

Edgeworth Streaming Model for redshift space distortions

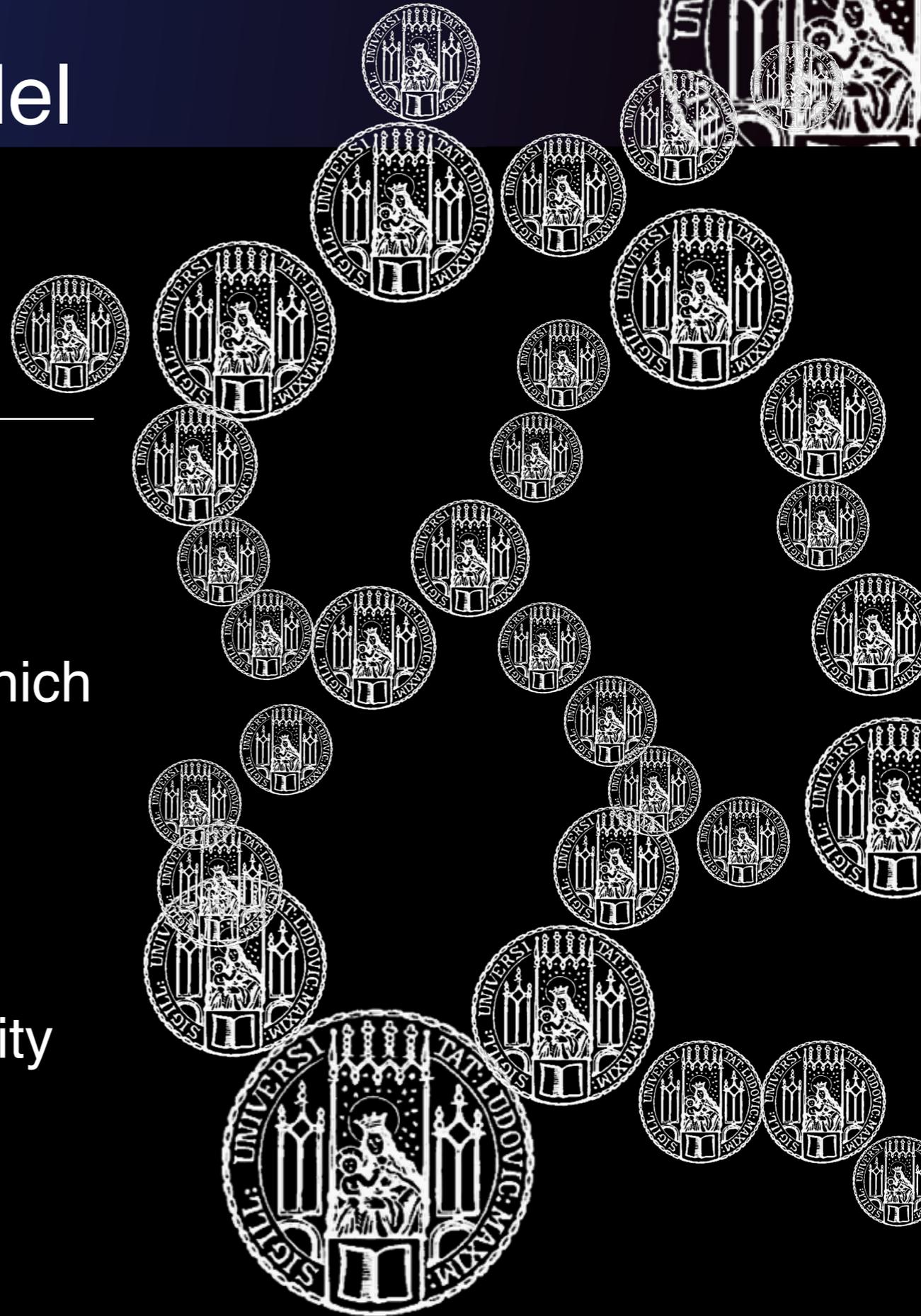
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& Excellence Cluster Universe

in collaboration with

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PhD advisor: Stefan Hofmann



Redshift space correlation function



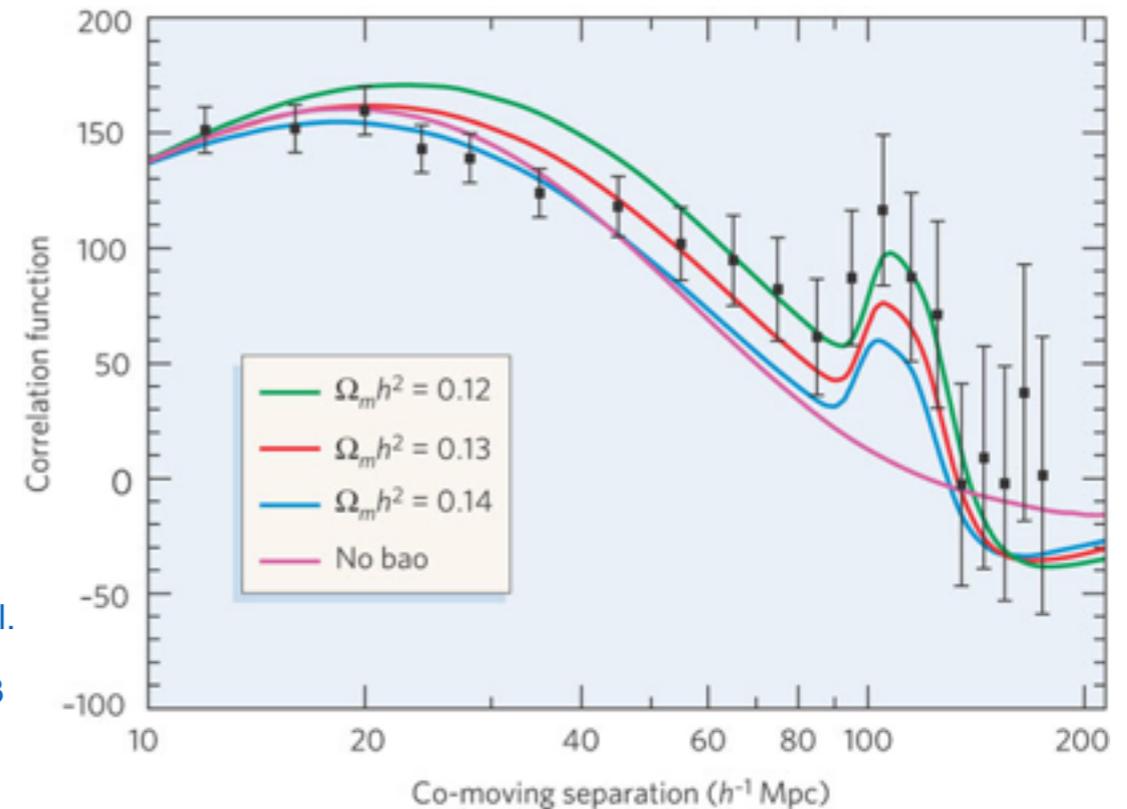
Correlation function

- measures excess probability

$$1 + \xi_X(s) = \left\langle (1 + \delta_X(s_1))(1 + \delta_X(s_2)) \right\rangle$$

