

$f_{\text{NL}}$  &  $\sum m_\nu$

# FROM THE CLUSTERING OF VOIDS

NICO HAMAUS

in collaboration with

K.C. CHAN, M. BIAGETTI, V. DESJACQUES, Y. LI,  
N. SCHUSTER, G. POLLINA, K. PAECH, J. WELLER,  
A. PISANI, P. SUTTER, G. LAVAUX, B. WANDELT,  
M. BALDI, C. CARBONE, K. DOLAG, C. KREISCH,  
J. FANG, B. JAIN, S. PANDEY, C. SÁNCHEZ, A. KOVÁCS

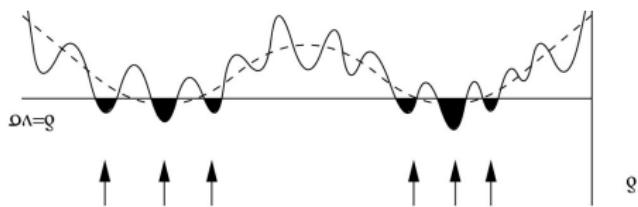


# OUTLINE

- 1 Finding voids
- 2 Void Clustering
- 3 Primordial Non-Gaussianity
- 4 Massive Neutrinos
- 5 Void Lensing
- 6 Conclusions

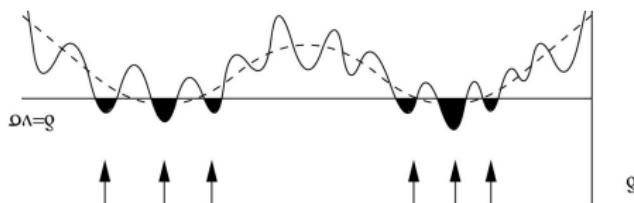
# DEFINITION OF VOIDS

Search for local minima in density field

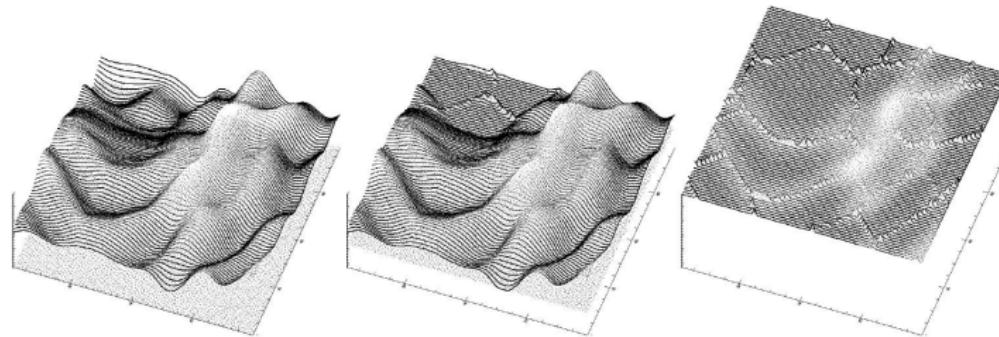


# DEFINITION OF VOIDS

Search for local minima in density field

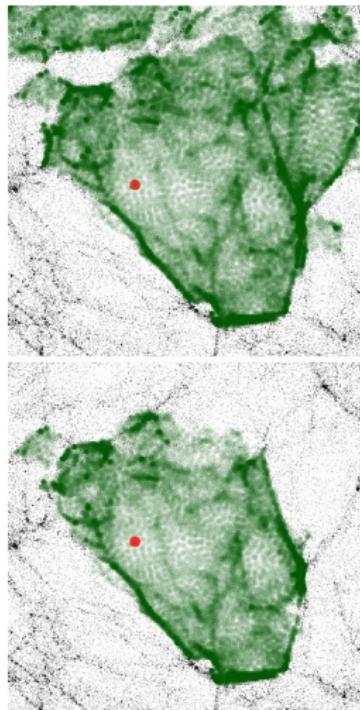


and raise a density threshold until a saddle point is reached

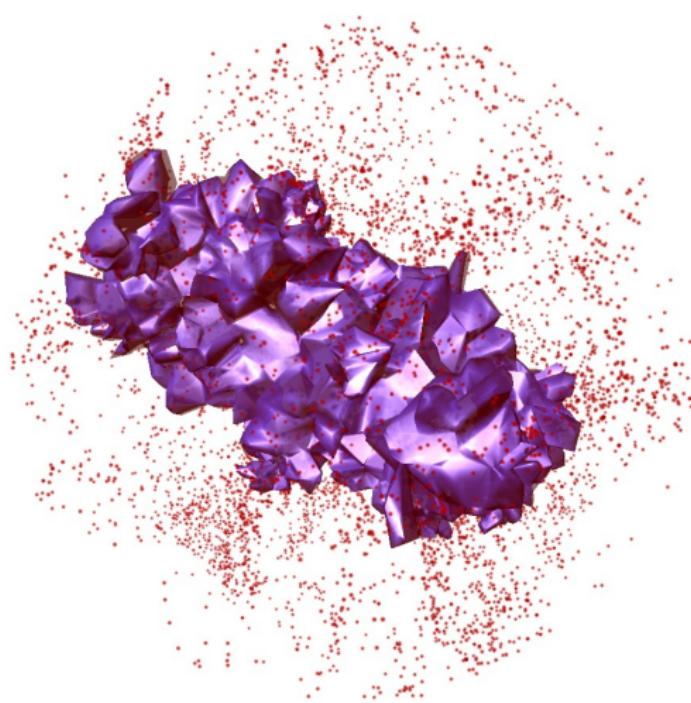


Watershed algorithm: Platen et al. (2007, MNRAS 380, 551)

# DEFINITION OF VOIDS



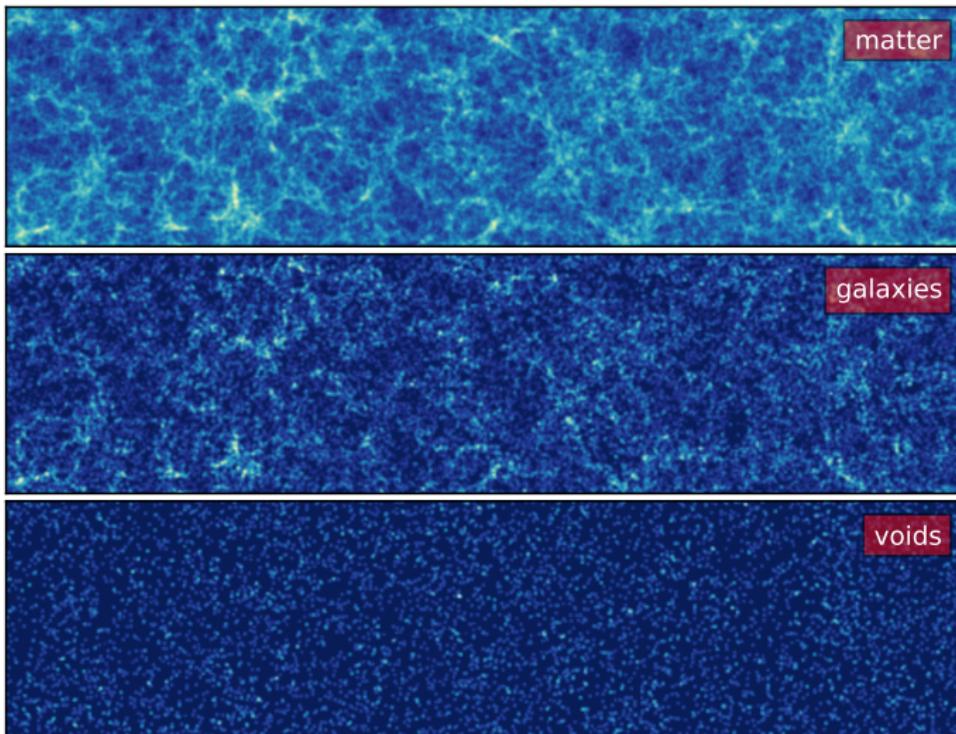
ZOBOV: Neyrinck (2008, MNRAS 386, 4)



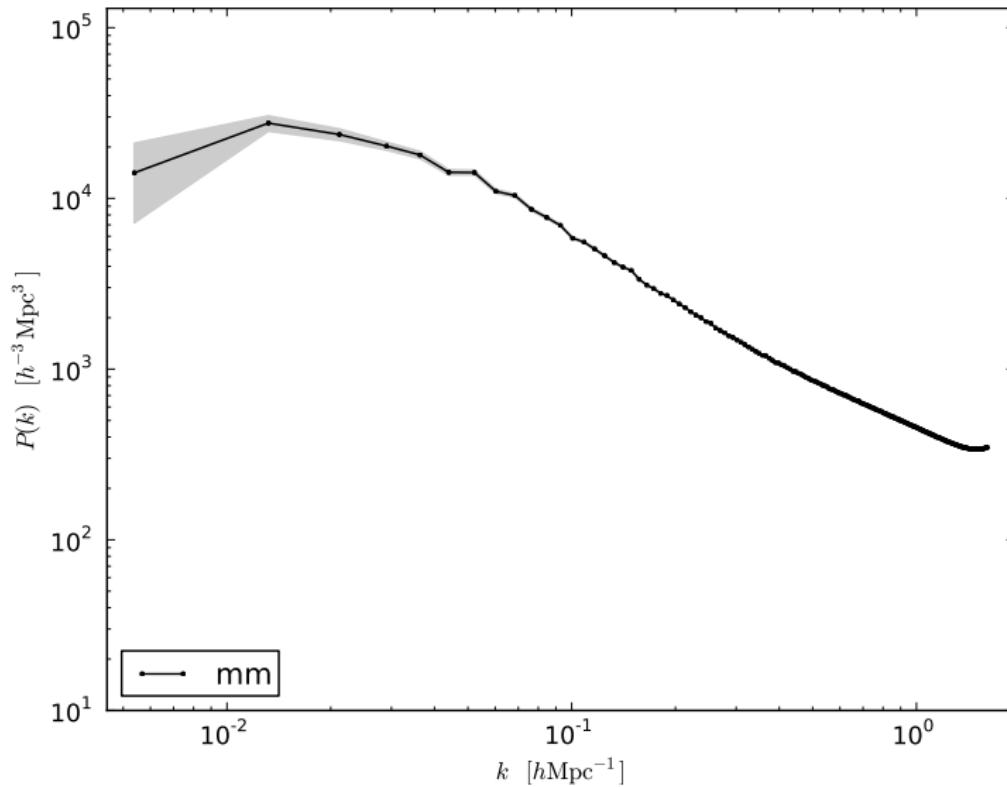
VIDE: Sutter, Lavaux, Hamaus et al. (2015, A&C 9, 1)

# DENSITY FIELDS

Voids are biased tracers of LSS (less clustered and sparser than galaxies)

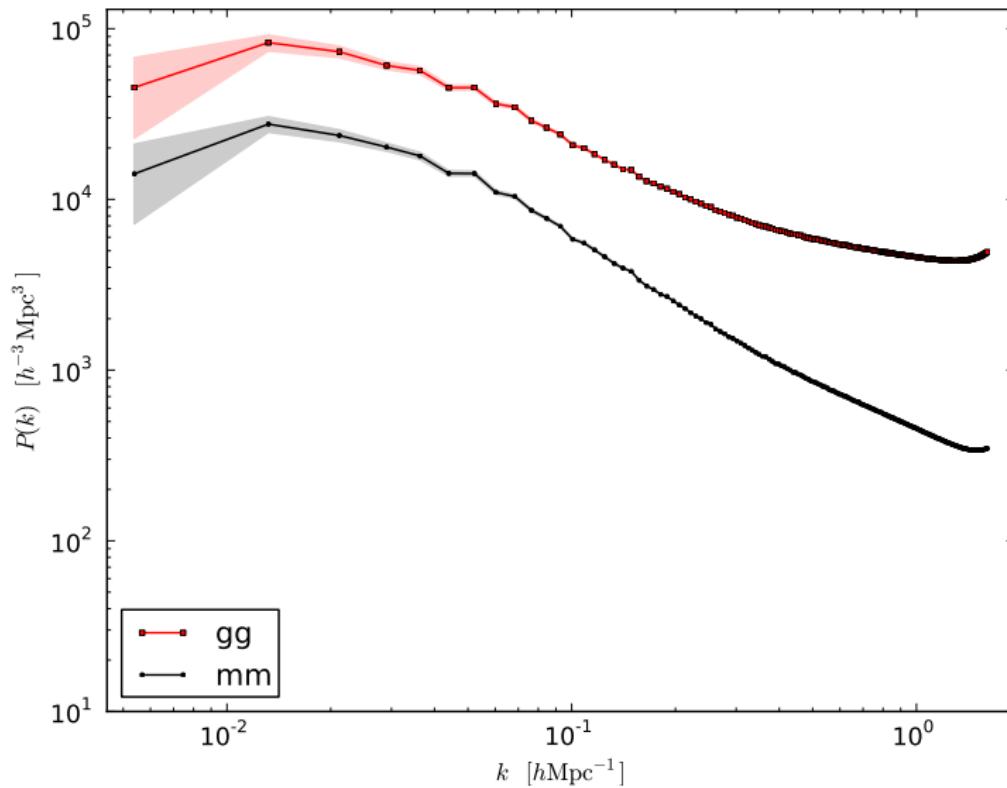


# POWER SPECTRUM



Hamaus et al. (2014, PRL 112, 041304)

# POWER SPECTRUM



Hamaus et al. (2014, PRL 112, 041304)

