

# Nonlinear reconstruction of redshift space distortions

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We apply nonlinear reconstruction to the dark matter density field in redshift space and solve for the nonlinear mapping from the initial Lagrangian positions to the final redshift space positions. The reconstructed anisotropic field inferred from the nonlinear displacement correlates with the linear initial conditions to much smaller scales than the redshift space density field. The number of linear modes in the density field is improved by a factor of  $30 \sim 40$ . We thus expect this reconstruction approach to substantially expand the cosmological information including BAO and RSD for dense low-redshift large scale structure surveys including for example SDSS main sample, DESI BGS, and 21 cm intensity mapping initiatives.

## I. INTRODUCTION

Measuring the three-dimensional large-scale structure of the Universe provides a powerful method to probe particle physics and cosmology. Precision measurements of baryon acoustic oscillations (BAO) and redshift space distortions (RSD) can constrain dark energy models and modified theories of gravity [1–9]. The ongoing and upcoming surveys are mapping large swaths of the visible Universe [1, 10–13]. However, the precision of the measured cosmological parameters (e.g. the BAO scale and the growth rate) is usually limited by the strong non-Gaussianities of the nonlinear data, which prevent a simple mapping to the linear initial conditions which are predicted by cosmological theories. Therefore, to better extract cosmological information from an observed nonlinear map, it is crucial to understand the nonlinearities that deviate the observed signatures from the theoretical predictions and how to reduce these nonlinearities in the nonlinear map to obtain better statistics.

Reconstruction methods have been developed to reverse the nonlinear degradation of the BAO signal [14–17]. The current standard BAO reconstruction method uses a linear mapping to reduce the shift nonlinearities in the observed nonlinear field [14]. The linear method is successfully applied to galaxy surveys and improves the

measurements of the BAO scale [2, 3, 18–23]. However, the density field reconstruction methods based on the linearized continuity equation only captures the shift terms from large-scale linear bulk flows [24–29]. Recently a new reconstruction method has been proposed by Ref. [30]. (See also Refs. [31, 32] for different implementations but based on the same principle.) The new method solves a nonlinear bijective mapping between the Eulerian coordinate system and a new coordinate system, where the mass per volume element is constant [30]. The nonlinear mapping provides a significantly better estimate of the nonlinear displacement compared to standard reconstruction. The linear density field inferred from the nonlinear bijective mapping gives a much better correlation with the linear initial conditions [30]. These nonlinear methods are expected to substantially improve BAO measurements in the near future.

Most nonlinear reconstruction methods developed recently focus on improving measurements of the linear BAO signal [30–33]. The current standard reconstruction also only works for reversing the nonlinear degradation of the BAO feature in galaxy redshift surveys, though different conventions for the RSD effect are adopted. (See Ref. [29] for more discussions about different RSD conventions.) The RSD effect is also one of the most powerful probes of the Universe, complementary to the BAO measurements. Precise measurements of the growth rate of structure with the RSD effect can provide tests of General Relativity. While the distinct BAO feature can be measured easily, extracting information from RSD measurements requires the modeling of the power spectrum

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amplitude. Thus the analysis of RSD is often limited to much smaller wavenumbers (e.g.  $k < 0.15 \ h/\text{Mpc}$  compared to  $k < 0.3 \ h/\text{Mpc}$  for the BAO-only analysis [2, 5]). A lot of efforts have been made to model the nonlinear RSD effect [34–61]. However, even if the nonlinear RSD effect can be modeled to nonlinear scales, this still does not improve the power spectrum information content of a nonlinear map. Present RSD analysis is limited to the strong nonlinearities of the density and velocity fields and the nonlinear mapping from real to redshift space, with a small amount of cosmological information being extracted from galaxy surveys. Another possible way to handle the nonlinear RSD effect would be trying to reduce non-Gaussianities of the nonlinear density field in redshift space, in order to unscramble the cosmological information encoded in higher order statistics of the nonlinear data and increase the power spectrum information content of the reconstructed linear map.

In this paper, we apply nonlinear reconstruction to the dark matter density field in redshift space and solve for the nonlinear mapping from the initial Lagrangian positions to the final redshift space positions. The reconstructed anisotropic field inferred from the nonlinear displacement correlates with the linear initial conditions to much smaller scales than the redshift space density field. Therefore, we are able to significantly increase the constraining power of anisotropic galaxy clustering measurements by including small scale density fluctuations in the analysis and provide tighter constraints on the growth rate of structure. The idea of RSD reconstruction proposed here would be crucial for improving the RSD measurements from current and future galaxy surveys.