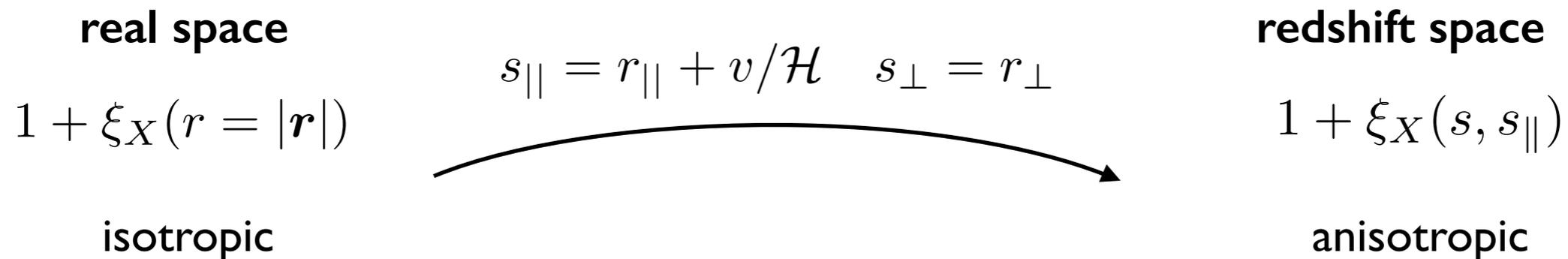
- halos as biased DM tracers
 - input for halo model
 - powerful cosmological probe

Eisenstein et al.
2005 ApJ, 633



Redshift space distortions

- redshift observations affected by peculiar velocities



Redshift space correlation function



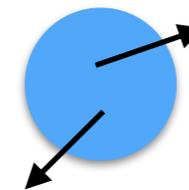
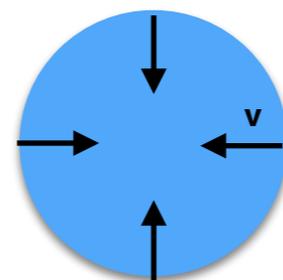
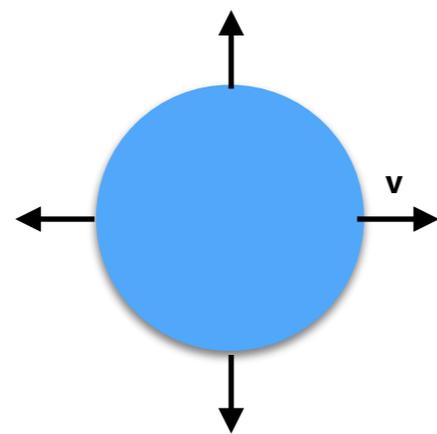
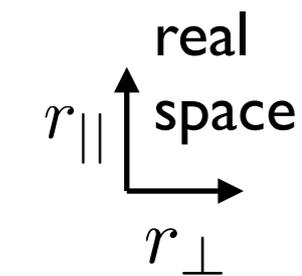
Redshift space distortions

- observations in redshift space from galaxy surveys
- impact of peculiar velocities along line of sight

$$s_{\parallel} = r_{\parallel} + v/\mathcal{H} \quad s_{\perp} = r_{\perp}$$

linear evolution

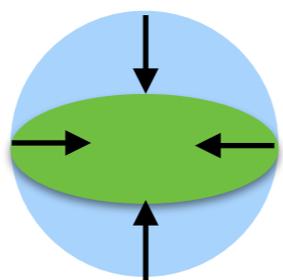
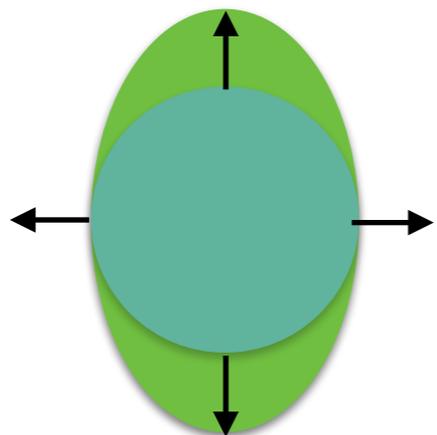
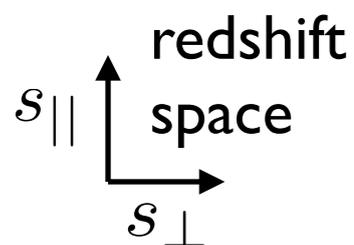
nonlinear structure



underdensity

overdensity

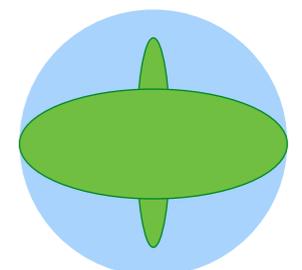
streaming model



+



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Streaming model



Gaussian Streaming Model

$$1 + \xi_X(s_{\parallel}, s_{\perp}, t) = \int_{-\infty}^{\infty} \frac{dr_{\parallel}}{\sqrt{2\pi}\sigma_{12}} \underbrace{(1 + \xi_X(r, t))}_{\text{real space correlation}} \exp \left[\underbrace{-\frac{(s_{\parallel} - r_{\parallel} - v_{12}(r, t)r_{\parallel}/r)^2}{2\sigma_{12}^2(r, r_{\parallel}, t)}}_{\text{Gaussian pairwise velocity distribution mean pairwise velocity \& dispersion}} \right]$$

Fisher (1995, *Astrophys.J.* 448,
Reid & White (2011, *MNRAS* 417)

- Edgeworth expansion around a Gaussian PDF
- similar as for density δ or velocity divergence θ in SPT
Juszkiewicz, Chodorowski et. al (*APJ* 442, 39 ,1995)

Streaming model



Edgeworth Streaming Model

$$1 + \xi_X(s_{\parallel}, s_{\perp}, t) = \int_{-\infty}^{\infty} \frac{dr_{\parallel}}{\sqrt{2\pi}\sigma_{12}} \underbrace{(1 + \xi_X(r, t))}_{\text{real space correlation}} \exp \left[\underbrace{-\frac{(s_{\parallel} - r_{\parallel} - v_{12}(r, t)r_{\parallel}/r)^2}{2\sigma_{12}^2(r, r_{\parallel}, t)}}_{\text{Gaussian pairwise velocity distribution mean pairwise velocity \& dispersion}} \right]$$