OUTLINE

FINDING VOIDS

VOID CLUSTERING

NON-GAUSSIANITY

MASSIVE NEUTRINOS

VOID LENSING

CONCLUSIONS

O

OO

●○○○○

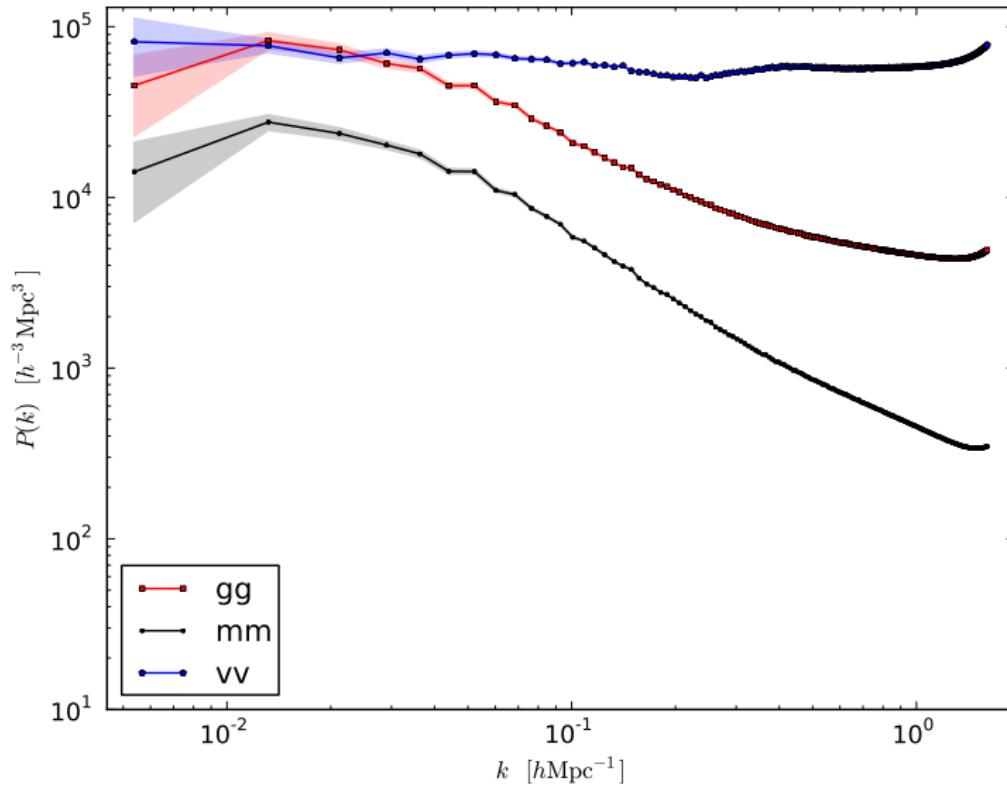
○○○○

○○○○○

○○○○

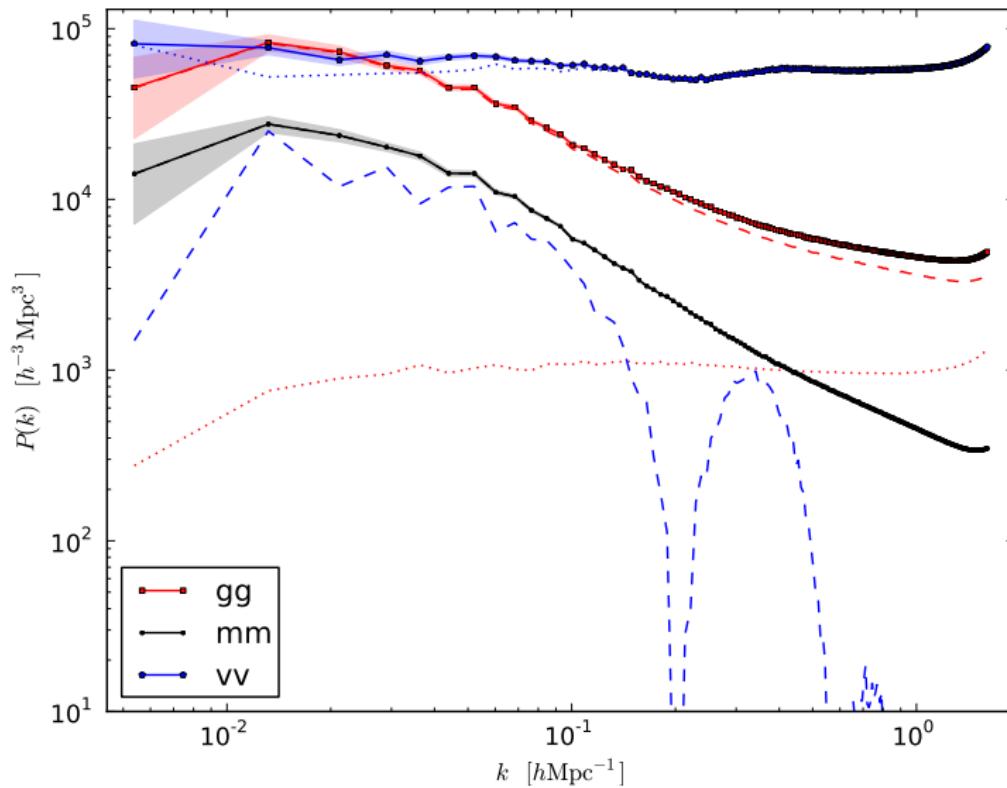
○○○○○○○

# POWER SPECTRUM



Hamaus et al. (2014, PRL 112, 041304)

# POWER SPECTRUM



Hamaus et al. (2014, PRL 112, 041304)

OUTLINE

FINDING VOIDS

VOID CLUSTERING

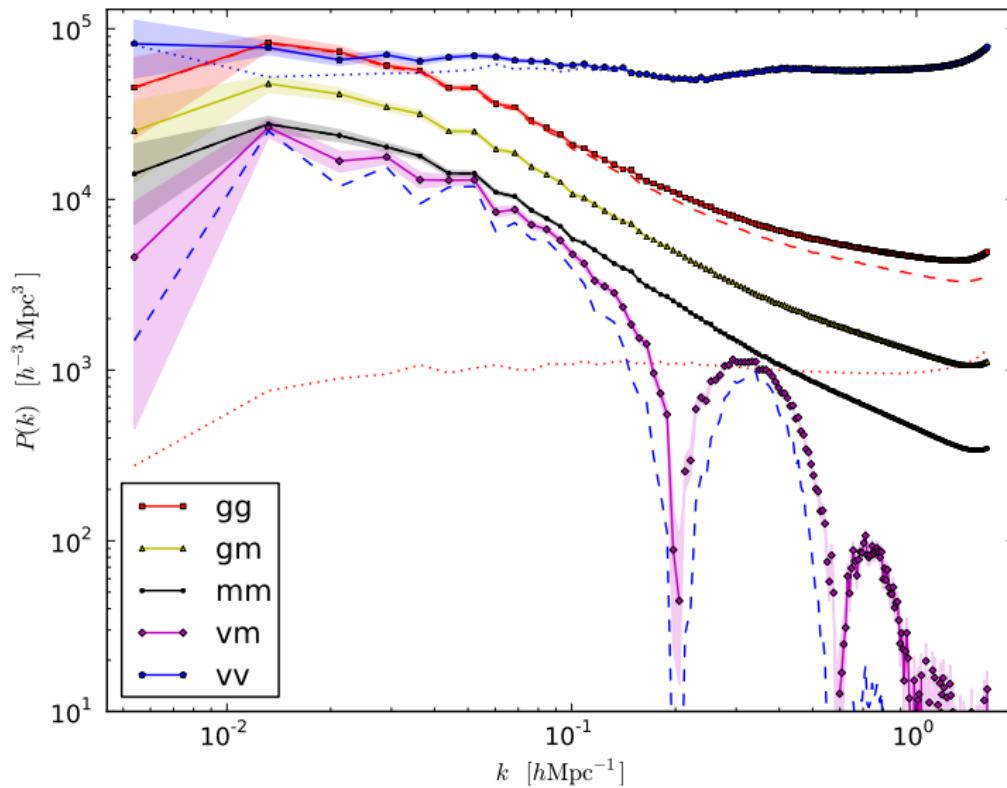
NON-GAUSSIANITY

MASSIVE NEUTRINOS

VOID LENSING

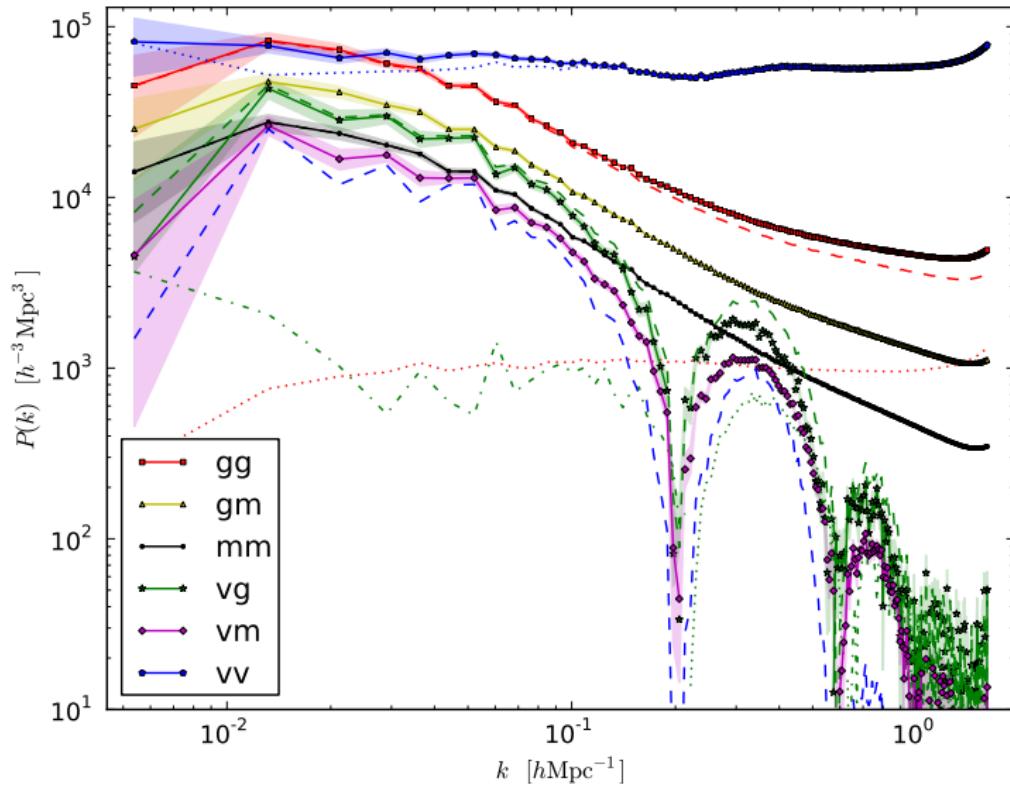
CONCLUSIONS

# POWER SPECTRUM



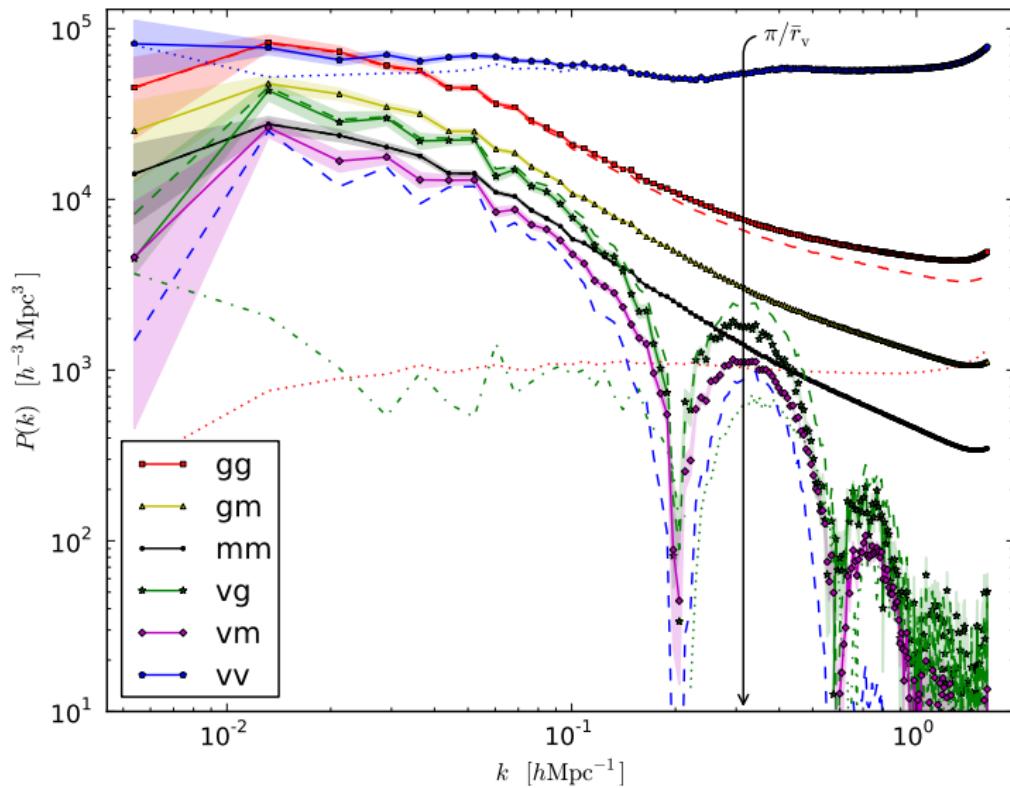
Hamaus et al. (2014, PRL 112, 041304)