This paper is organized as follows. In Section II, we describe the motivation for reconstruction of RSD. Section III describes the performance tests of reconstruction and shows the reconstruction results. In Section IV, we describe some applications of reconstruction. We conclude in Section V. Appendix A shows the statistical properties of Lagrangian displacements. In Appendix B and C, we present the study of Lagrangian velocities and shift velocities. Appendix D describes how to implement nonlinear reconstruction in the presence of mask. Appendix E shows the RSD reconstruction results for halos.

## II. PHYSICAL MOTIVATION

The Lagrangian space perturbation theories which capture the nonlinear shift terms provide a successful description of the matter distribution on large scales [62–73]. In the Lagrangian picture of structure formation, the Eulerian position  $\mathbf{x}$  of each mass element is given by the sum of its initial Lagrangian coordinate  $\mathbf{q}$  and the subsequent displacement  $\boldsymbol{\psi}(\mathbf{q})$ :

$$\mathbf{x}(\mathbf{q}) = \mathbf{q} + \boldsymbol{\psi}(\mathbf{q}). \quad (1)$$

The displacement field which defines the mapping from initial Lagrangian coordinates to Eulerian coordinates is

the key variable in Lagrangian space based perturbation theories. Since the initial density field is almost uniform, the nonlinear density at the Eulerian position  $\mathbf{x}$  is

$$\rho(\mathbf{x})d^3x = \rho(\mathbf{q})d^3q, \quad (2)$$

where  $\rho(\mathbf{q}) \approx \text{constant}$ . The mass conservation equation predicts the exact nonlinear density  $\rho(\mathbf{x}) = d^3q/d^3x - 1$  at  $\mathbf{x}$  in the absence of shell crossing. Therefore, the key ingredient of Lagrangian perturbation theory calculation is to solve the nonlinear displacement field and calculate the Jacobian of the mapping from  $\mathbf{q}$  to  $\mathbf{x}$ .

The redshift space position  $\mathbf{s}$  of an object is shifted from its real space position  $\mathbf{x}$  by its peculiar velocity,

$$\mathbf{s}(\mathbf{x}) = \mathbf{x} + \frac{v_z(\mathbf{x})}{aH}\hat{z}, \quad (3)$$

where  $a$  is the scale factor and  $H$  is the Hubble parameter. In this paper we adopt the plane-parallel approximation and take the line of sight direction to be the  $z$ -axis. The density field in redshift space can be obtained by imposing mass conservation,

$$\rho(\mathbf{s})d^3s = \rho(\mathbf{x})d^3x. \quad (4)$$

Since the growth of displacement  $\boldsymbol{\psi}(\mathbf{q})$  is the *Lagrangian velocity*  $\mathbf{v}(\mathbf{q})$ , the Eulerian velocity  $\mathbf{v}(\mathbf{x})$  of a mass element at  $\mathbf{x} = \mathbf{q} + \boldsymbol{\psi}(\mathbf{q})$  is simply the Lagrangian velocity  $\mathbf{v}(\mathbf{q})$ . Thus we can define a mapping from the initial Lagrangian coordinates  $\mathbf{q}$  to the redshift space coordinates  $\mathbf{s}$  as

$$\mathbf{s}(\mathbf{q}) = \mathbf{q} + \boldsymbol{\psi}^s(\mathbf{q}), \quad (5)$$

where

$$\boldsymbol{\psi}^s(\mathbf{q}) = \boldsymbol{\psi}(\mathbf{q}) + \mathbf{v}^s(\mathbf{q})/aH. \quad (6)$$

Here  $\mathbf{v}^s(\mathbf{x}) \equiv [\hat{z} \cdot \mathbf{v}(\mathbf{x})]\hat{z}$  is called the *shift velocity*. The nonlinear density field in redshift space at  $\mathbf{s}$  is

$$\rho(\mathbf{s})d^3s = \rho(\mathbf{q})d^3q. \quad (7)$$

Again the nonlinear density at  $\mathbf{s}$  predicted by Eq. (7) with the redshift space displacement field  $\boldsymbol{\psi}^s(\mathbf{q})$  is exact before shell crossing happens.