CU, Kopp and Haugg (2015, arXiv: 1503.08837)

$$\times \left(1 + \frac{\Lambda_{12}}{6\sigma_{12}^3} \left[\left(\frac{\Delta_{srv}}{\sigma_{12}} \right)^3 - 3 \frac{\Delta_{srv}}{\sigma_{12}} \right] \right)$$

first non-Gaussian correction
pairwise velocity skewness

- Edgeworth expansion around a Gaussian PDF
- similar as for density δ or velocity divergence θ in SPT

Juszkiewicz, Chodorowski et. al (APJ 442, 39 ,1995)

Streaming model

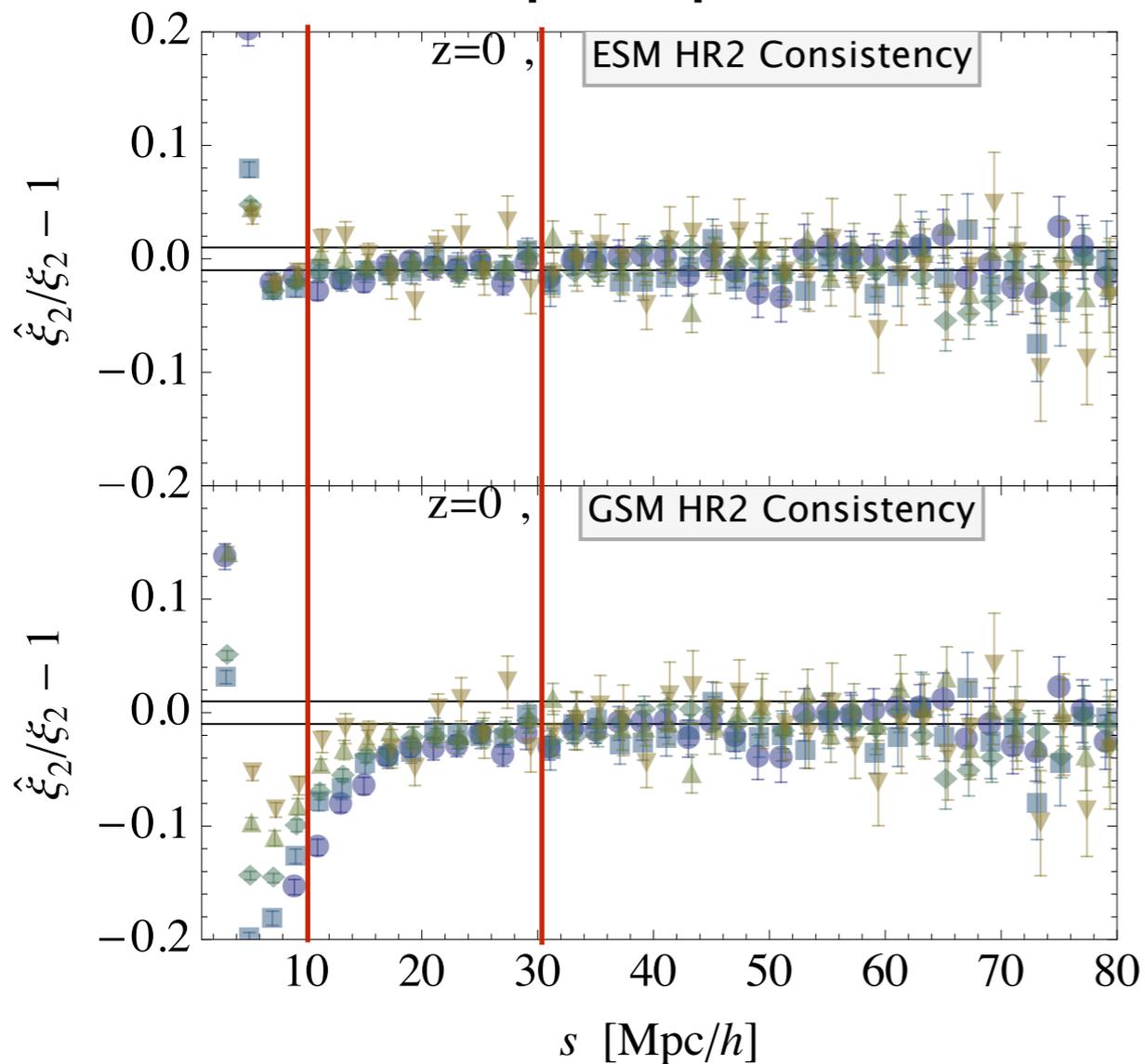


Accuracy of Streaming Models

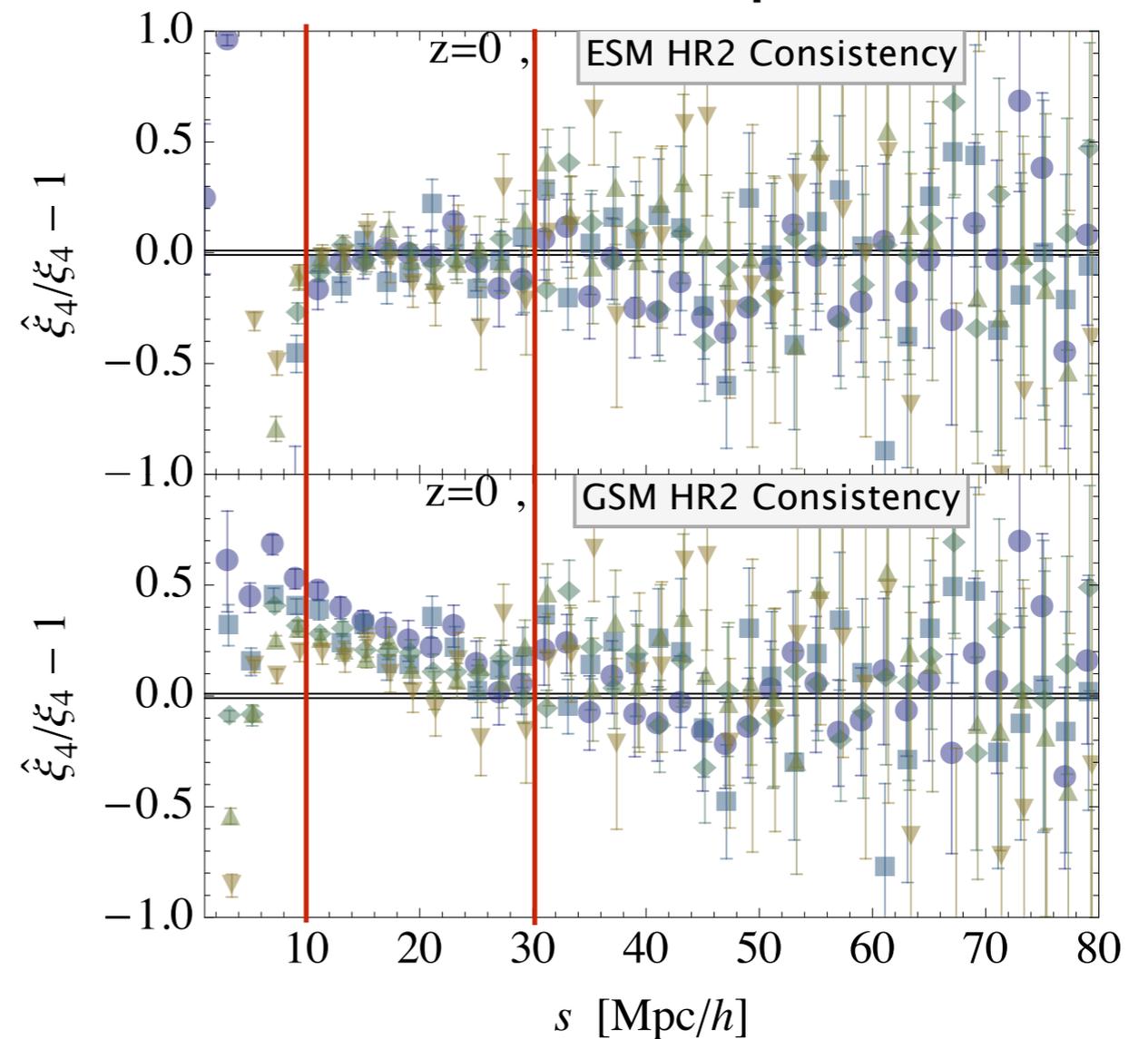
- compare multipoles of redshift space correlation function
- **model independent**: measure ingredients in halo catalog
- 2% down to 10 Mpc/h (ESM) vs. 30 Mpc/h (GSM)