# POWER SPECTRUM



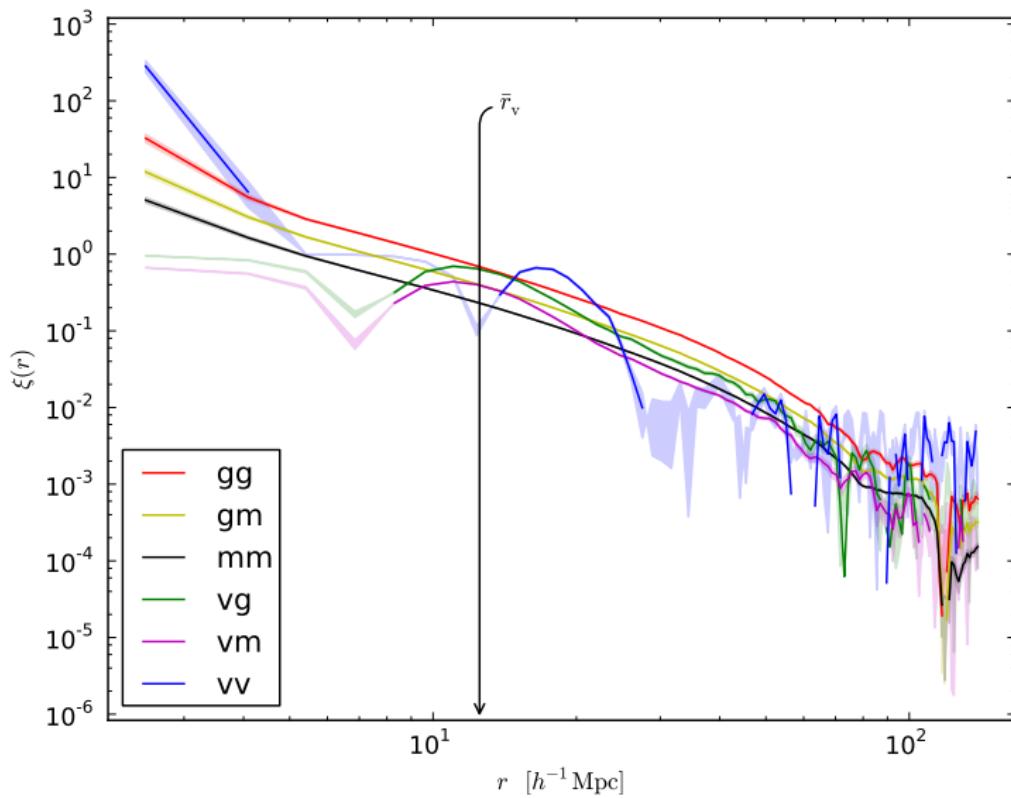
Hamaus et al. (2014, PRL 112, 041304)

# POWER SPECTRUM



Hamaus et al. (2014, PRL 112, 041304)

# CORRELATION FUNCTION



Hamaus et al. (2014, PRL 112, 041304)

# “VOID MODEL”

$$P_{\text{vg}}^{(1\mathcal{V})}(k) = \frac{1}{\bar{n}_{\text{v}}\bar{n}_{\text{g}}} \int \frac{dn_{\text{v}}(r_{\text{v}})}{dr_{\text{v}}} N_{\text{g}}(r_{\text{v}}) u_{\text{v}}(k|r_{\text{v}}) dr_{\text{v}}$$

$$P_{\text{vg}}^{(2\mathcal{V})}(k) = \frac{1}{\bar{n}_{\text{v}}\bar{n}_{\text{g}}} \iint \frac{dn_{\text{v}}(r_{\text{v}})}{dr_{\text{v}}} \frac{dn_{\text{g}}(m_{\text{g}})}{dm_{\text{g}}} b_{\text{v}}(r_{\text{v}}) b_{\text{g}}(m_{\text{g}}) u_{\text{v}}(k|r_{\text{v}}) P(k) dr_{\text{v}} dm_{\text{g}}$$

# “VOID MODEL”

$$P_{\text{vg}}^{(1\mathcal{V})}(k) = \frac{1}{\bar{n}_{\text{v}}\bar{n}_{\text{g}}} \int \frac{\text{d}n_{\text{v}}(r_{\text{v}})}{\text{d}r_{\text{v}}} N_{\text{g}}(r_{\text{v}}) u_{\text{v}}(k|r_{\text{v}}) \text{d}r_{\text{v}}$$

$$P_{\text{vg}}^{(2\mathcal{V})}(k) = \frac{1}{\bar{n}_{\text{v}}\bar{n}_{\text{g}}} \iint \frac{\text{d}n_{\text{v}}(r_{\text{v}})}{\text{d}r_{\text{v}}} \frac{\text{d}n_{\text{g}}(m_{\text{g}})}{\text{d}m_{\text{g}}} b_{\text{v}}(r_{\text{v}}) b_{\text{g}}(m_{\text{g}}) u_{\text{v}}(k|r_{\text{v}}) P(k) \text{d}r_{\text{v}} \text{d}m_{\text{g}}$$

Model for voids of given radius

$$\begin{aligned} P_{\text{vg}}(k) &\simeq b_{\text{v}} b_{\text{g}} u_{\text{v}}(k) P(k) + \bar{n}_{\text{v}}^{-1} u_{\text{v}}(k) \\ P_{\text{vv}}(k) &\simeq b_{\text{v}}^2 P(k) + \bar{n}_{\text{v}}^{-1} \\ P_{\text{vm}}(k) &\simeq b_{\text{v}} u_{\text{v}}(k) P(k) \end{aligned}$$

# “VOID MODEL”

$$P_{\text{vg}}^{(1\mathcal{V})}(k) = \frac{1}{\bar{n}_{\text{v}}\bar{n}_{\text{g}}} \int \frac{\text{d}n_{\text{v}}(r_{\text{v}})}{\text{d}r_{\text{v}}} N_{\text{g}}(r_{\text{v}}) u_{\text{v}}(k|r_{\text{v}}) \text{d}r_{\text{v}}$$

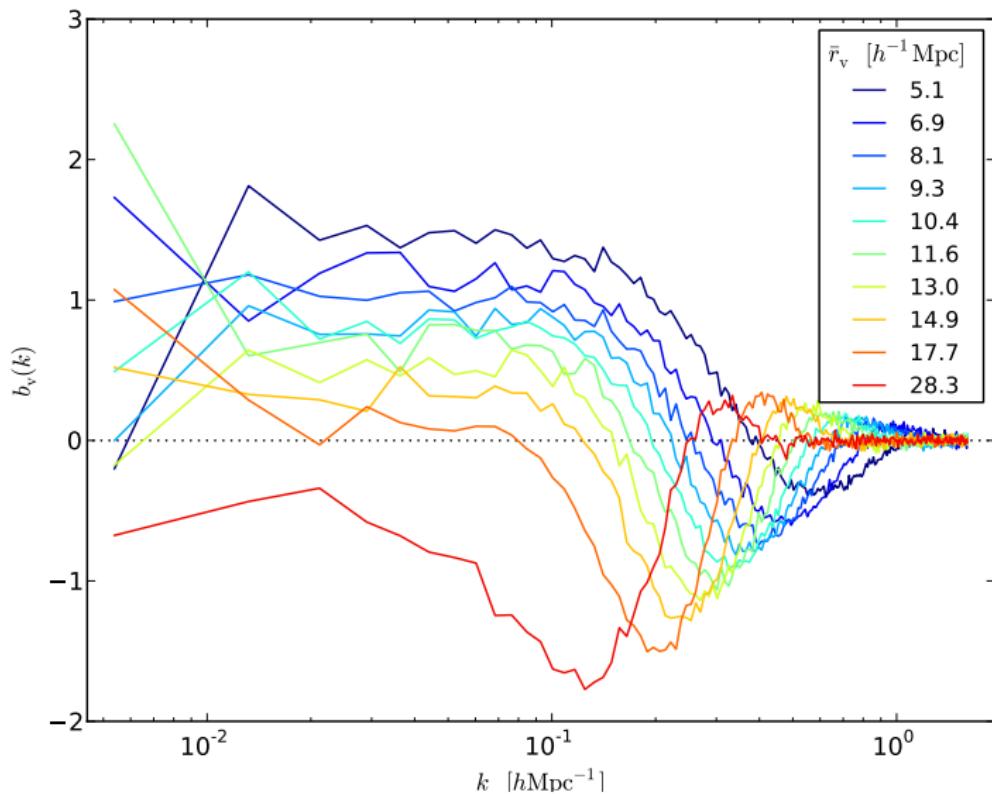
$$P_{\text{vg}}^{(2\mathcal{V})}(k) = \frac{1}{\bar{n}_{\text{v}}\bar{n}_{\text{g}}} \iint \frac{\text{d}n_{\text{v}}(r_{\text{v}})}{\text{d}r_{\text{v}}} \frac{\text{d}n_{\text{g}}(m_{\text{g}})}{\text{d}m_{\text{g}}} b_{\text{v}}(r_{\text{v}}) b_{\text{g}}(m_{\text{g}}) u_{\text{v}}(k|r_{\text{v}}) P(k) \text{d}r_{\text{v}} \text{d}m_{\text{g}}$$

Model for voids of given radius

$$\begin{aligned} P_{\text{vg}}(k) &\simeq b_{\text{v}} b_{\text{g}} u_{\text{v}}(k) P(k) + \bar{n}_{\text{v}}^{-1} u_{\text{v}}(k) \\ P_{\text{vv}}(k) &\simeq b_{\text{v}}^2 P(k) + \bar{n}_{\text{v}}^{-1} \\ P_{\text{vm}}(k) &\simeq b_{\text{v}} u_{\text{v}}(k) P(k) \end{aligned}$$

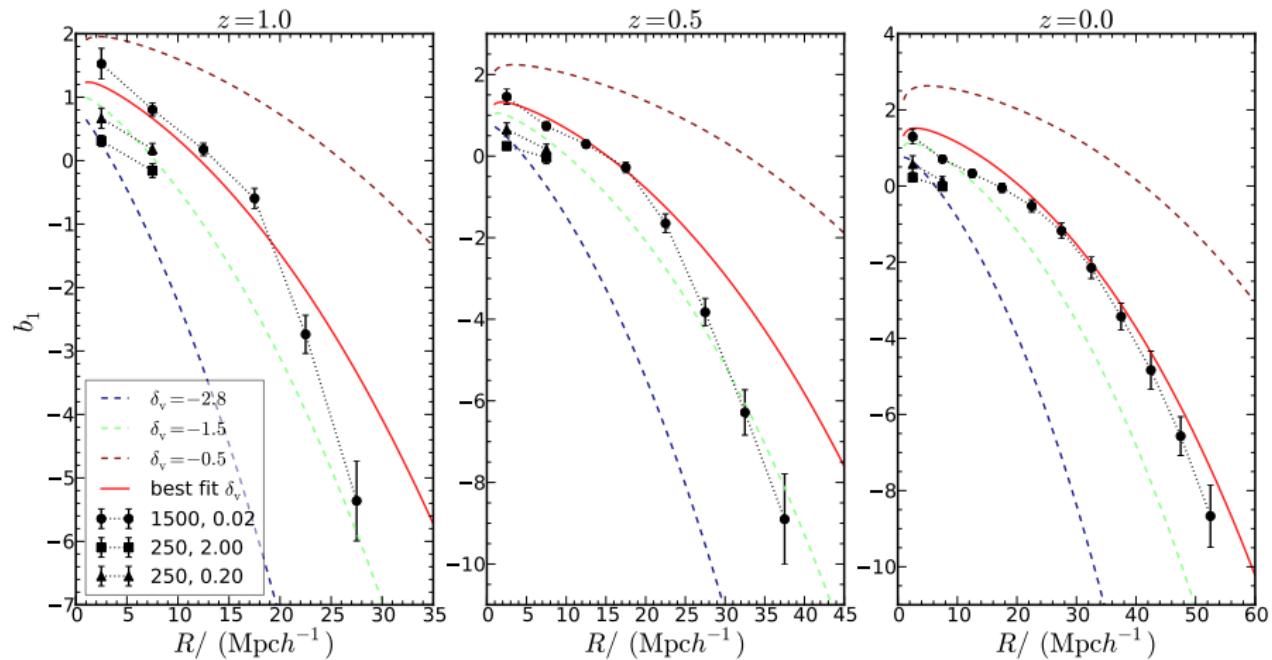
- ➡ Void density profile is normalized to  $u_{\text{v}}(k \rightarrow 0) = 1$
- ➡ Large scales: linearly biased tracers of dark matter
- ➡ Small scales: dominated by internal void structure