The nonlinear mapping from real to redshift space has been studied extensively, which is necessary to model the nonlinear RSD effect. A successful nonlinear RSD model can reduce theoretical systematics in the analysis. However, the cosmological information extracted from redshift space two-point statistics is still limited by the shift nonlinearities in the observed density field. In order to remove the nonlinear shift terms induced by the mapping from  $\mathbf{q}$  to  $\mathbf{x}$  and then from  $\mathbf{x}$  to  $\mathbf{s}$ , we can try to invert the nonlinear mapping and solve for the Lagrangian displacement and the shift velocity, which are Lagrangian quantities that are not affected by shift terms.

The basic idea of nonlinear reconstruction is to build a bijective mapping between the redshift space coordinates

$\mathbf{s}$  and the potential isobaric coordinates  $\boldsymbol{\xi}$ , where  $\rho(\boldsymbol{\xi})d^3\xi$  is approximately constant. We define a coordinate transformation that is a pure gradient,

$$s^i = \xi^\mu \delta_\mu^i + \frac{\partial \phi^s}{\partial \xi^\nu} \delta^{i\nu}, \quad (8)$$

where  $\phi^s(\boldsymbol{\xi})$  is the *displacement potential*. The potential isobaric gauge is unique as long as we require the coordinate transformation defined above is positive definite, i.e.,  $\det(\partial s^i / \partial \xi^\alpha) > 0$ . It becomes analogous to synchronous gauge and Lagrangian coordinates before shell crossing, but allows a unique mapping even after shell crossing. The unique displacement potential  $\phi^s(\boldsymbol{\xi})$  consistent with the nonlinear density and positive definite coordinate transformation can be solved using the moving mesh approach, which evolves the coordinate system towards a state of constant mass per volume element, i.e.,  $\rho(\boldsymbol{\xi})d^3\xi = \text{constant}$  [74, 75]. See Ref. [30] for the details of this calculation.

The gradient of the reconstructed displacement potential  $\nabla_\xi \phi^s(\boldsymbol{\xi})$  provides an estimate of the redshift space displacement  $\boldsymbol{\psi}^s(\mathbf{q})$ . We define the negative Laplacian of the reconstructed displacement potential the reconstructed density field,

$$\delta_r^s(\boldsymbol{\xi}) = -\nabla_\xi \cdot \nabla_\xi \phi^s(\boldsymbol{\xi}) = -\nabla_\xi^2 \phi^s(\boldsymbol{\xi}). \quad (9)$$

Note that the reconstructed density field is computed on the potential isobaric coordinates instead of the redshift space coordinates.

Because of the limited degrees of freedom, we can only solve the scalar part of the nonlinear mapping, while the real nonlinear displacement also includes the curl part. The gradient and curl parts of the redshift space nonlinear displacement are

$$\boldsymbol{\psi}_E^s(\mathbf{q}) = \boldsymbol{\psi}_E(\mathbf{q}) + \mathbf{v}_E^s(\mathbf{q})/aH, \quad (10)$$

and

$$\boldsymbol{\psi}_B^s(\mathbf{q}) = \boldsymbol{\psi}_B(\mathbf{q}) + \mathbf{v}_B^s(\mathbf{q})/aH, \quad (11)$$

respectively, where

$$\boldsymbol{\psi}_E(\mathbf{k}) = [\mathbf{k} \cdot \boldsymbol{\psi}(\mathbf{k})]\mathbf{k}/k^2, \quad \boldsymbol{\psi}_B(\mathbf{k}) = \boldsymbol{\psi}(\mathbf{k}) - \boldsymbol{\psi}_E(\mathbf{k}), \quad (12)$$

and

$$\mathbf{v}_E^s(\mathbf{k}) = [\mathbf{k} \cdot \mathbf{v}^s(\mathbf{k})]\mathbf{k}/k^2, \quad \mathbf{v}_B^s(\mathbf{k}) = \mathbf{v}^s(\mathbf{k}) - \mathbf{v}_E^s(\mathbf{k}). \quad (13)$$

Since the displacement field is anisotropic in the presence of redshift space distortions, the reconstructed displacement potential is also anisotropic. In addition to the real space  $E$ -mode displacement  $\boldsymbol{\psi}_E(\mathbf{q})$ , the solved displacement also contains the  $E$ -mode shift velocity  $\mathbf{v}_E^s(\mathbf{q})/aH$ . The shift velocity is the most important observable in the RSD measurements since it describes the growth of the displacement, which provides cosmological information about the logarithmic growth rate  $f = d\ln D/d\ln a$ , where  $D$  is the linear growth factor. This is new to all

previous reconstruction algorithms since it is about reconstruction of the redshift space distortions.

