_{\text{Poisson}}(y) = \frac{c b^{a/c}}{\Gamma(b/c)} y^{a-1} \exp(-by^c)$$

with a = 4.8065, b = 4.06342, c = 1.16391 [Tanemura 2003]



### **Voronoi volume function**

#### Clustered tracers

For **clustered** tracers, generalising [Gilbert 1962] we can write

$$\langle y^2 \rangle = \frac{8\pi^2}{3} \int_0^\infty dz \, z^2 \int_{-1}^1 d\mu \frac{1}{v(z,\mu)^2} n_{\text{trc}}^2 \int_0^\infty dV_U V_U \exp\left(W_0(n_{\text{trc}}, V_U)\right)$$

where  $\exp(W_0(n_{tre}, V))$  is the void probability function for the volume V [White 1979]:

$$W_0(n_{\text{trc}}, V) = \sum_{k=1}^{\infty} \frac{\left(-n_{\text{trc}}V\right)^k}{k!} \bar{\xi}_k(V) \equiv \left(-n_{\text{trc}}V\right) \chi(n_{\text{trc}}, V)$$

where  $\bar{\xi}_k(V)$  is the connected k-point correlation function averaged over V.

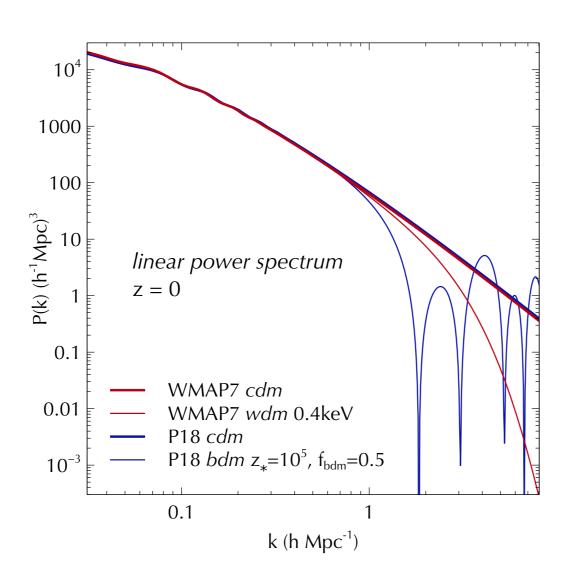
Thus  $\langle y^2 \rangle$  depends on the infinite hierarchy of tracer correlation functions.

[For Poisson distributed tracers,  $\chi(n_{\rm trc}, V) = 1$  and we recover  $\langle y^2 \rangle_{\rm Poisson} \simeq 1.179$ .]



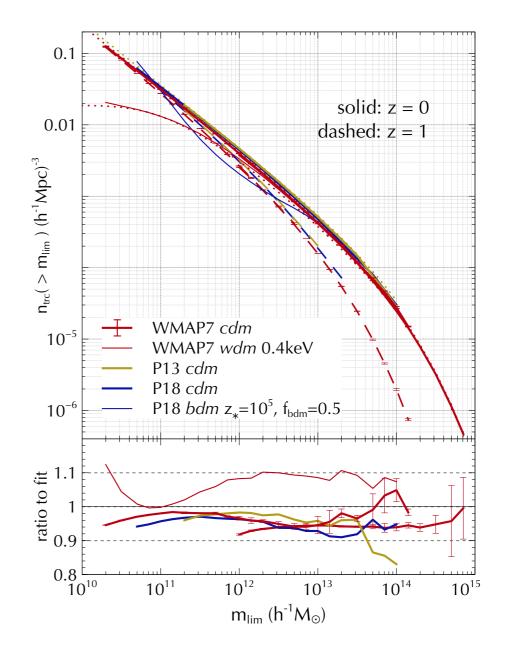
### **Simulation suite**

### Gravity-only GADGET-2 N-body runs



#### **Cold Dark Matter**

**P13** ( $\Omega_{\rm m} = 0.315, h = 0.673, \sigma_8 = 0.829$ ) L150\_N512 **WMAP7** ( $\Omega_{\rm m} = 0.276, h = 0.7, \sigma_8 = 0.811$ ) L150\_N1024, L300\_N1024, L600\_N1024 **P18** ( $\Omega_{\rm m} = 0.306, h = 0.678, \sigma_8 = 0.815$ ) L200\_N1024