# VOID BIAS



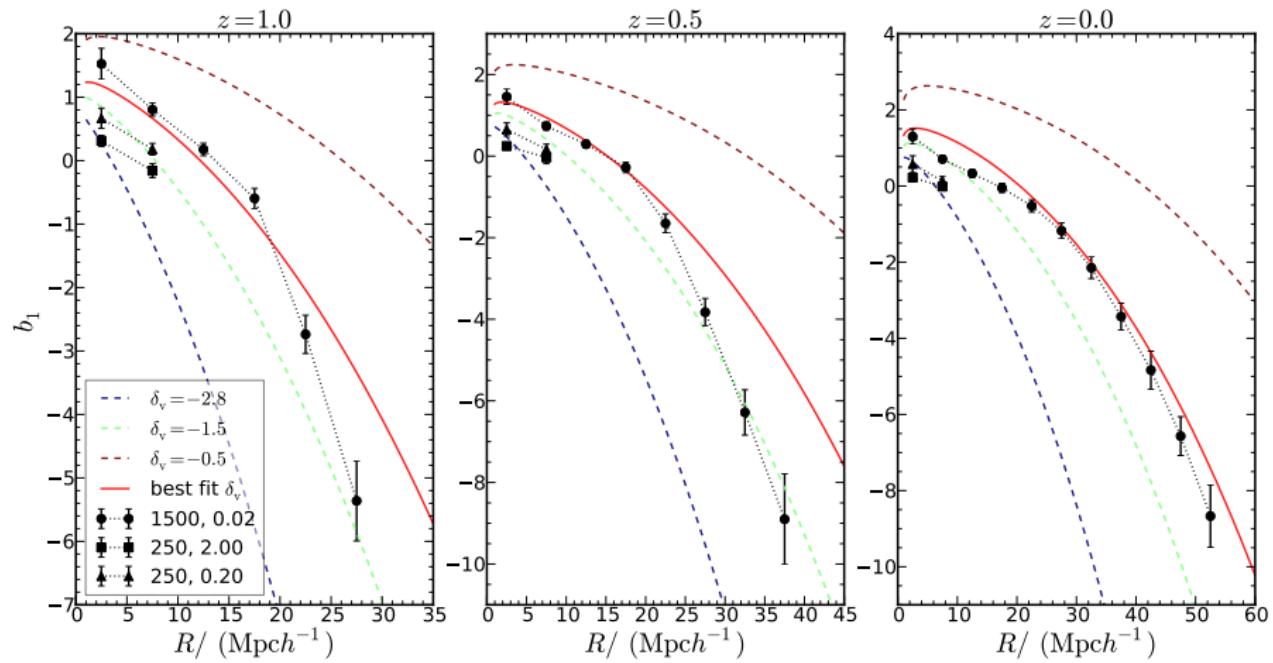
Hamaus et al. (2014, PRL 112, 041304)

# LINEAR VOID BIAS



Chan, Hamaus, Desjacques (2014, PRD 90, 103521)

# LINEAR VOID BIAS



Chan, Hamaus, Desjacques (2014, PRD 90, 103521)

Separate Universe Simulations: Chan, Li, Biagetti, Hamaus (2019, arXiv:1909.03736)

# PRIMORDIAL NON-GAUSSIANITY

Expand primordial potential locally around Gaussian to first order

$$\Phi(\mathbf{x}) = \Phi_G(\mathbf{x}) + f_{NL} [\Phi_G^2(\mathbf{x}) - \langle \Phi_G^2 \rangle]$$

Coupling between short- and long-wavelength modes effectively causes local rescaling in the amplitude of matter fluctuations  $\sigma_8$ .

# PRIMORDIAL NON-GAUSSIANITY

Expand primordial potential locally around Gaussian to first order

$$\Phi(\mathbf{x}) = \Phi_G(\mathbf{x}) + f_{NL} [\Phi_G^2(\mathbf{x}) - \langle \Phi_G^2 \rangle]$$

Coupling between short- and long-wavelength modes effectively causes local rescaling in the amplitude of matter fluctuations  $\sigma_8$ .

Additional *scale-dependent* contribution to linear bias

$$b_{NG} = 2f_{NL}\mathcal{M}^{-1} \frac{\partial \ln n}{\partial \ln \sigma_8}, \quad \mathcal{M}(k) = \frac{2}{3} \frac{k^2 T(k) D(z)}{\Omega_m H_0^2}$$

# PRIMORDIAL NON-GAUSSIANITY

Expand primordial potential locally around Gaussian to first order

$$\Phi(\mathbf{x}) = \Phi_G(\mathbf{x}) + f_{NL} [\Phi_G^2(\mathbf{x}) - \langle \Phi_G^2 \rangle]$$

Coupling between short- and long-wavelength modes effectively causes local rescaling in the amplitude of matter fluctuations  $\sigma_8$ .

Additional *scale-dependent* contribution to linear bias

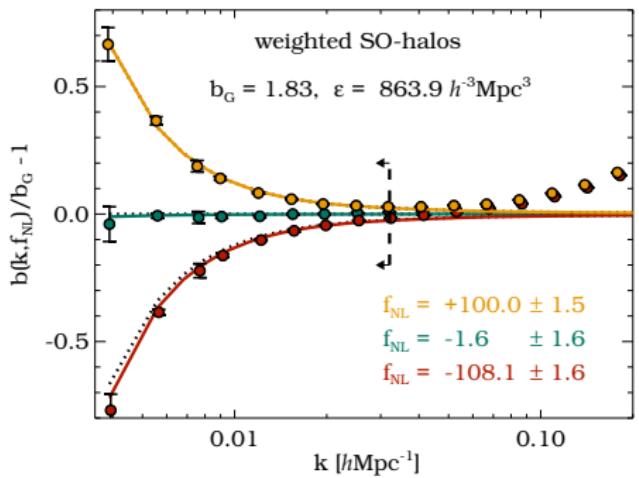
$$b_{NG} = 2f_{NL}\mathcal{M}^{-1} \frac{\partial \ln n}{\partial \ln \sigma_8}, \quad \mathcal{M}(k) = \frac{2}{3} \frac{k^2 T(k) D(z)}{\Omega_m H_0^2}$$

With a *universal* halo-mass function / void-size function:

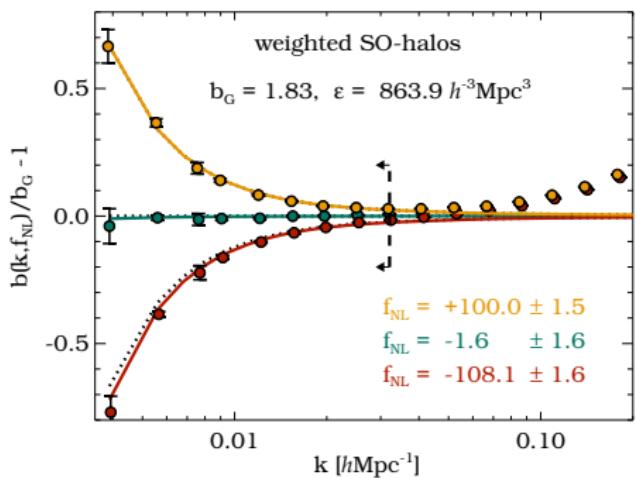
$$b(k, f_{NL}) = b_G + f_{NL}(b_G - 1)\delta_c \frac{3\Omega_m H_0^2}{k^2 T(k) D(z)}$$

Dalal et al. (2008, PRD 77, 123514); Slosar et al. (2008, JCAP 08, 031)

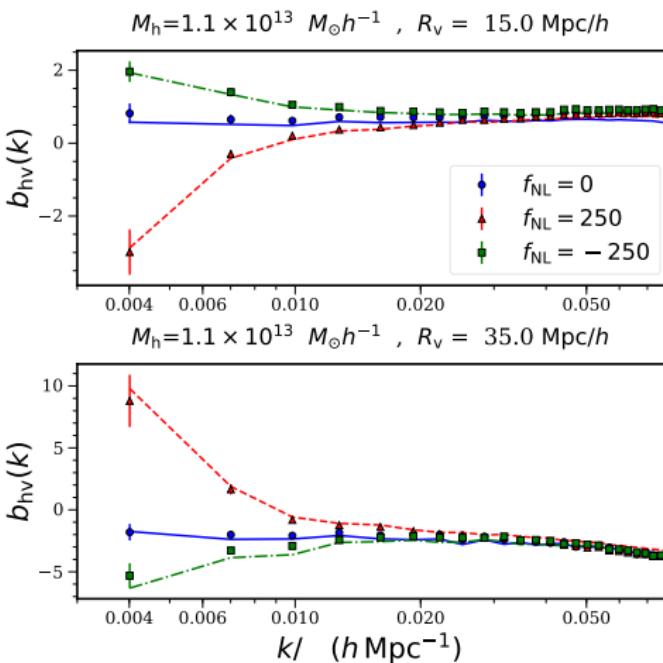
# SCALE-DEPENDENT BIAS



# SCALE-DEPENDENT BIAS

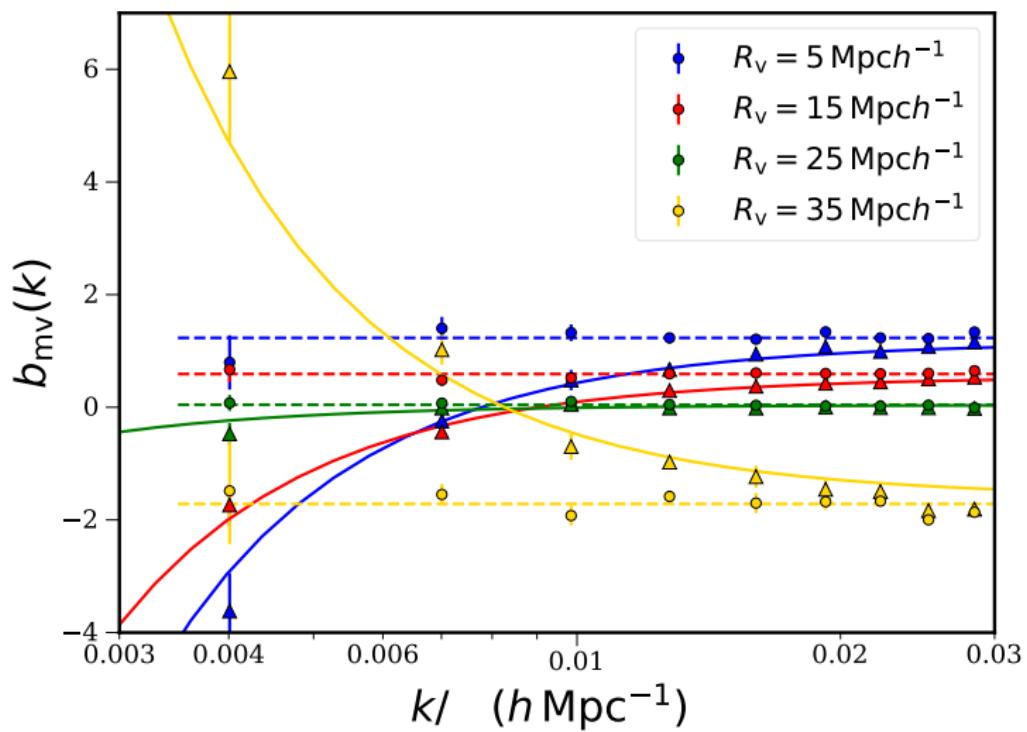


Hamaus, Seljak, Desjacques (2011, PRD 84, 083509)



Chan, Hamaus, Biagetti (2019, PRD 99, 121304)

# SCALE-DEPENDENT VOID BIAS



Chan, Hamaus, Biagetti (2019, PRD 99, 121304)

# FISHER FORECAST

Perform *multi-tracer* analysis (5 halo-mass bins, 3 void-size bins):

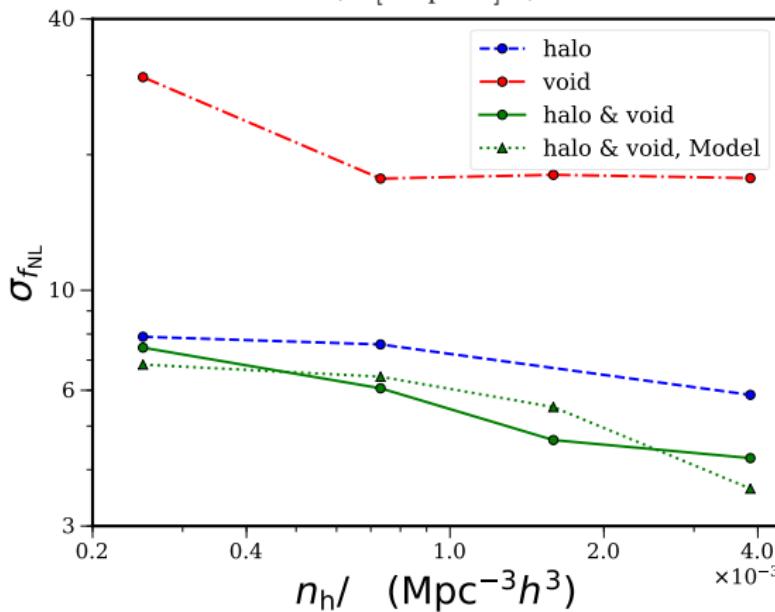
$$F_{f_{\text{NL}} f_{\text{NL}}} = V \int \frac{d^3 k}{(2\pi)^3} \frac{1}{2} \text{Tr} \left( \Sigma^{-1} \frac{\partial \Sigma}{\partial f_{\text{NL}}} \Sigma^{-1} \frac{\partial \Sigma}{\partial f_{\text{NL}}} \right), \quad \Sigma_{ij} \equiv \langle \delta_i(\mathbf{k}) \delta_j(\mathbf{k}) \rangle$$

# FISHER FORECAST

Perform *multi-tracer* analysis (5 halo-mass bins, 3 void-size bins):