Notice that the shift velocity  $\mathbf{v}^s(\mathbf{q})$  is different from  $\mathbf{v}(\mathbf{q})$ , where the former corresponds to the  $z$ -component of the latter. The Lagrangian velocities  $\mathbf{v}_E$  and  $\mathbf{v}_B$  describe the growth of the Lagrangian displacements  $\boldsymbol{\psi}_E$  and  $\boldsymbol{\psi}_B$ , respectively. The statistical properties of the Lagrangian displacements have been studied in Appendix A, while the Lagrangian and shift velocities have been investigated in Appendix B and Appendix C.

### III. PERFORMANCE TESTS

To test the performance of reconstruction with redshift space density field, we use a set of ten  $N$ -body simulations run with the CUBEP<sup>3</sup>M code [76]. These simulations evolve  $1024^3$  dark matter particles in a cubic box of length 600 Mpc/ $h$ . We take the snapshots at redshift  $z = 0$  and compute the density on a  $512^3$  grid. We solve the displacement potential  $\phi^s$  using the multigrid algorithm and then compute the reconstructed density field using Eq. (9) in the potential isobaric gauge [30].

To check convergence of the reconstruction results, we run a high resolution simulation of  $2048^3$  particles with the same box size. We also perform reconstruction with the redshift space density field on a  $512^3$  grid from the high resolution simulation. We find the reconstruction results with different simulation resolutions are almost indistinguishable. Therefore, we use the high resolution simulation for visual comparison due to its better image resolution and use the statistical averaged power spectra from the other ten simulations to reduce statistical errors.

In the following subsections, we assess the performance of reconstruction in redshift space in terms of the density field slice, the correlation with the initial conditions, the correlation with the nonlinear fields, the shape of the power spectrum, and the correlation with the Lagrangian displacements and velocities.

#### A. Nonlinear map with potential isobaric gauge

Figure 1 shows a slice of the nonlinear density through the high resolution simulation, while the corresponding redshift space density field is presented in Fig. 2. We also plot the Eulerian positions of a uniform grid of potential isobaric coordinates on the corresponding density field slices. At small scales the random velocity dispersion elongates cosmic structure along the line of sight, leading to the Finger of God effect. The Finger of God effect smoothes the density fluctuations along the line of sight, blurring the clear small scale structure in real space. This causes the loss of cosmological information about the initial conditions. Thus we expect the performance in redshift space will be degraded compared to the result in real space.



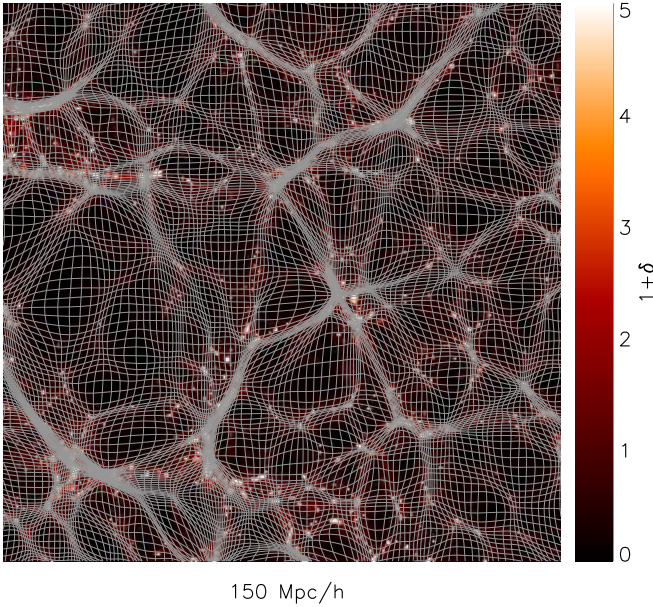


FIG. 1: A slice of the nonlinear density field from the high resolution simulation without the RSD effect. The curvilinear grid shows the Eulerian coordinate of each grid point of the potential isobaric coordinate.