#### **Alternate Dark Matter** (paired to CDM)

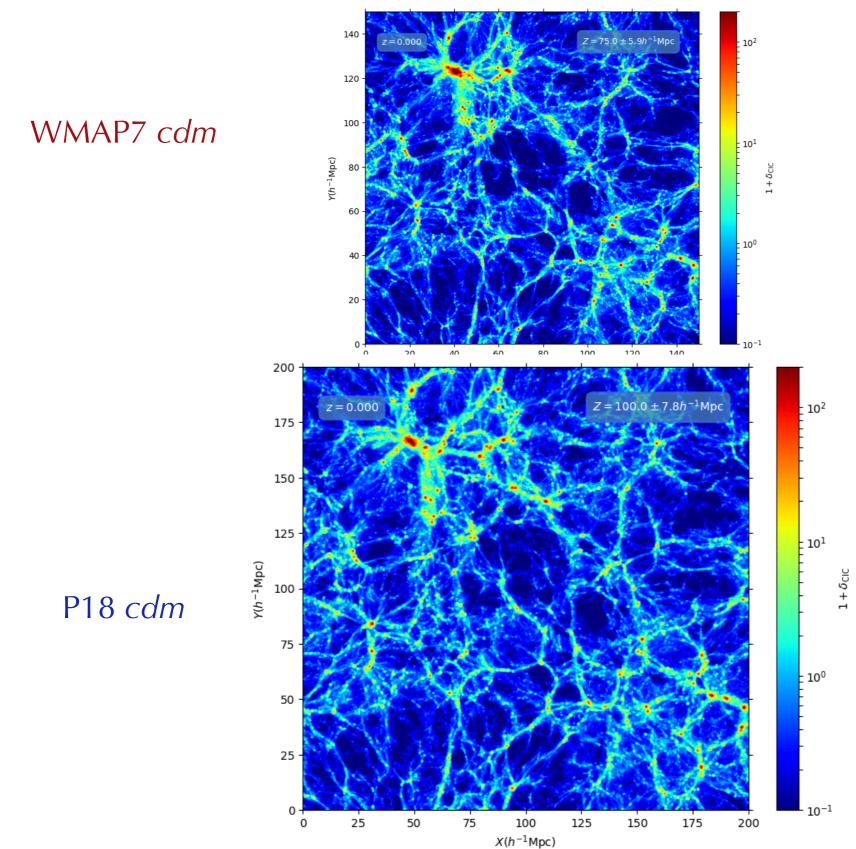
**WMAP7 warm** DM  $(m_{\rm DM} = 0.4 \, \text{keV})$ 

L150\_N1024

**P18 ballistic** DM  $(z_* = 10^5, f_{\text{bdm}} = 0.5)$ 



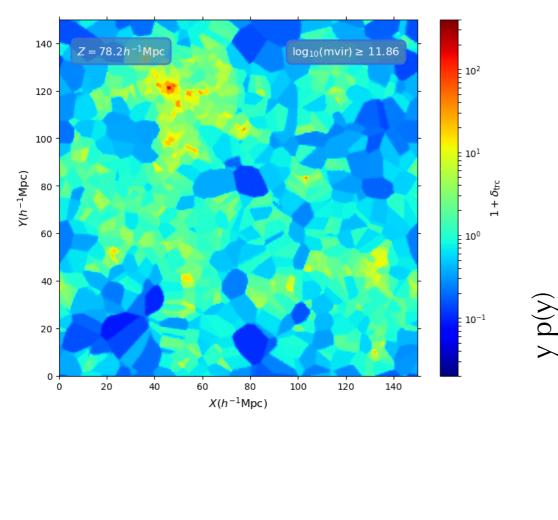
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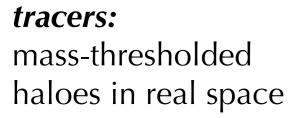