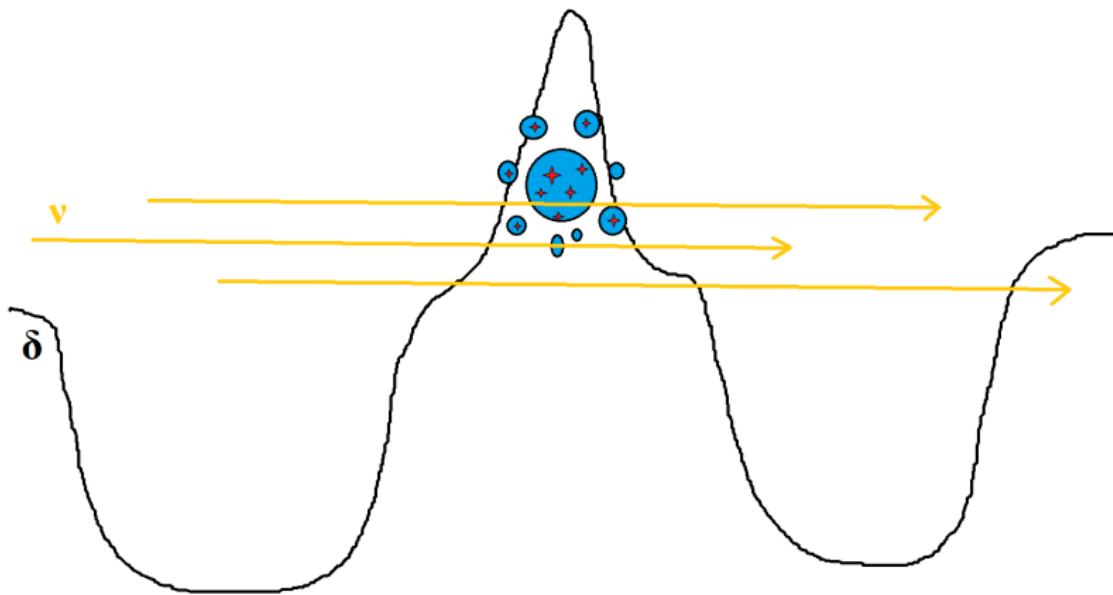
$$F_{f_{\text{NL}} f_{\text{NL}}} = V \int \frac{d^3k}{(2\pi)^3} \frac{1}{2} \text{Tr} \left( \Sigma^{-1} \frac{\partial \Sigma}{\partial f_{\text{NL}}} \Sigma^{-1} \frac{\partial \Sigma}{\partial f_{\text{NL}}} \right), \quad \Sigma_{ij} \equiv \langle \delta_i(\mathbf{k}) \delta_j(\mathbf{k}) \rangle$$

$$0.004 < k [h\text{Mpc}^{-1}] < 0.08$$



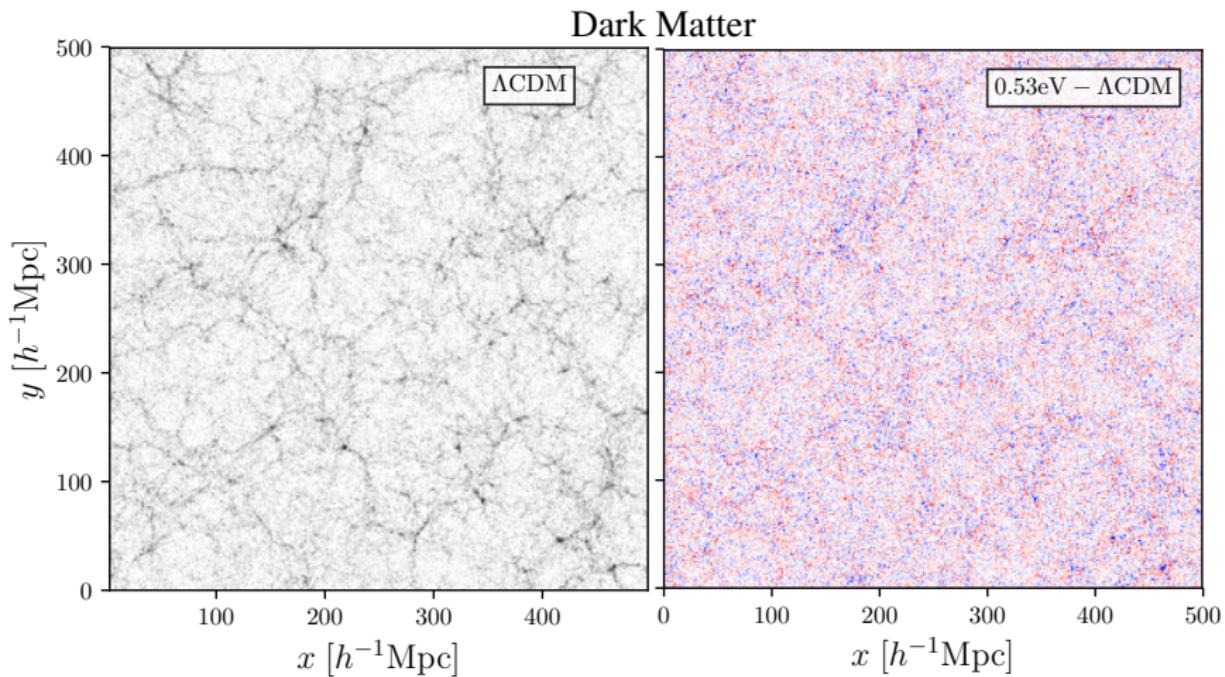
# NEUTRINOS AND VOIDS

Neutrinos freely stream into voids



- Massara et al. (2015, JCAP 11, 018)  
Banerjee & Dalal (2016, JCAP 11, 015)  
Kreisch et al. (2019, MNRAS 488, 4413)

# DEMUNI SIMULATIONS

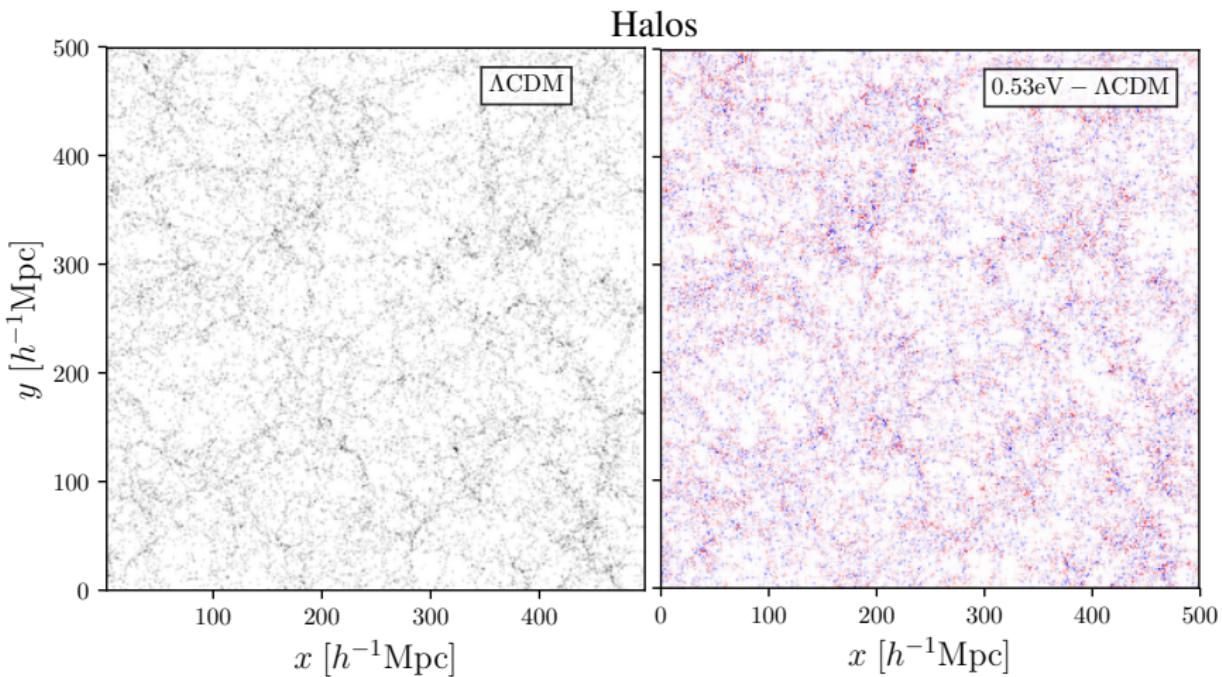


$$L_{\text{box}} = 2h^{-1}\text{Gpc}, N_{\text{cdm}} = 2048^3, N_{\nu} = 2048^3, \sum m_{\nu} [\text{eV}] = 0.17, 0.30, 0.53$$

Castorina et al. (2015, JCAP 7, 043); Carbone, Petkova, Dolag (2016, JCAP 7, 034)

OUTLINE  
OFINDING VOIDS  
OOVOID CLUSTERING  
OOOOOONON-GAUSSIANITY  
OOOOMASSIVE NEUTRINOS  
OO●OOOVOID LENSING  
OOOOCONCLUSIONS  
OOOOOOO

# DEMUNI SIMULATIONS

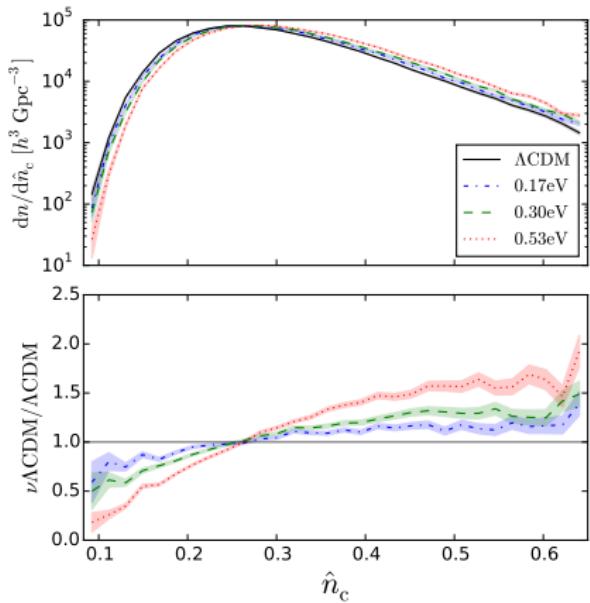


$$L_{\text{box}} = 2h^{-1}\text{Gpc}, N_{\text{cdm}} = 2048^3, N_{\nu} = 2048^3, \sum m_{\nu} [\text{eV}] = 0.17, 0.30, 0.53$$

Castorina et al. (2015, JCAP 7, 043); Carbone, Petkova, Dolag (2016, JCAP 7, 034)