The grid lines become closer in the higher density regions and sparser in the lower density regions. As a result, the mass enclosed in each curvilinear grid cell is approximately constant. Since the structure in redshift space is blurred along the line of sight, the grid lines also move apart in the dense regions compared to the corresponding real space grid line separation. This suppresses the small scale power of the reconstructed density field along the line of sight. As the reconstructed density field is given by

$$\delta_r^s(\xi) = -\nabla_\xi \cdot (s(\xi) - \xi) = 3 - \nabla_\xi \cdot s(\xi), \quad (14)$$

where  $\nabla_\xi \cdot s(\xi) = \partial s_x / \partial \xi_x + \partial s_y / \partial \xi_y + \partial s_z / \partial \xi_z$ , a larger separation  $\Delta s_z$  between neighbouring curvilinear grid lines implies a smaller density. Although the squashing effect on large scales is not obvious in the density field slice presented here, we still expect the reconstructed density field have similar statistical properties as the redshift space nonlinear density field but with much less nonlinearities.

## B. Correlation with the linear initial conditions

To directly measure the linear initial conditions recovered with nonlinear reconstruction, we calculate the cross-correlation coefficient between the density field and the linear initial conditions. The power spectra used for computing the cross correlation coefficients are averaged over ten simulation to reduce the cosmic variance.

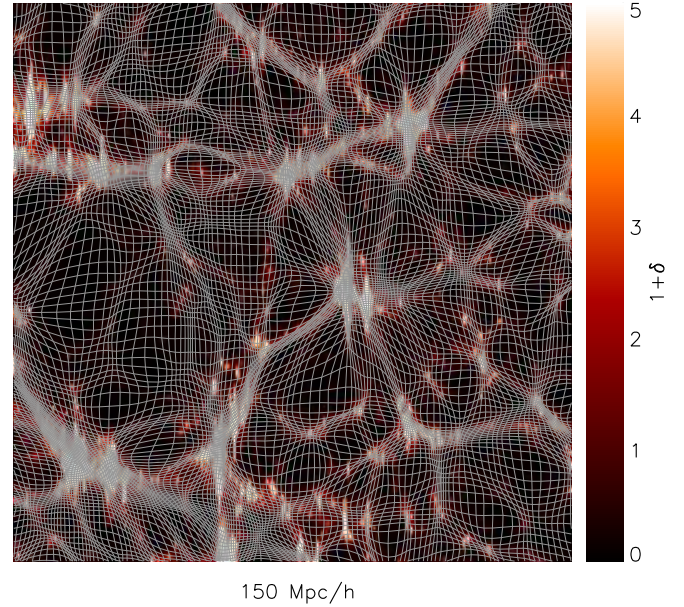


FIG. 2: A slice of the nonlinear density field from the high resolution simulation with the RSD effect. The curvilinear grid shows the Eulerian coordinate of each grid point of the potential isobaric coordinate.

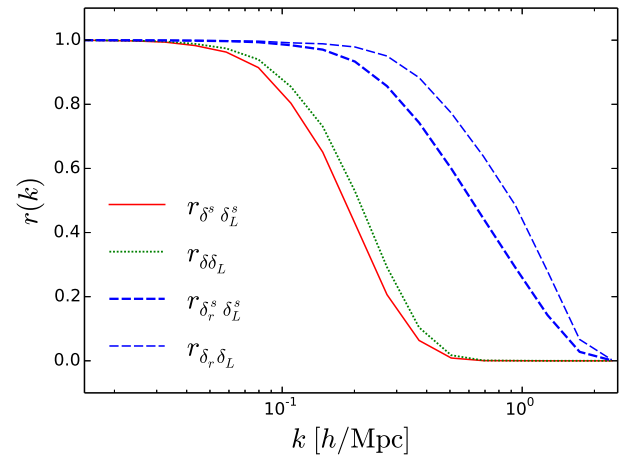


FIG. 3: The cross-correlation coefficients with the initial conditions for the nonlinear density field (solid line), the reconstructed density field (thick-dashed line), and the nonlinear (dotted line) and reconstructed (thin-dashed line) density field without redshift space distortions. The correlation with the linear field for the redshift space field after reconstruction is better than 50% at  $k \lesssim 0.6 h/\text{Mpc}$  and 80% at  $k \lesssim 0.3 h/\text{Mpc}$ .