Horizon Run 2
halo catalog
mass range $\lg M \sim 13 - 15$

quadrupole



hexadecapole



Streaming model ingredients



Gaussian Streaming Model

$$1 + \xi_X(s_{\parallel}, s_{\perp}, t) = \int_{-\infty}^{\infty} \frac{dr_{\parallel}}{\sqrt{2\pi}\sigma_{12}(r, r_{\parallel}, t)} (1 + \xi_X(r, t)) \exp \left[-\frac{(s_{\parallel} - r_{\parallel} - v_{12}(r, t)r_{\parallel}/r)^2}{2\sigma_{12}^2(r, r_{\parallel}, t)} \right]$$

real space correlation

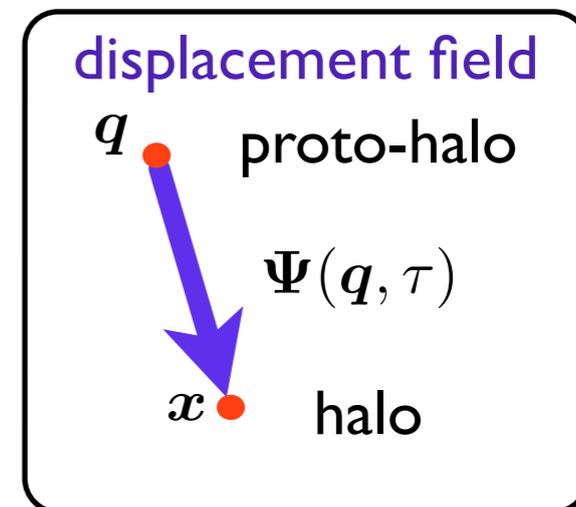
Gaussian velocity distribution
mean pairwise velocity & dispersion

Wang, Reid & White (2014, MNRAS 437)

Lagrangian Perturbation Theory
+ local Lagrangian bias: $\xi \rightarrow \xi_X$

- **Zel'dovich approximation** Zel'dovich (1970, A&A 5, 84)
 - 1st order Lagrangian PT
 - physically motivated resummation of SPT
- **Post Zel'dovich approximation**
 - higher order Lagrangian PT
 - partial resummation: Convolution LPT

Carlson et al. (2012, MNRAS 429)



Streaming model ingredients



Gaussian Streaming Model

real space correlation

$$1 + \xi_X(s_{\parallel}, s_{\perp}, t) = \int_{-\infty}^{\infty} \frac{dr_{\parallel}}{\sqrt{2\pi}\sigma_{12}(r, r_{\parallel}, t)} (1 + \xi_X(r, t)) \exp \left[-\frac{(s_{\parallel} - r_{\parallel} - v_{12}(r, t)r_{\parallel}/r)^2}{2\sigma_{12}^2(r, r_{\parallel}, t)} \right]$$

Gaussian velocity distribution
mean pairwise velocity & dispersion

$$\exp \left[-\frac{(s_{\parallel} - r_{\parallel} - v_{12}(r, t)r_{\parallel}/r)^2}{2\sigma_{12}^2(r, r_{\parallel}, t)} \right]$$

- **Why smoothing?**
 - implement halo size in fluid description
 - improves Zel'dovich predictions in N-body
- **truncated Zel'dovich approximation**
 - Zel'dovich with smoothed input PS
 - improves agreement with N-body