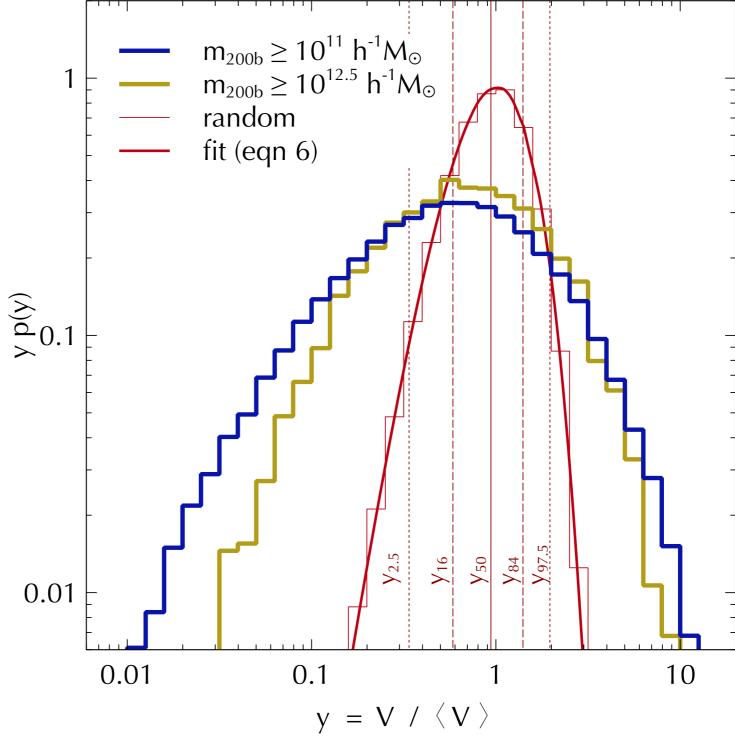


# **VVF** shape

### Dependence on halo mass



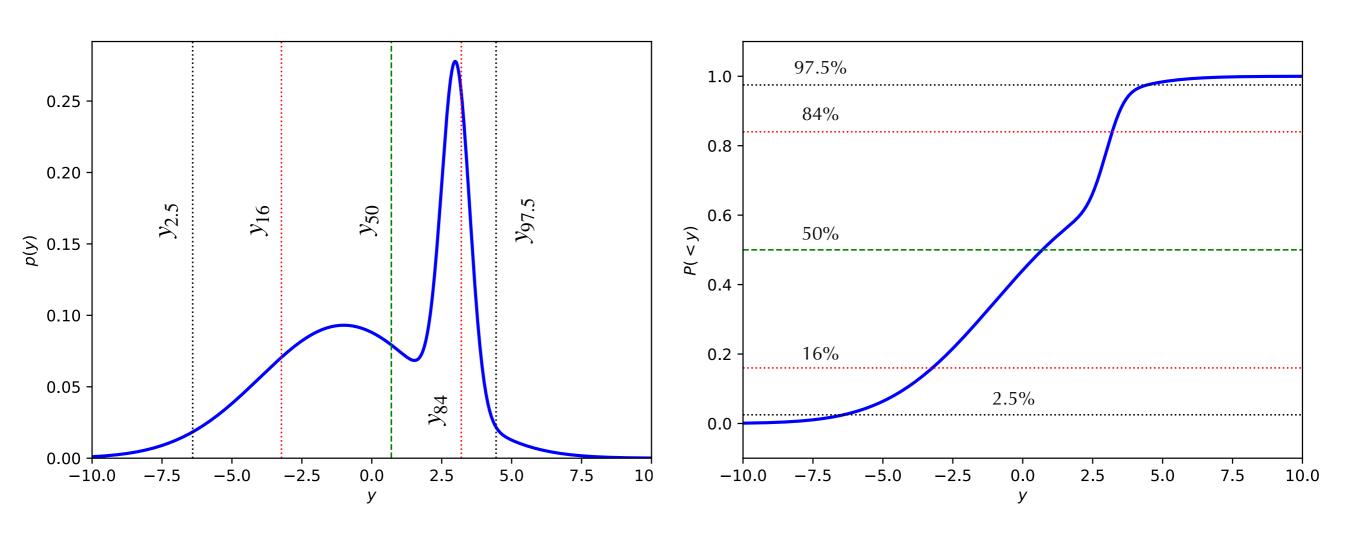






# **Aside**

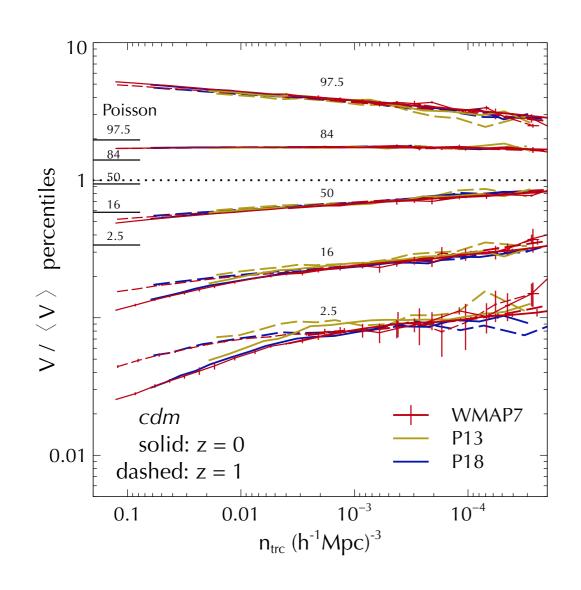
### Percentiles of a distribution

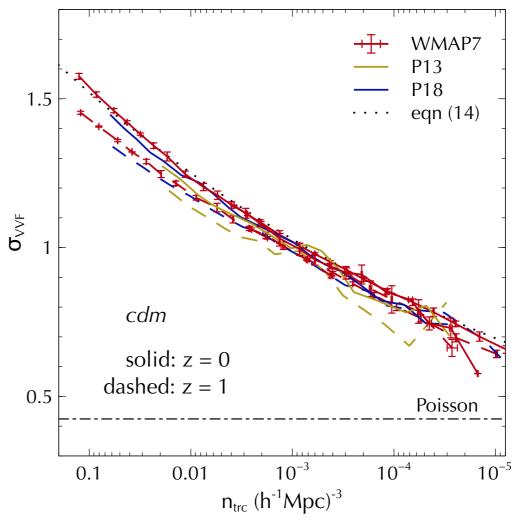




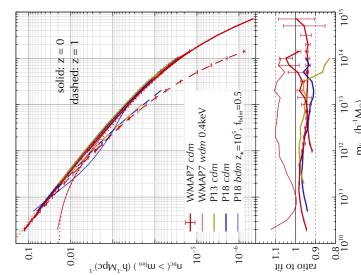
## **VVF** percentiles + std dev

### Dependence on halo mass and redshift





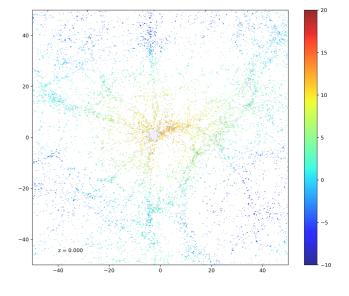
*tracers:* mass-thresholded haloes in real space