We correlate the nonlinear density field  $\delta(\mathbf{k})$  and the reconstructed density field  $\delta_r(\mathbf{k})$  without the RSD effect with the linear density  $\delta_L(\mathbf{k})$ . For the redshift space density field  $\delta_L^s(\mathbf{k})$  and the reconstructed density field  $\delta_r^s(\mathbf{k})$ , we instead correlate them with the redshift space displace-

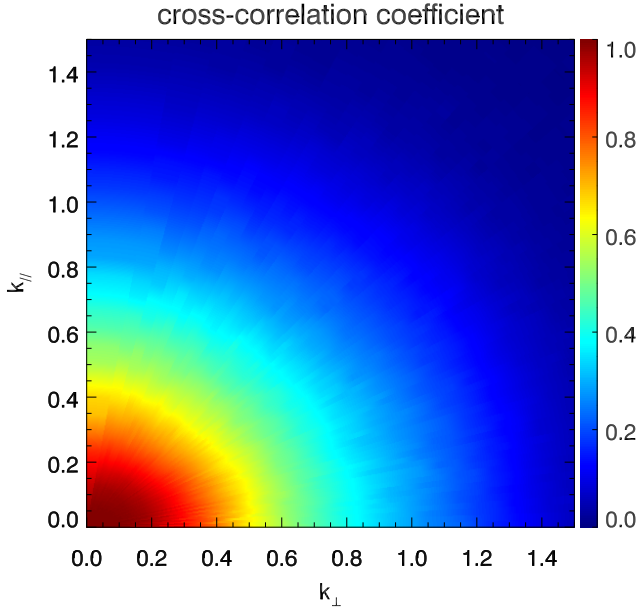


FIG. 4: The two-dimensional cross-correlation coefficient of the reconstructed density field with the redshift space linear density field.

ment divergence,

$$\delta_L^s(\mathbf{k}) = -i\mathbf{k} \cdot \boldsymbol{\psi}_L^s(\mathbf{k}) = (1 + f\mu^2)\delta_L(\mathbf{k}), \quad (15)$$

where  $\mu = k_z/k$ . Figure 3 shows the correlation of the redshift space density field with the linear initial conditions before and after nonlinear reconstruction. We also plot the cross-correlation coefficients for real space density fields for comparison. The correlation with the linear field for the redshift space field after reconstruction is better than 50% at  $k \lesssim 0.6$  h/Mpc and 80% at  $k \lesssim 0.3$  h/Mpc. For the redshift space nonlinear density field, the correlation is better than 50% and 80% only at  $k \lesssim 0.17$  h/Mpc and  $k \lesssim 0.1$  h/Mpc. Nonlinear reconstruction significantly reduces the shift nonlinearities and improves the number of linear modes by a factor  $30 \sim 40$  depending on the scale used for comparison.

We note that RSD significantly degrades the performance of reconstruction. However, the nonlinear density field is less affected by RSD; the correlation with the linear density field is slightly smaller than without RSD. This is because RSD mainly scramble small scale nonlinear structures and these small-scale structures almost have no correlation with the initial conditions. However, we can recover the cosmological information which is still present in these structures. While RSD blur these visible small scale structures, we can no longer reconstruct the cosmological information that is present in the small scale fluctuations of real space density field.

As the solved displacement potential is anisotropic in the presence of RSD, we expect the performance of reconstruction will depend on the direction. Figure 4 shows the two-dimensional correlation coefficient  $r_{\delta_r^s \delta_L^s}(k_\perp, k_\parallel)$