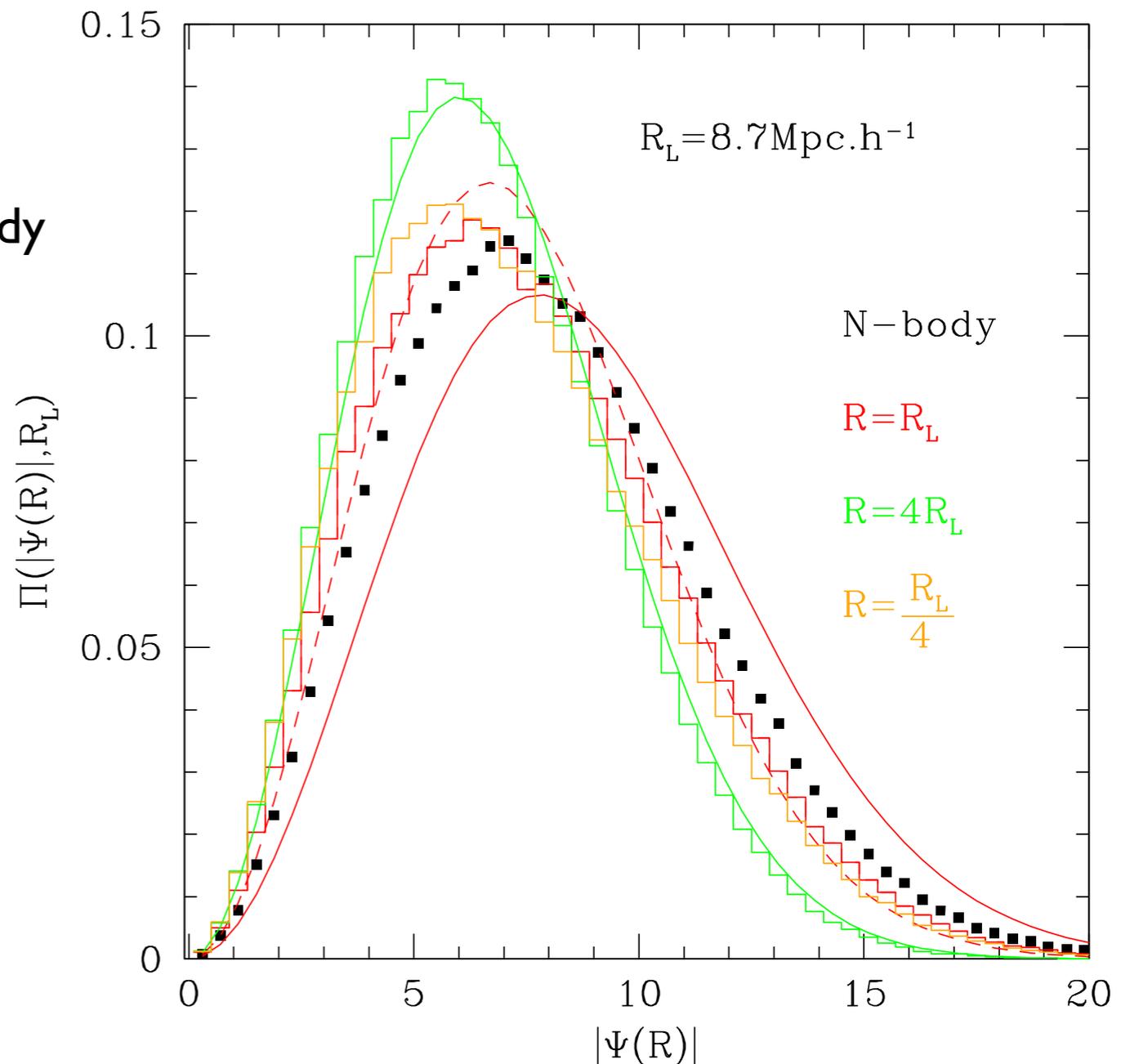
Coles, Melott, Shandarin (1993, MNRAS 260)

coarse-grained dust model

CU, Kopp & Haugg (arXiv: 1503.08837)

truncated CLPT

Kopp, CU & Aчитouv (in preparation)



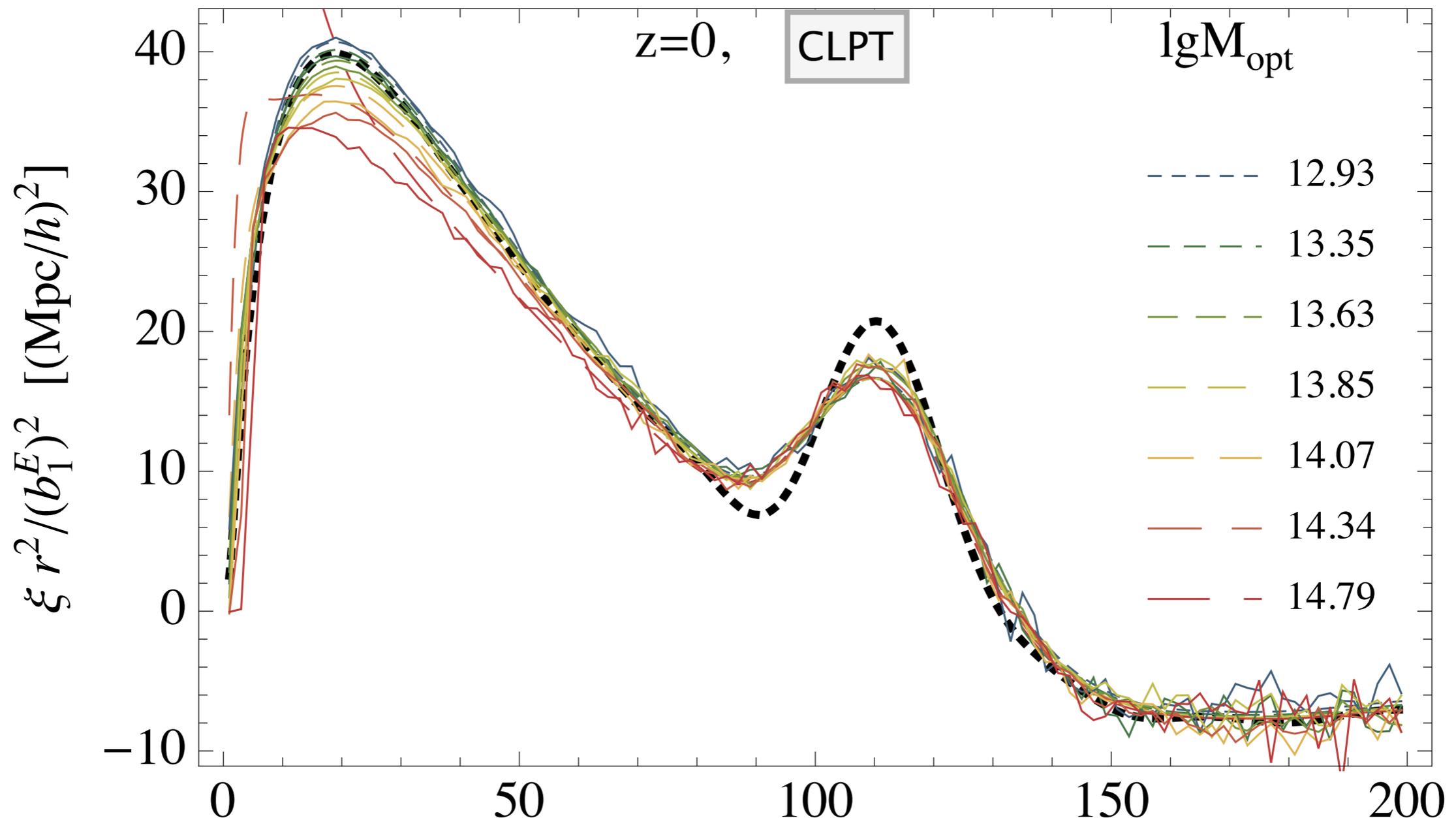
Redshift space distortions



Streaming parameters from truncated CLPT

Real space halo correlation $\xi(r)$: best agreement for $1 \text{ Mpc}/h$

- smoothing in $R(M)$ worse: need to include peak bias [Baldauf, Desjacques & Seljak \(arXiv: 1405.5885\)](#)