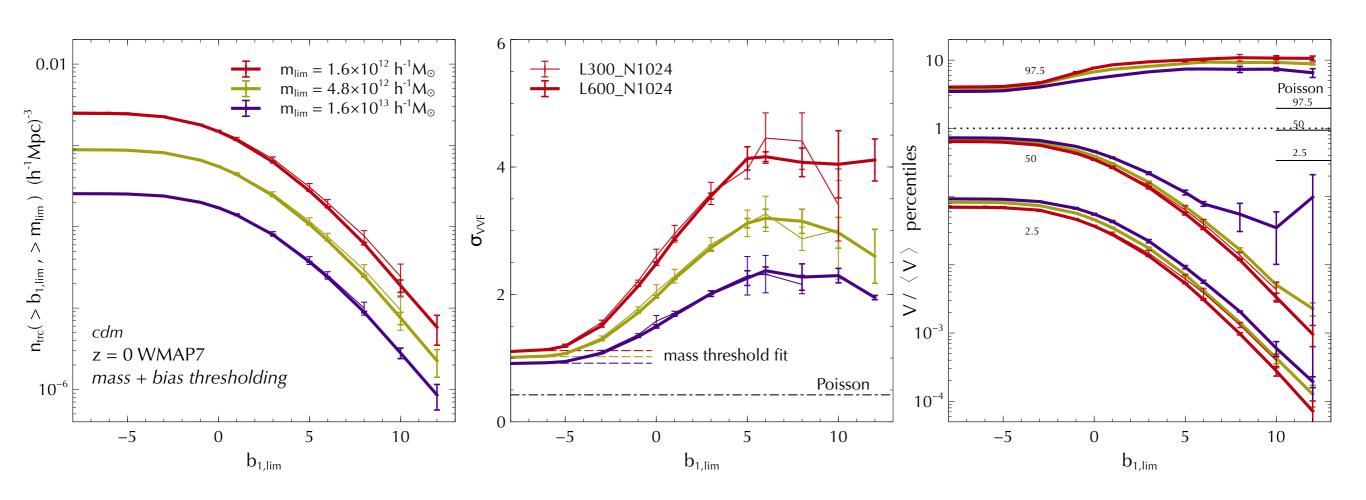


## **VVF** percentiles + std dev

Dependence on halo clustering



[AP, Hahn & Sheth 2018]