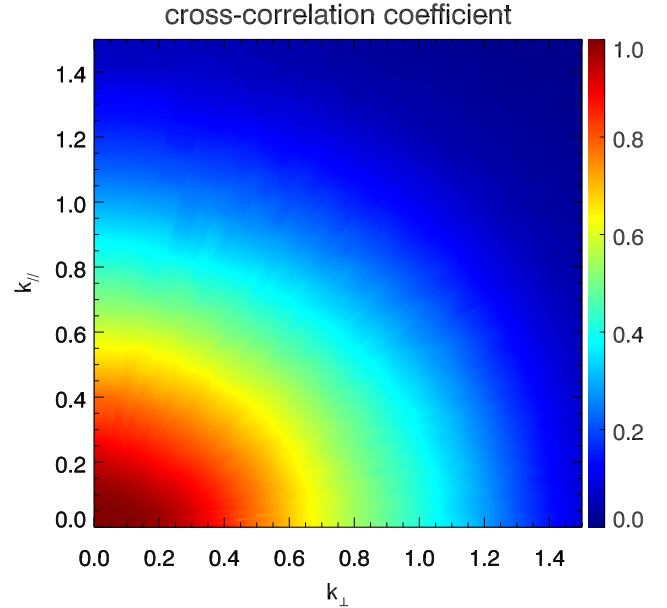


FIG. 5: The two-dimensional cross-correlation coefficient of the reconstructed density field with the redshift space nonlinear field.

between the redshift space reconstructed density field and the linear density field. Note that the correlation with the linear field along the  $k_\parallel$  axis drops faster than along the  $k_\perp$  axis. One reason for this is that the Lagrangian velocity  $\mathbf{v}(\mathbf{q})$  is much more nonlinear than the Lagrangian displacement  $\boldsymbol{\psi}(\mathbf{q})$ . The Lagrangian displacement is an integral with respect to time of the Lagrangian velocity,

$$\boldsymbol{\psi}(\mathbf{q}, t) = \int_0^t \mathbf{v}(\mathbf{q}, t') dt'. \quad (16)$$

At higher redshifts the velocity field is more linear than the velocity field at later time. The Lagrangian displacement is a linear combination of velocities at different redshifts with different amount of nonlinearities, while the present day Lagrangian velocity corresponds to the most nonlinear contribution to the displacement. More details about this discussion is presented in Appendix A and B. Since the combination of  $\nabla \cdot \boldsymbol{\psi}$  and  $\nabla \cdot \mathbf{v}^s/aH$  is more nonlinear than  $\nabla \cdot \boldsymbol{\psi}$  and most power of  $\nabla \cdot \mathbf{v}^s$  is concentrated in  $\mu$  bins close to  $\mu = 1$ , the correlation with the linear density field drops faster in these  $\mu$  bins.

### C. Correlation with the nonlinear field

The correlation with the linear field quantifies the linear signal recovered with nonlinear reconstruction, which is relevant for improving the measurements of BAO scale in redshift surveys. To answer how well we reconstruct the nonlinear displacement, we need to correlate the reconstructed density field with the redshift space nonlin-

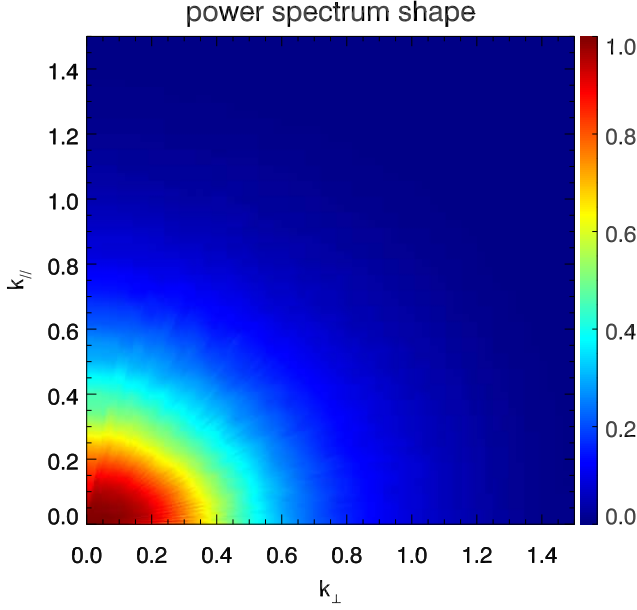


FIG. 6: The shape of the power spectrum for the filtered reconstructed density field normalized to the linear field.

ear displacement divergence,

$$\delta_E^s(\mathbf{k}) = -i\mathbf{k} \cdot \boldsymbol{\psi}^s(\mathbf{k}) = -i\mathbf{k} \cdot (\boldsymbol{\psi} + \mathbf{v}^s/aH)(\mathbf{k}). \quad (17)$$

Figure 5 demonstrates the two-dimensional correlation coefficient  $r_{\delta_r^s \delta_E^s}(k_\perp, k_\parallel)$  between the redshift space reconstructed density field and the redshift space nonlinear  $E$ -mode displacement. The correlation with this nonlinear field is much better than that with the linear field as expected. The redshift space nonlinear displacement divergence represents the best possible description of the reconstructed density field.