# VOID ABUNDANCE

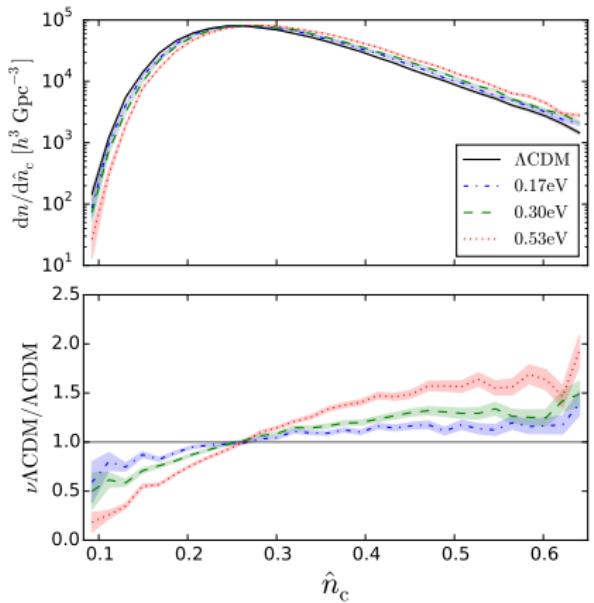
Dark matter voids



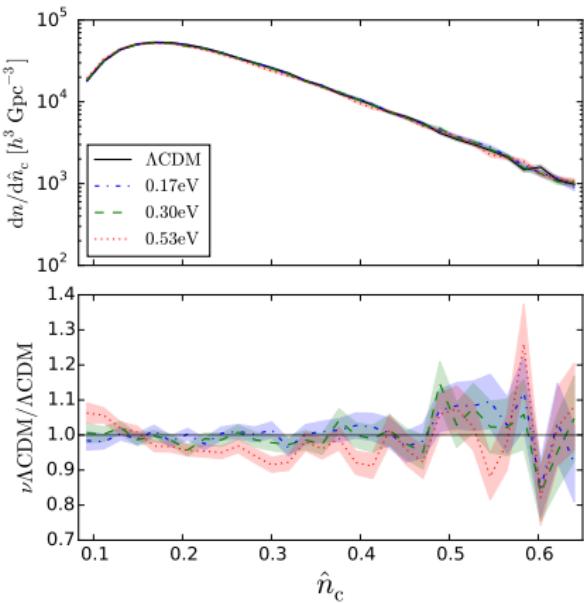
Halo voids

# VOID ABUNDANCE

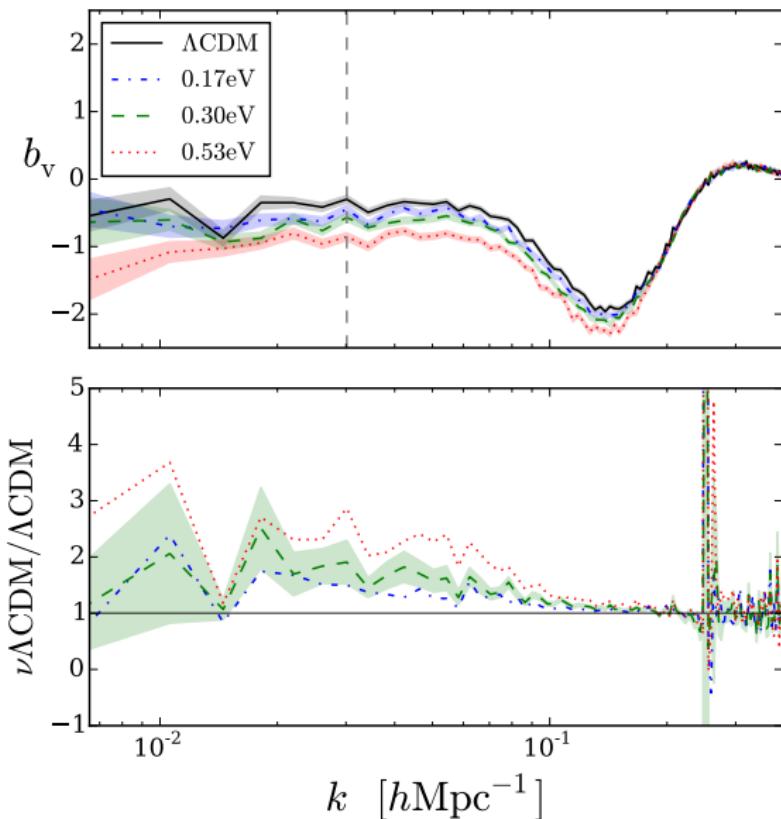
## Dark matter voids



## Halo voids

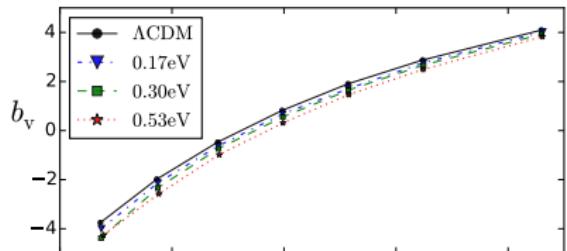


# VOID BIAS

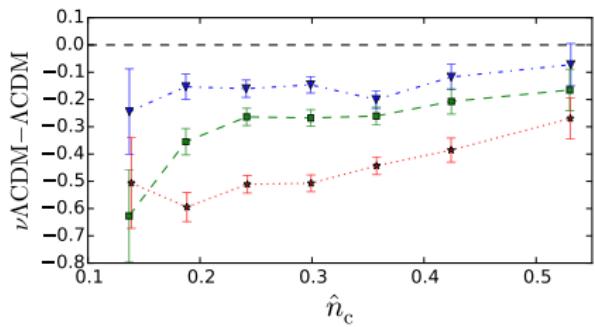


# LINEAR VOID BIAS

Dark matter voids

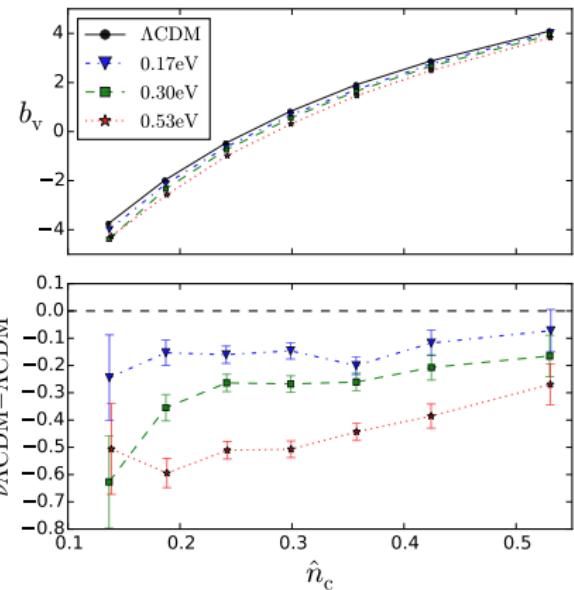


Halo voids

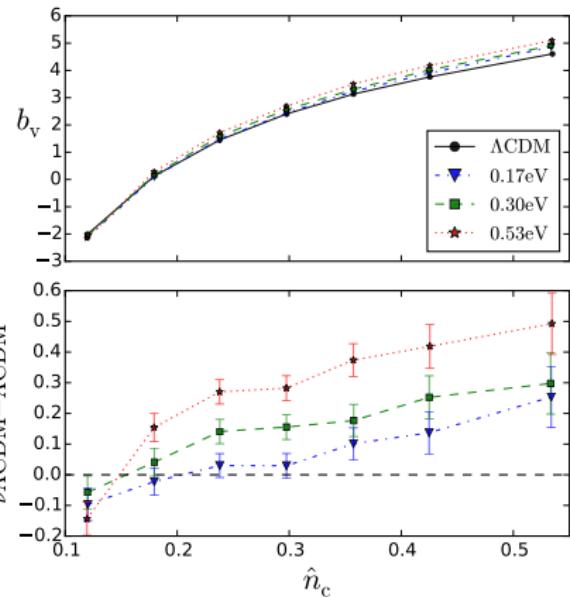


# LINEAR VOID BIAS

Dark matter voids



Halo voids



Schuster, Hamaus et al. (2019, JCAP in press, arXiv:1905.00436)

OUTLINE  
O

FINDING VOIDS  
OO

VOID CLUSTERING  
OOOOOO

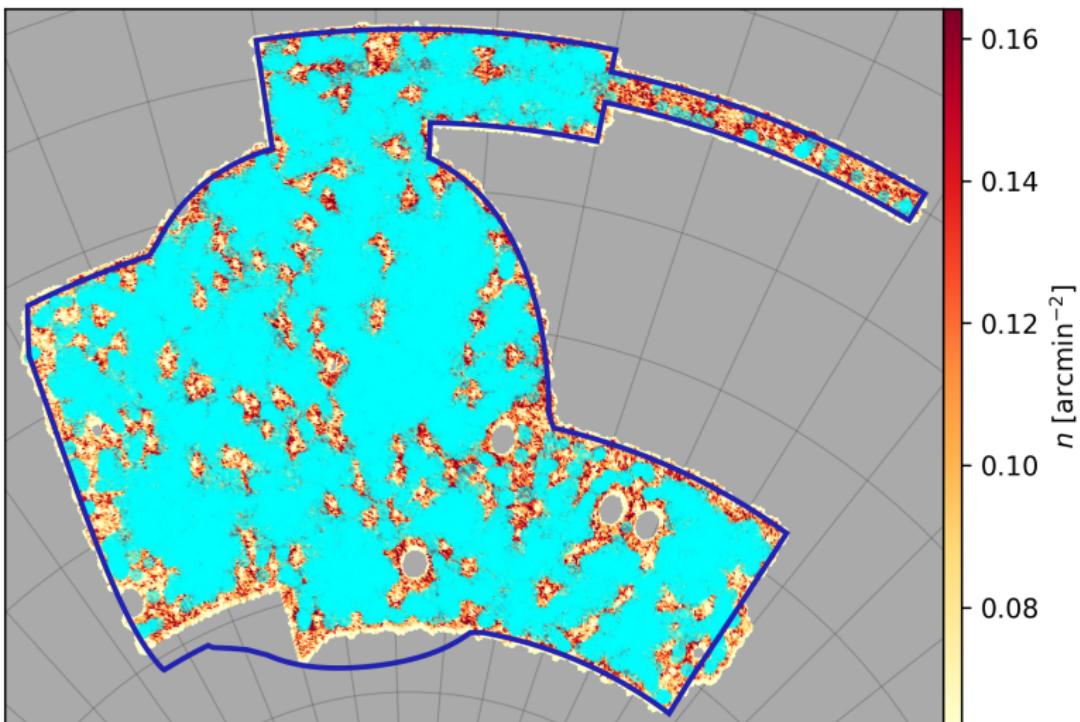
NON-GAUSSIANITY  
OOOO

MASSIVE NEUTRINOS  
OOOOOO

VOID LENSING  
●OOO

CONCLUSIONS  
OOOOOOO

# VOIDS IN THE DARK ENERGY SURVEY



Pollina, Hamaus et al. (in prep. for DES)

OUTLINE  
O

FINDING VOIDS  
OO

VOID CLUSTERING  
OOOOOO

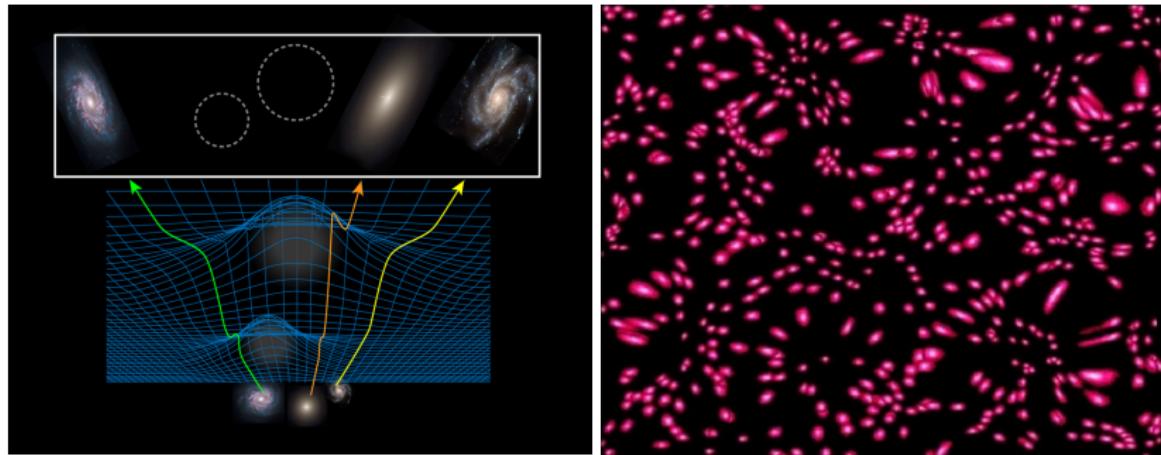
NON-GAUSSIANITY  
OOOO

MASSIVE NEUTRINOS  
OOOOOO

VOID LENSING  
OO●OO

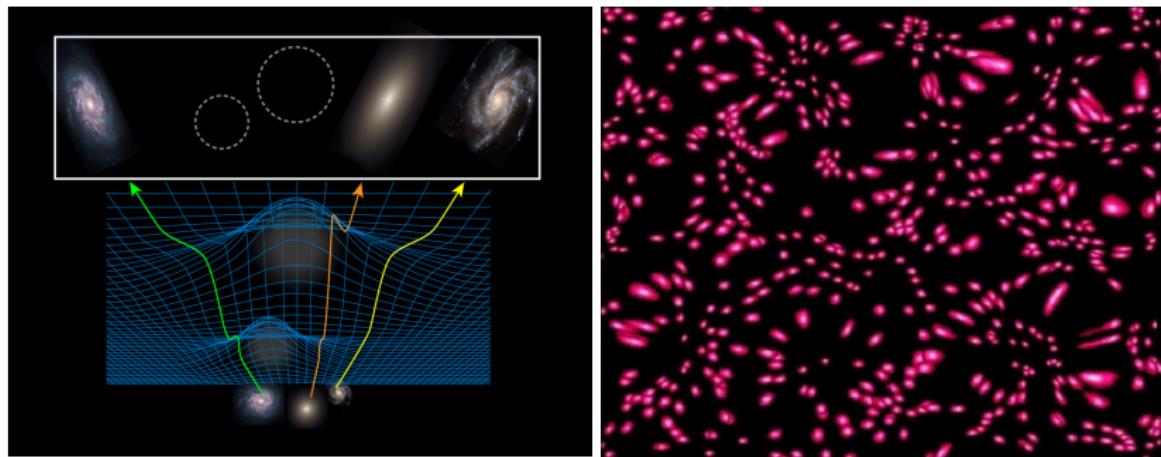
CONCLUSIONS  
OOOOOOOO

# VOID LENSING



OUTLINE  
OFINDING VOIDS  
OOVOID CLUSTERING  
OOOOOONON-GAUSSIANITY  
OOOOMASSIVE NEUTRINOS  
OOOOOOVOID LENSING  
OO●OOCONCLUSIONS  
OOOOOOO

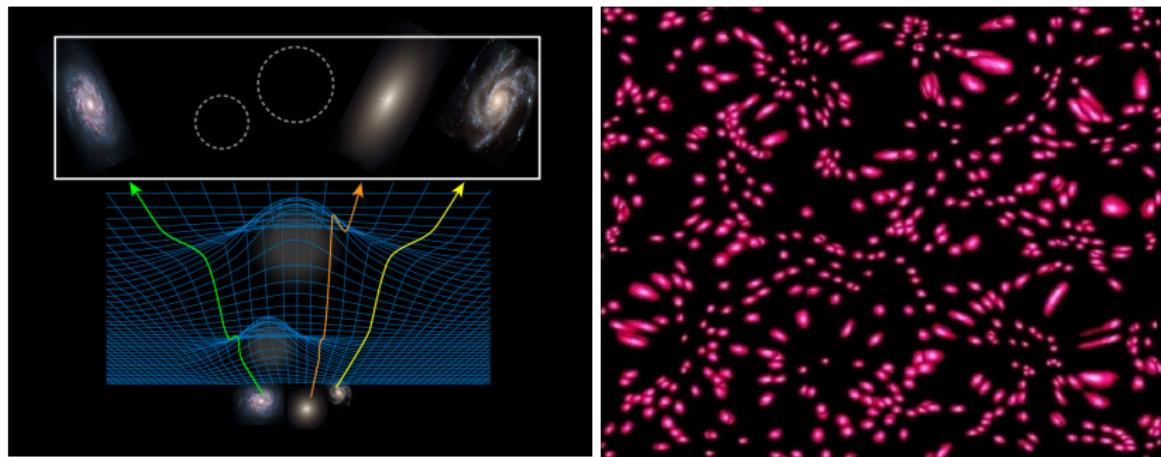
# VOID LENSING



$$\Delta\Sigma(r_p) \equiv \bar{\Sigma}(< r_p) - \Sigma(r_p) = \Sigma_{\text{crit}} \gamma_+(r_p), \quad \Sigma_{\text{crit}} = \frac{c^2}{4\pi G} \frac{D_A(z_s)}{D_A(z_l) D_A(z_l, z_s)}$$