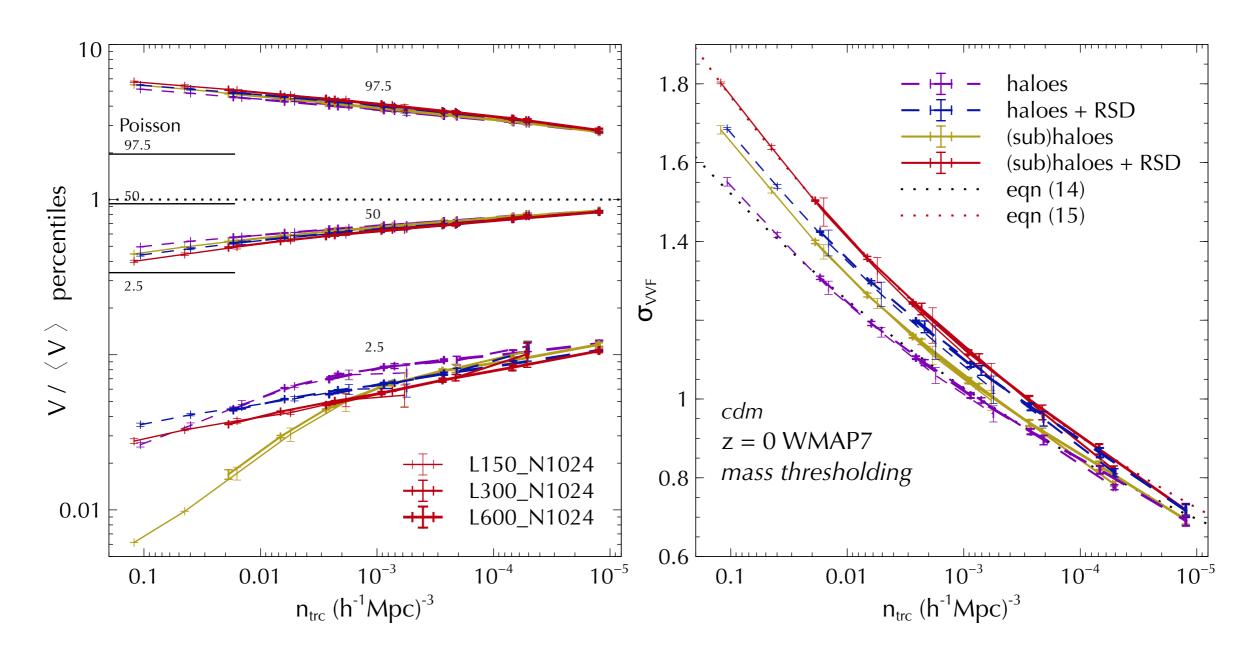


*tracers:* (mass + bias)-thresholded haloes in real space at z=0



## **VVF** percentiles + std dev

### Effects of substructure and redshift-space distortions

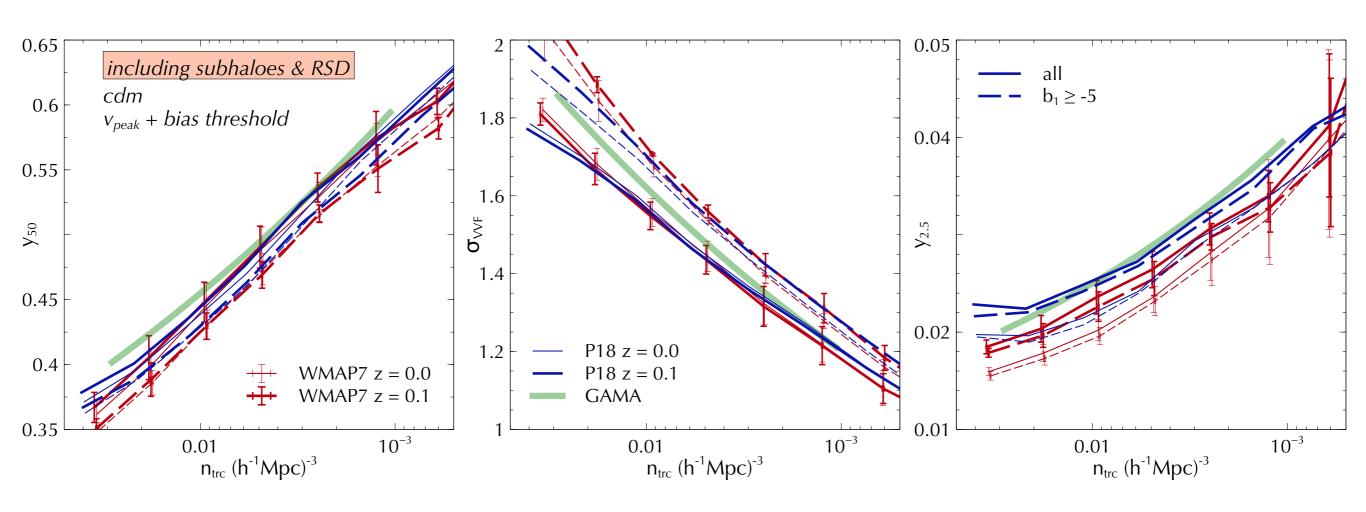


*tracers:* mass-thresholded (sub)haloes in real and redshift space at z=0



# **Comparison with data**

Subhalo abundance matching: CDM simulations

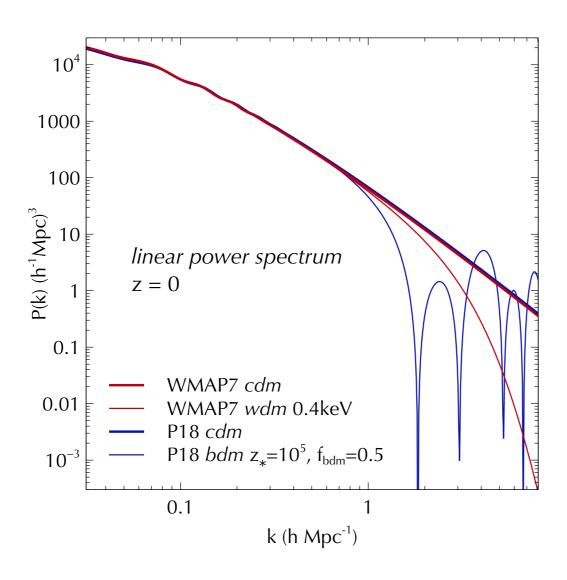


*tracers:* ( $v_{\text{peak}}$ +bias)-thresholded (sub)haloes in redshift space at z=0 and z=0.1