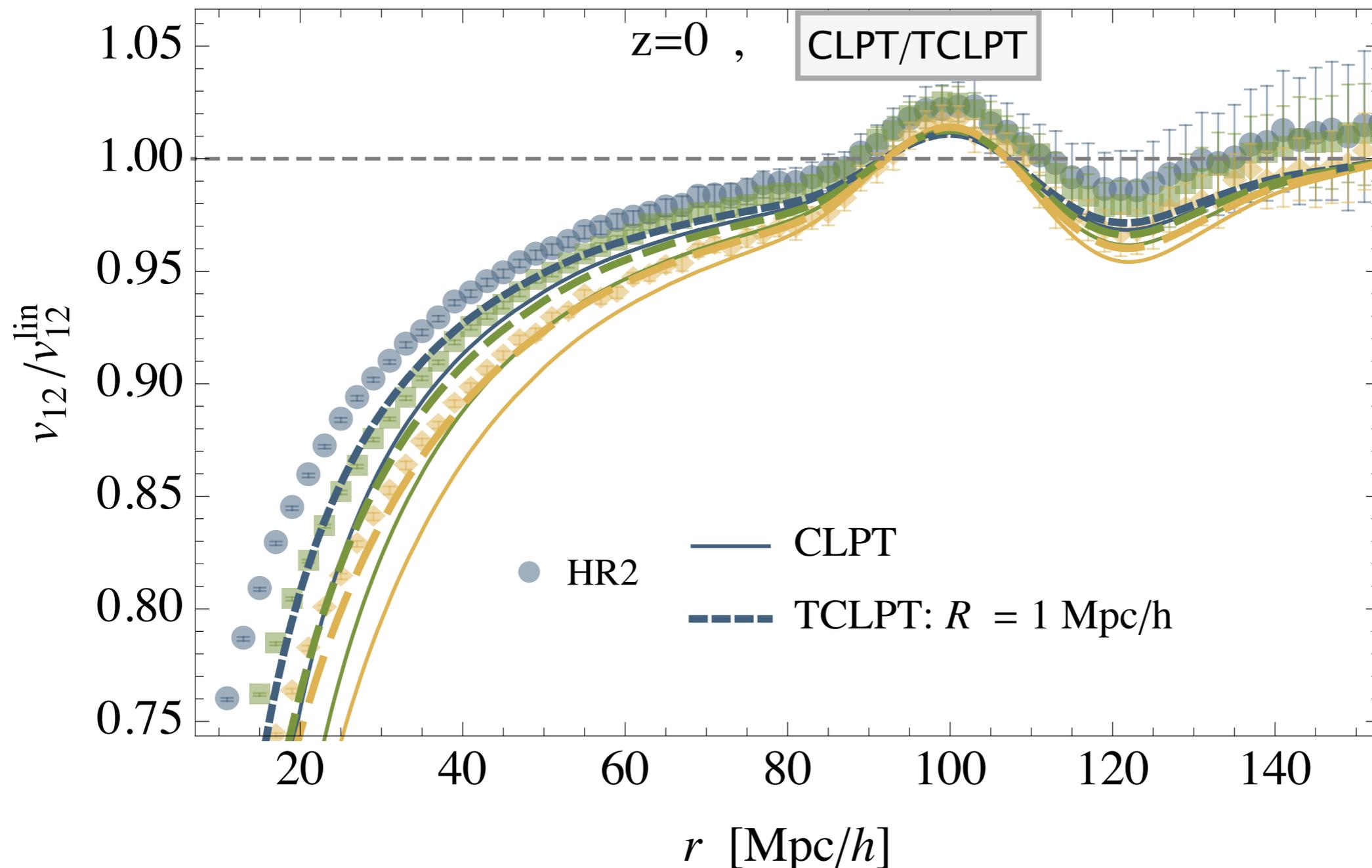
Redshift space distortions



Streaming parameters from truncated CLPT

Pairwise velocity $v_{12}(r)$: best agreement for 1 Mpc/h

- smoothing in $R(M)$ worse for high M - velocity bias?



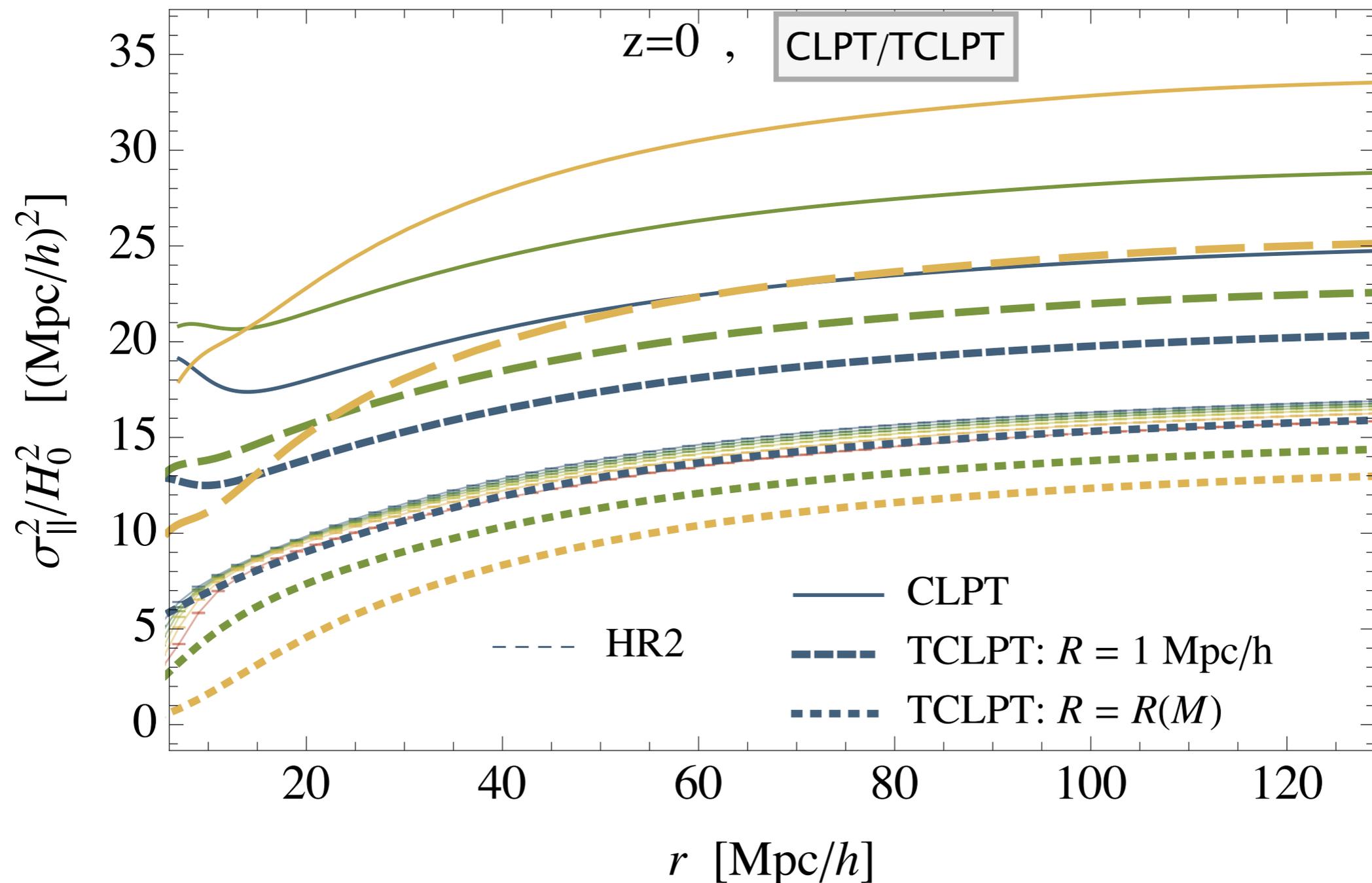
Redshift space distortions



Streaming parameters from truncated CLPT

Pairwise velocity dispersion $\sigma_{12}(r)$: best agreement for R(M)