OUTLINE  
OFINDING VOIDS  
OOVOID CLUSTERING  
OOOOOONON-GAUSSIANITY  
OOOOMASSIVE NEUTRINOS  
OOOOOOVOID LENSING  
OO●OOCONCLUSIONS  
OOOOOOO

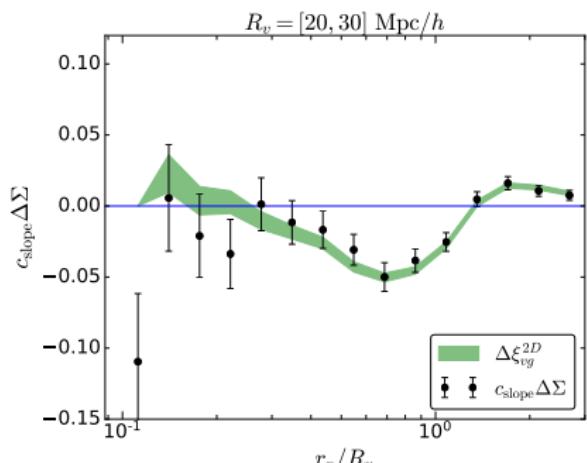
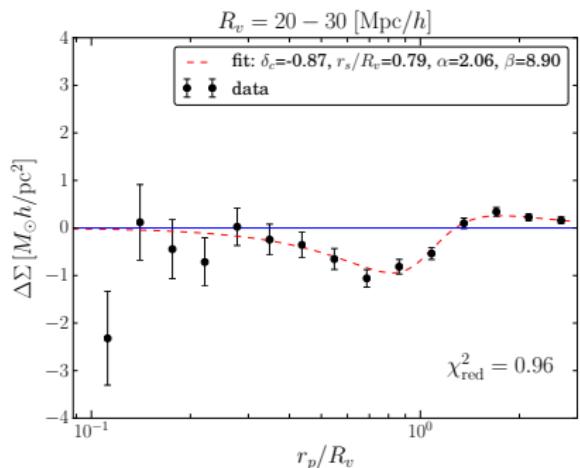
# VOID LENSING



$$\Delta\Sigma(r_p) \equiv \bar{\Sigma}(< r_p) - \Sigma(r_p) = \Sigma_{\text{crit}} \gamma_+(r_p), \quad \Sigma_{\text{crit}} = \frac{c^2}{4\pi G} \frac{D_A(z_s)}{D_A(z_l) D_A(z_l, z_s)}$$

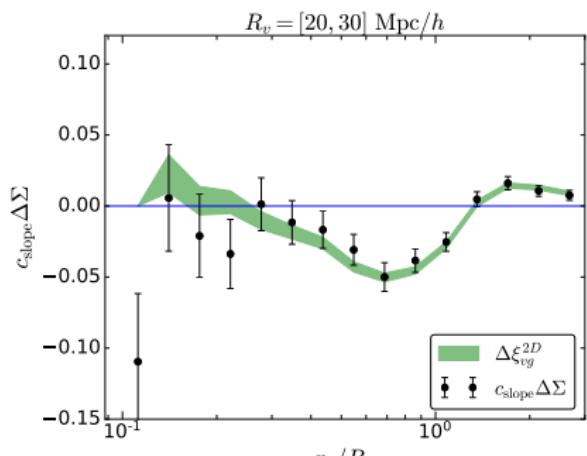
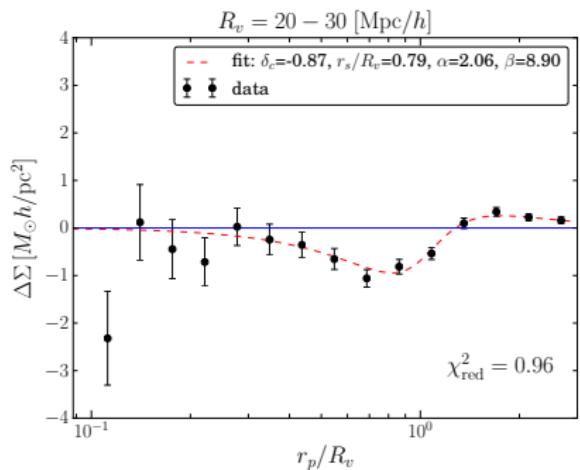
$$\Sigma(r_p) = \int \rho \left( \sqrt{[r_z - D_A(z_l)]^2 + r_p^2} \right) dr_z$$

# VOID LENSING: DES Y1



Fang, Hamaus et al. (2019, MNRAS 490, 3573)

# VOID LENSING: DES Y1



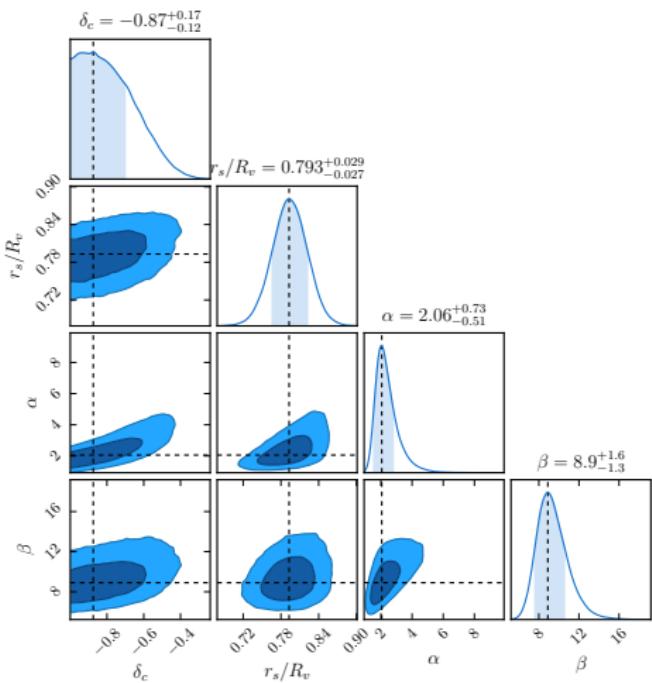
Fang, Hamaus et al. (2019, MNRAS 490, 3573)

Mass (dark matter + baryons) vs. Light (baryons)

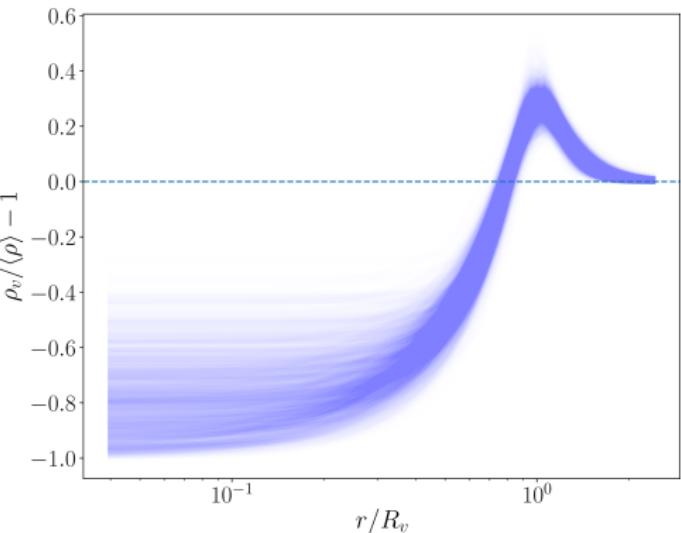
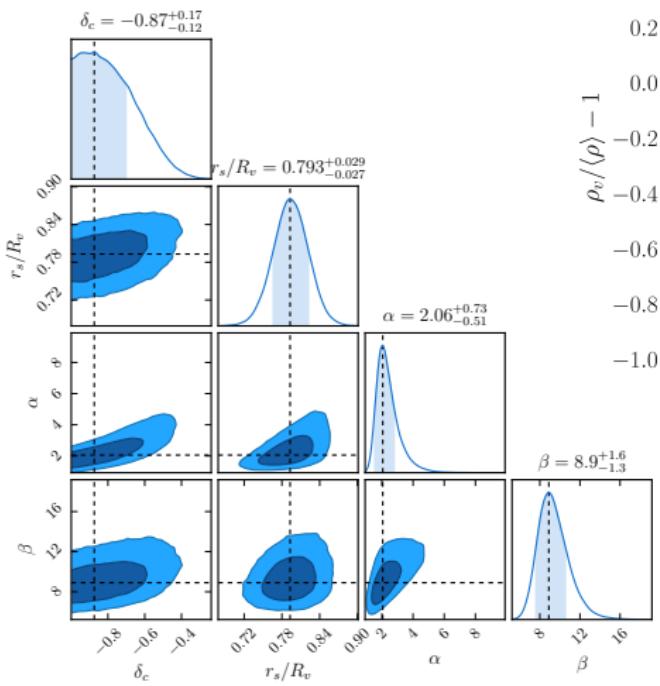
$$\left. \begin{aligned} \Delta\Sigma(r_p) &\equiv \bar{\Sigma}(< r_p) - \Sigma(r_p) \\ \Delta\xi_{\text{vg}}^{2D}(r_p) &\equiv \bar{\xi}_{\text{vg}}^{2D}(< r_p) - \xi_{\text{vg}}^{2D}(r_p) \end{aligned} \right\} \Rightarrow \Delta\xi_{\text{vg}}^{2D}(r_p) = \frac{b_g}{\langle \Sigma_m \rangle} \Delta\Sigma(r_p)$$

OUTLINE  
OFINDING VOIDS  
OOVOID CLUSTERING  
OOOOOONON-GAUSSIANITY  
OOOOMASSIVE NEUTRINOS  
OOOOOOVOID LENSING  
OOO●CONCLUSIONS  
OOOOOOO

# VOID LENSING: DES Y1



# VOID LENSING: DES Y1



$$\frac{\rho_v(r)}{\bar{\rho}} - 1 = \delta_c \frac{1 - (r/r_s)^\alpha}{1 + (r/r_v)^\beta}$$

Fang, Hamaus et al. (2019, MNRAS 490, 3573)

# CONCLUSIONS

- Voids are biased tracers of the LSS, like galaxies / 21cm / Ly- $\alpha$ , etc.

# CONCLUSIONS

- Voids are biased tracers of the LSS, like galaxies / 21cm / Ly- $\alpha$ , etc.
- Their linear bias extends far into the negative regime.

# CONCLUSIONS

- Voids are biased tracers of the LSS, like galaxies / 21cm / Ly- $\alpha$ , etc.
- Their linear bias extends far into the negative regime.
- Complementary clustering statistics enhance the attainable signal of  $f_{\text{NL}}$  &  $\sum m_\nu$ .

# CONCLUSIONS

- Voids are biased tracers of the LSS, like galaxies / 21cm / Ly- $\alpha$ , etc.
- Their linear bias extends far into the negative regime.
- Complementary clustering statistics enhance the attainable signal of  $f_{\text{NL}}$  &  $\sum m_\nu$ .
- A combination of void lensing and void clustering can offer observational constraints.

# CONCLUSIONS

- Voids are biased tracers of the LSS, like galaxies / 21cm / Ly- $\alpha$ , etc.
- Their linear bias extends far into the negative regime.
- Complementary clustering statistics enhance the attainable signal of  $f_{\text{NL}}$  &  $\sum m_\nu$ .
- A combination of void lensing and void clustering can offer observational constraints.
- Essentially for free, as voids are contained in survey data anyway.

# CONCLUSIONS

- Voids are biased tracers of the LSS, like galaxies / 21cm / Ly- $\alpha$ , etc.
- Their linear bias extends far into the negative regime.
- Complementary clustering statistics enhance the attainable signal of  $f_{\text{NL}}$  &  $\sum m_\nu$ .
- A combination of void lensing and void clustering can offer observational constraints.
- Essentially for free, as voids are contained in survey data anyway.

Thank you !