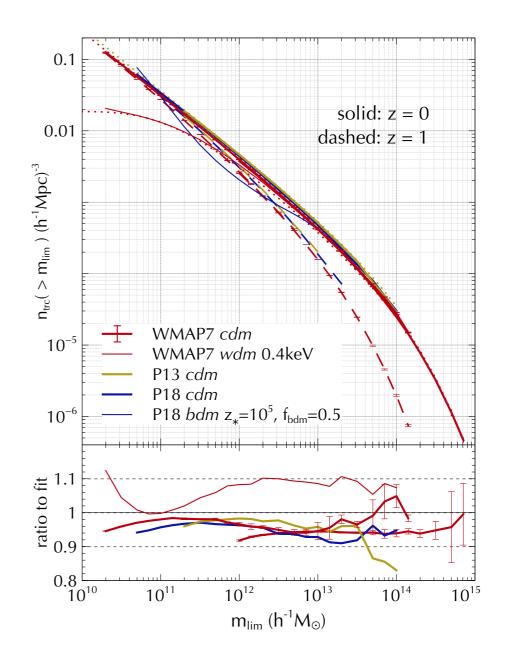
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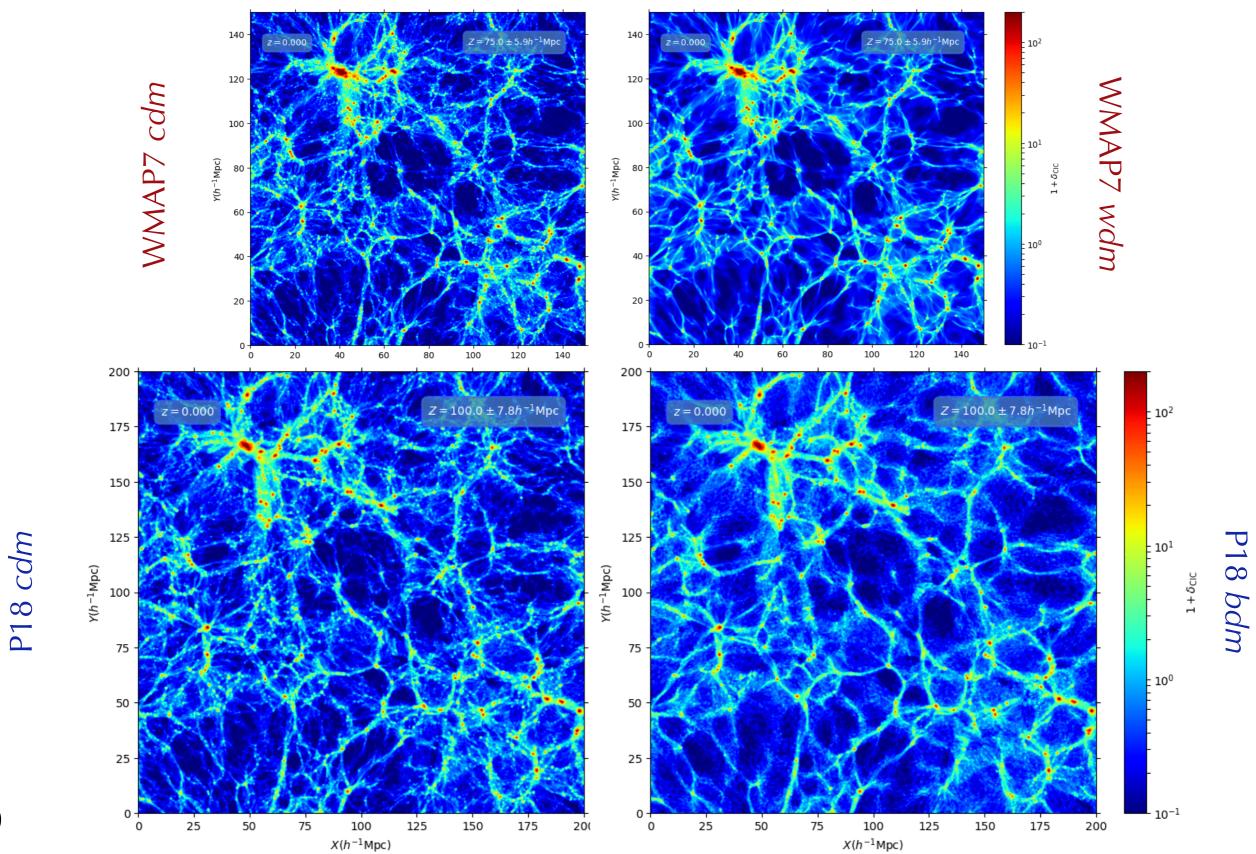
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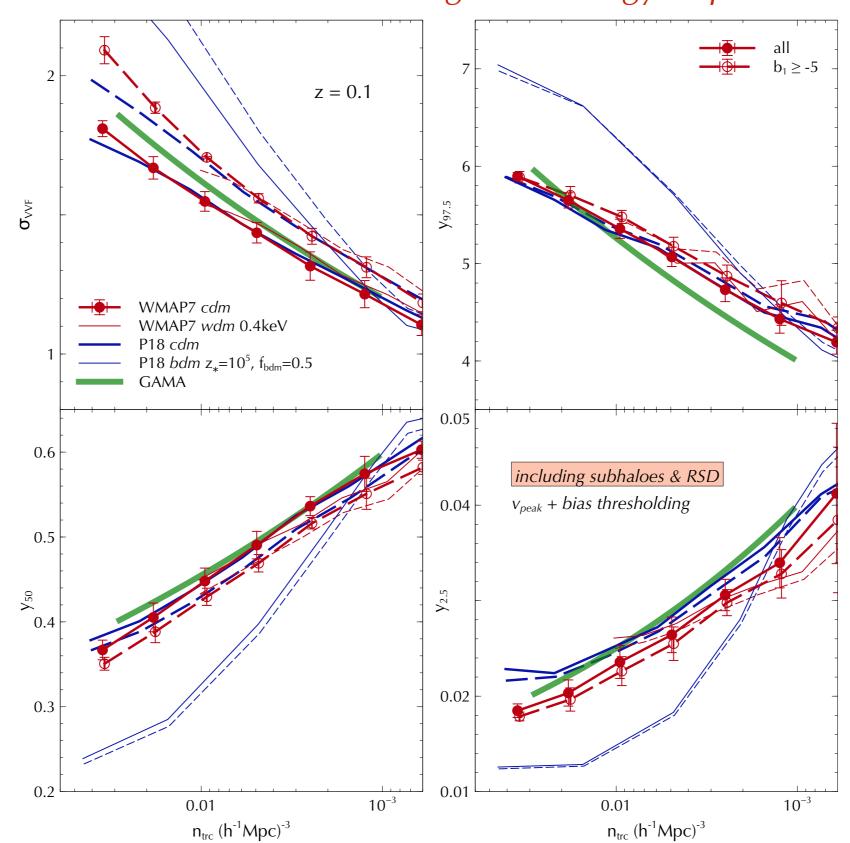
# **Simulation suite**





# **Comparison with data**

Subhalo abundance matching: cosmology dependence





### **Conclusions**

- ★ Voronoi volume function is a novel probe of nonlinear structure, sensitive to variety of tracer properties
  - Halo mass
  - Substructure content
  - Kinematics (RSD)
  - Large-scale clustering
  - Redshift evolution
- ★ Cosmology dependence (*cdm / wdm*):
  - VVF of mass-selected samples essentially universal
  - Weak cosmology dependence in realistic (SHAM) samples
- ★ Cosmology dependence (*bdm*):
  - Strong sensitivity to oscillatory features in P(k)