- huge improvement of the overall amplitude



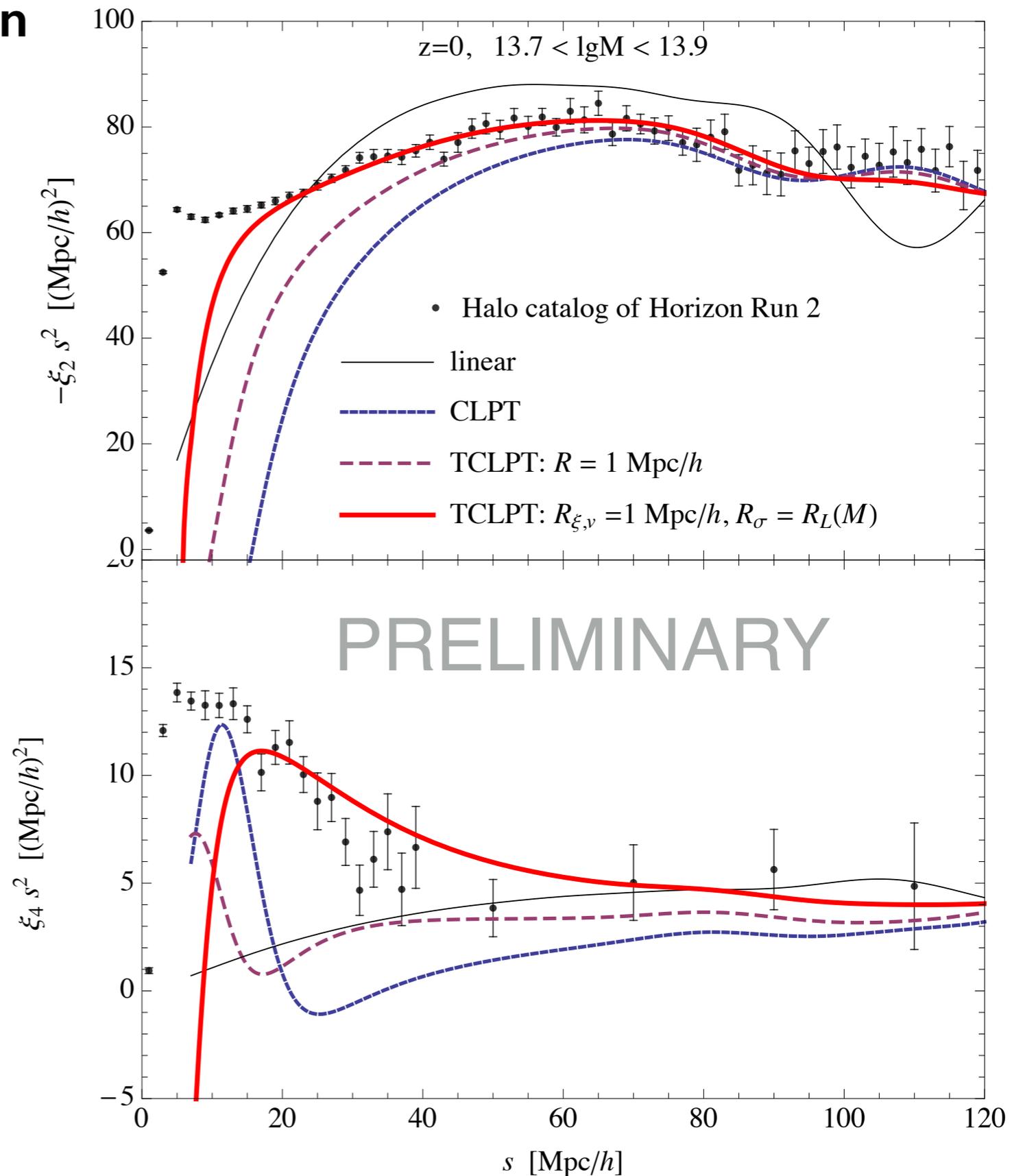
Streaming model predictions



Redshift space correlation function

Kopp, CU & Achitouv (in preparation)

- plug streaming model ingredients in
obtain redshift space multipoles
 - monopole $\xi_0(s)$
 - quadrupole $\xi_2(s)$
 - hexadecapole $\xi_4(s)$
- TCLPT outperforms CLPT
 - simultaneously improves all
higher redshift-space multipoles
- **optimal: two-filter TCLPT smoothing**
 - $\xi(r) & v_{12}(r)$: 1 Mpc/h
 - $\sigma_{12}(r)$: Lagrangian scale $R(M)$