OUTLINE

FINDING VOIDS

VOID CLUSTERING

NON-GAUSSIANITY

MASSIVE NEUTRINOS

VOID LENSING

CONCLUSIONS

O

OO

OOOOOO

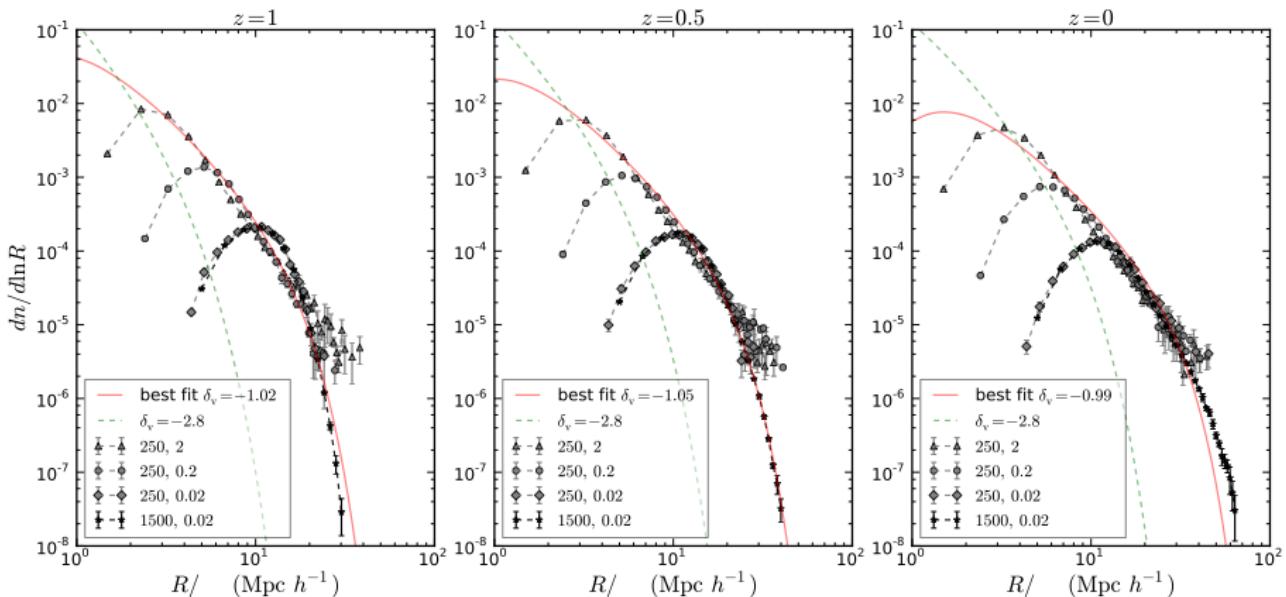
OOOO

OOOOOO

OOOO

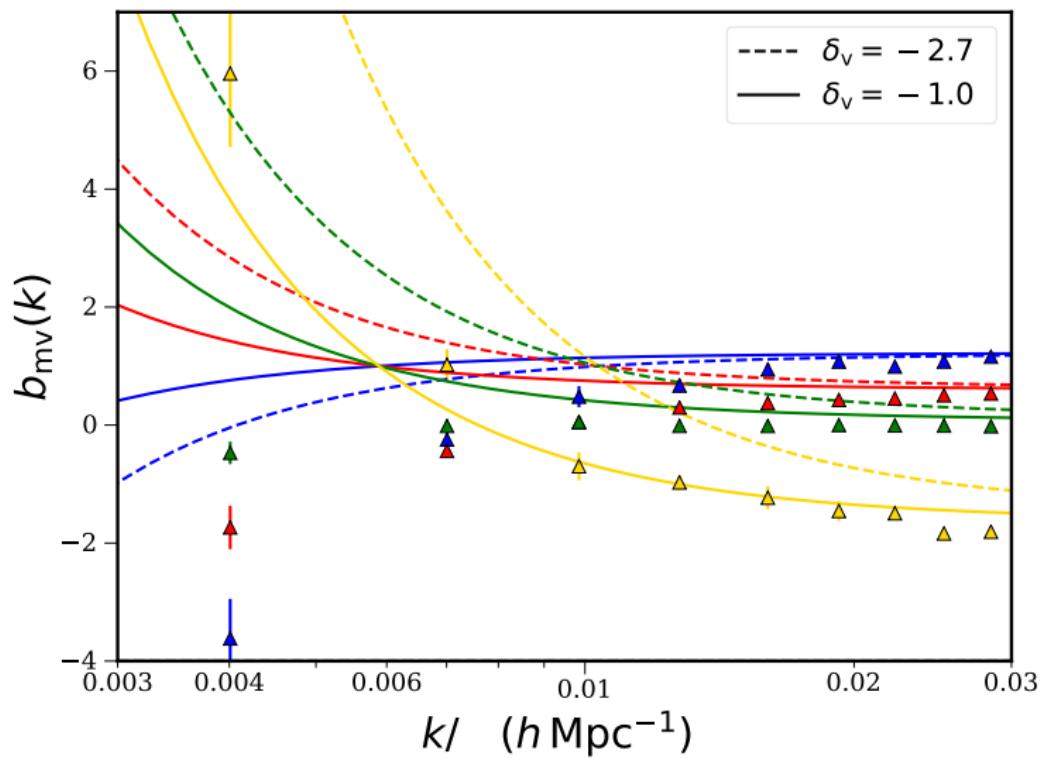
O●OOOO

# VOID ABUNDANCE



Chan, Hamaus, Desjacques (2014, PRD 90, 103521)

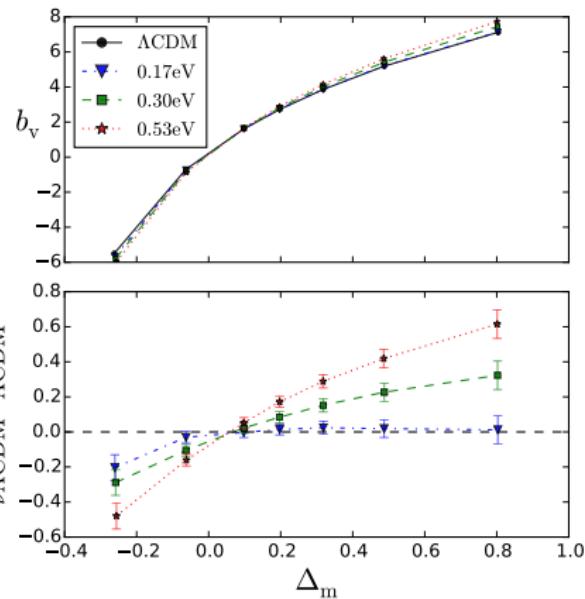
# SCALE-DEPENDENT VOID BIAS



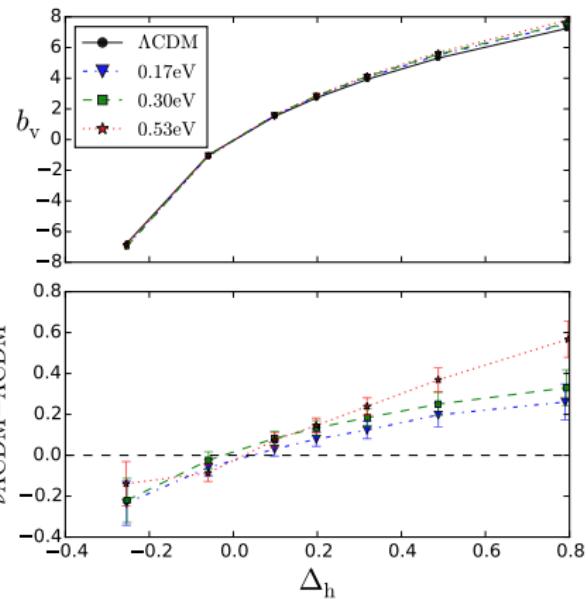
Chan, Hamaus, Biagetti (2019, PRD 99, 121304)

# LINEAR VOID BIAS

Dark matter voids

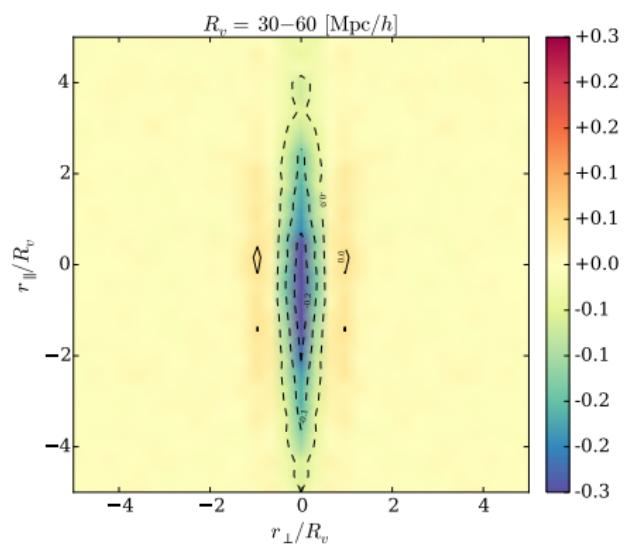
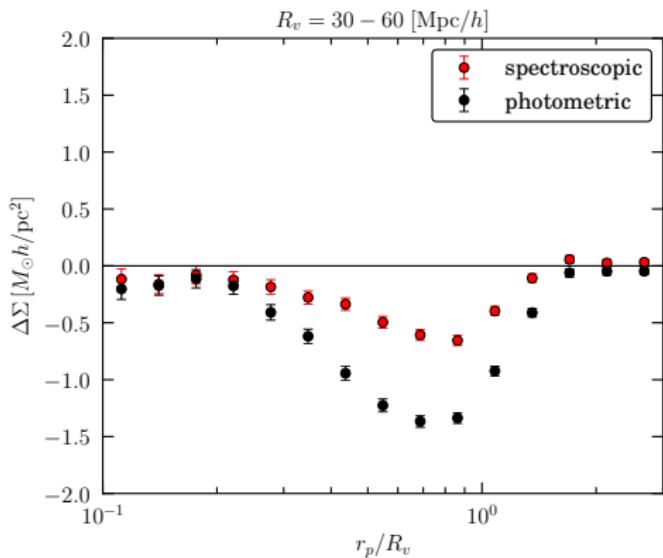


Halo voids



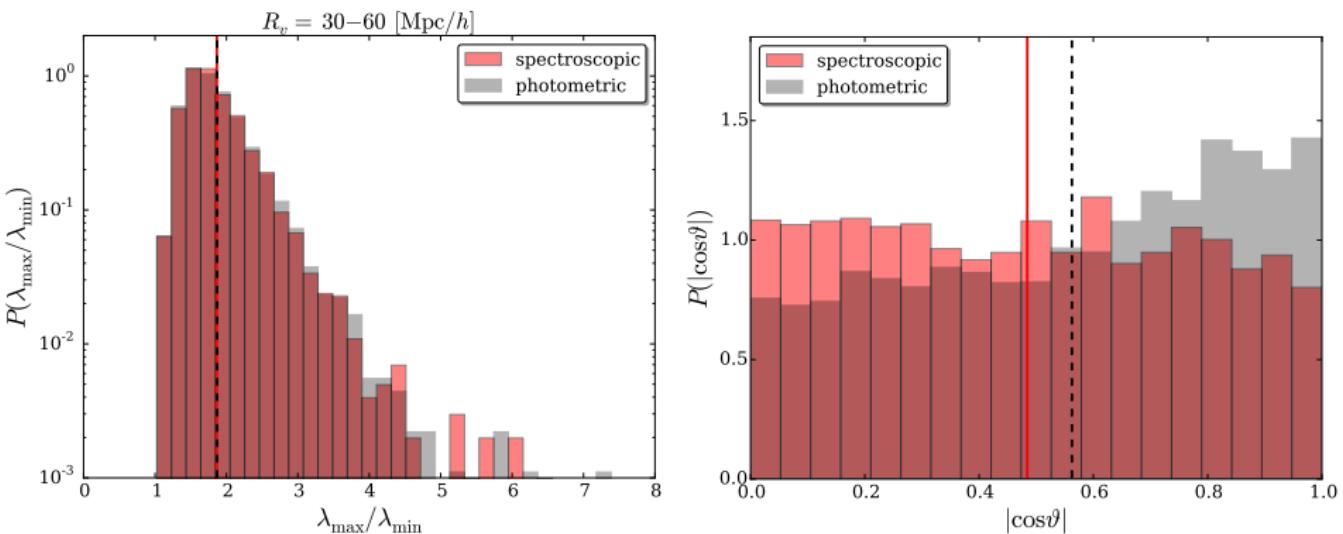
Schuster, Hamaus et al. (2019, JCAP in press, arXiv:1905.00436)

# VOID LENSING: DES Y1 MOCKS



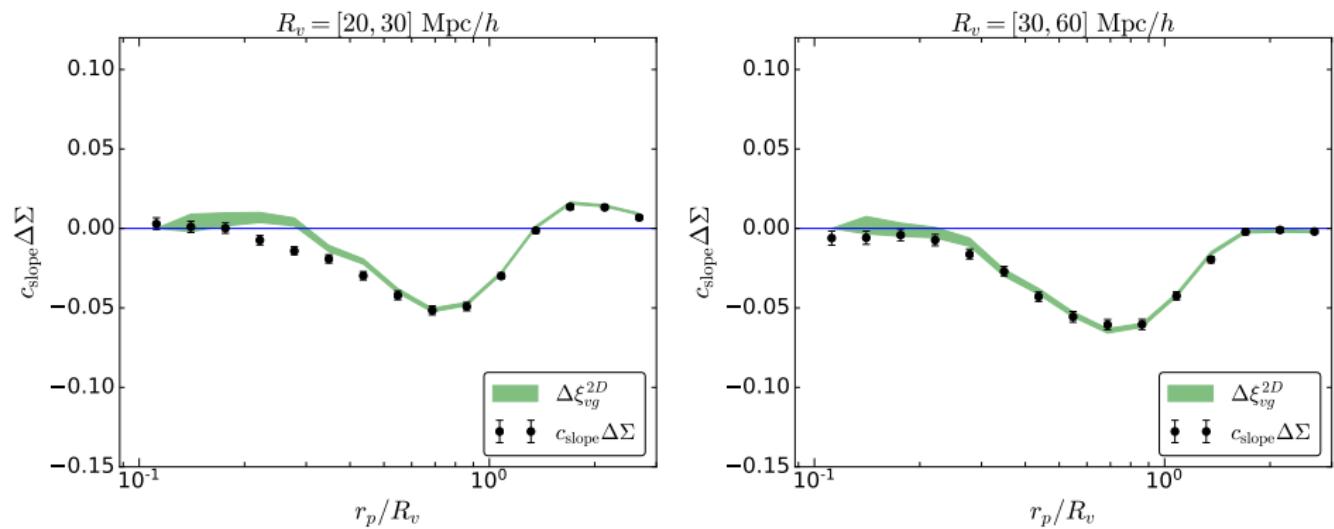
Fang, Hamaus et al. (2019, MNRAS 490, 3573)

# VOID LENSING: DES Y1 MOCKS



Fang, Hamaus et al. (2019, MNRAS 490, 3573)

# VOID LENSING: DES Y1 MOCKS



Fang, Hamaus et al. (2019, MNRAS 490, 3573)