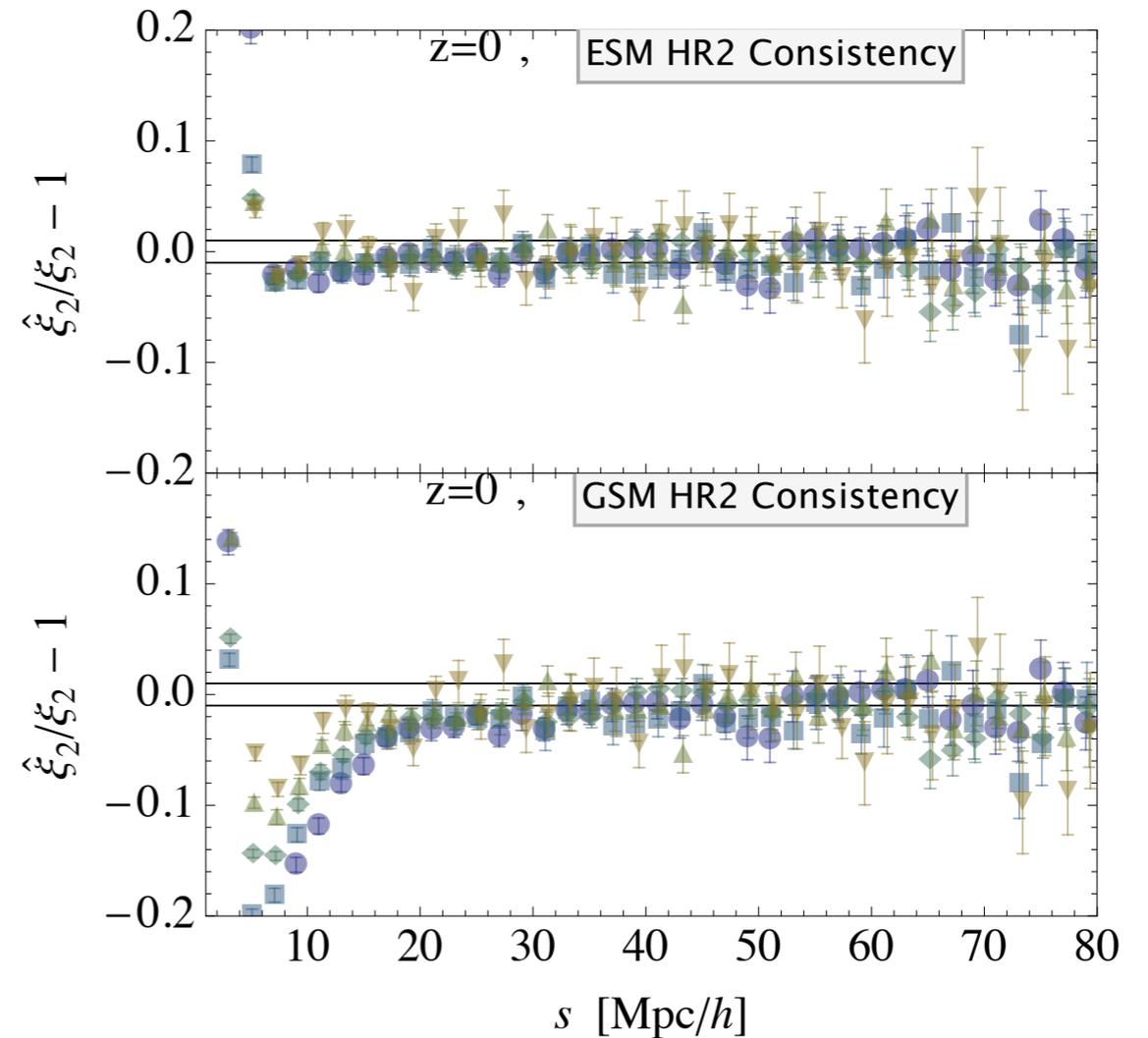
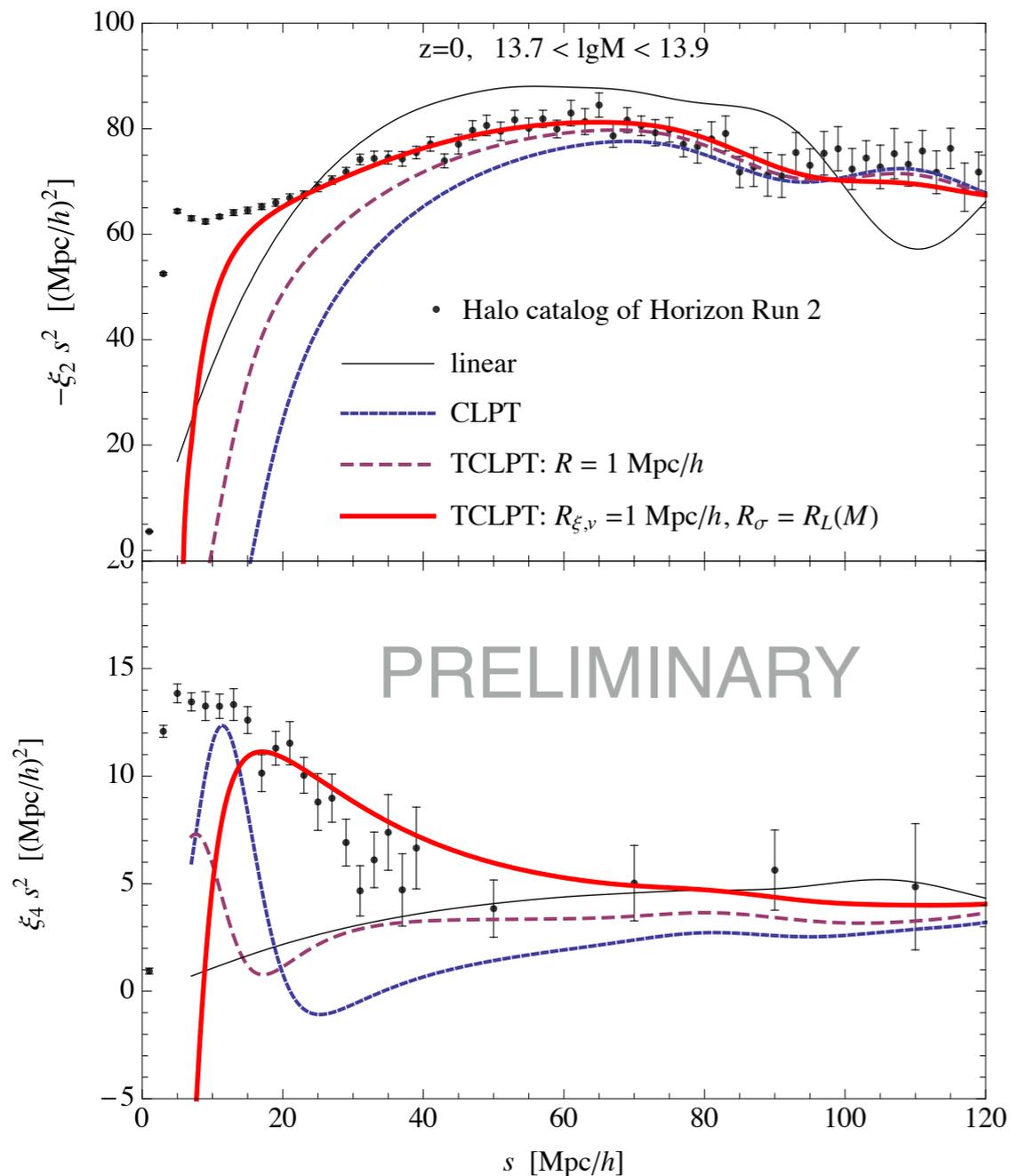
Summary



Edgeworth streaming model

- generalization of Gaussian streaming model
- pushed 2% accuracy from 30 down to 10 Mpc/h

CU, Kopp and Haugg (2015, arXiv: 1503.08837)



Truncated Zel'dovich approximation

- truncated Post-Zel'dovich approximation (TCLPT)
 - optimal with two filters: 1 Mpc/h & $R(M)$
 - consistent results for $\xi_0, \xi_2, \xi_4(s)$
- peak bias effects relevant

Kopp, CU & Aчитouv (in preparation)