We find that the cross correlation coefficient with the nonlinear field shows similar anisotropic feature as the cross correlation coefficient with the linear field. This is because our reconstruction method can not recover the  $B$ -mode displacement  $\boldsymbol{\psi}_B + \mathbf{v}_B^s/aH$ . The  $B$ -mode real space nonlinear displacement  $\boldsymbol{\psi}_B$  is negligible on scales we care. The  $x$ - and  $y$ -component of the  $B$ -mode shift velocity  $\mathbf{v}_B^s$  are also negligible. However, the  $z$ -component of the  $B$ -mode shift velocity  $\mathbf{v}_B^s$  contributes 8/35 of the total power of  $\mathbf{v}^s$  on large scales and even more on smaller scales (see Appendix C). Therefore, the correlation with the nonlinear field is slightly lower in the  $\mu$  bins close to  $\mu = 1$  as shown in Fig. 5. The effect may also cause part of the degradation of the correlation coefficient with the linear field along the line of sight.

#### D. Reconstruction with transfer functions

To recover the broadband power of the reconstructed density field, we can use the transfer functions [31]. We

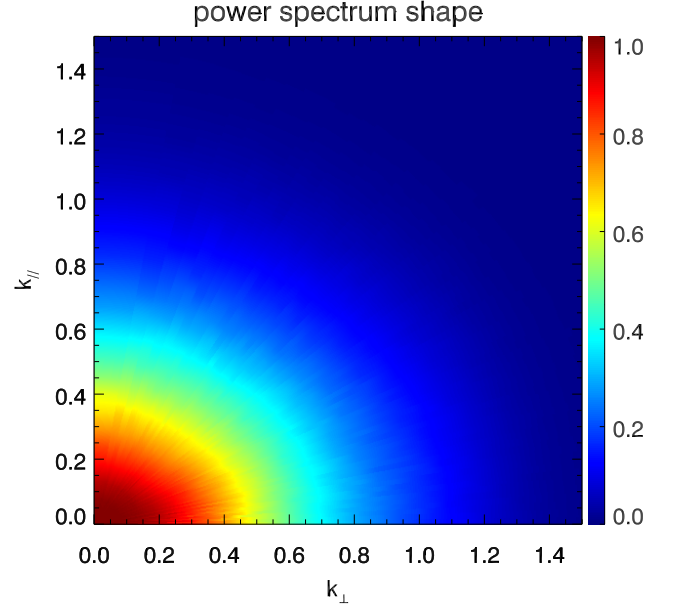


FIG. 7: The shape of the power spectrum for the filtered reconstructed density field normalized to the nonlinear field.

write the filtered reconstructed field as

$$\tilde{\delta}_r^s(\mathbf{k}) = t(\mathbf{k})\delta_r^s(\mathbf{k}), \quad (18)$$

where  $t(\mathbf{k})$  is the transfer function to be specified. Note that the transfer function is anisotropic in the presence of RSD. We can choose the transfer function by minimizing the difference between the reconstructed density field and the linear density field,

$$\langle (t(\mathbf{k})\delta_r^s(\mathbf{k}) - \delta_L^s(\mathbf{k}))^2 \rangle, \quad (19)$$

and have

$$t_L(\mathbf{k}) = \frac{P_{\delta_r^s \delta_L^s}(\mathbf{k})}{P_{\delta_r^s}(\mathbf{k})}, \quad (20)$$

where the subscript  $L$  denotes the transfer function is to recover the broadband power of the reconstructed density field with respect to the linear field. We can also choose the transfer function by minimizing the difference with the nonlinear field,

$$\langle (t(\mathbf{k})\delta_r^s(\mathbf{k}) - \delta_E^s(\mathbf{k}))^2 \rangle, \quad (21)$$

and have

$$t_E(\mathbf{k}) = \frac{P_{\delta_r^s \delta_E^s}(\mathbf{k})}{P_{\delta_r^s}(\mathbf{k})}, \quad (22)$$

where the subscript  $E$  denotes the transfer function recovers the broadband power spectrum with respect to the nonlinear field. Figure 6 shows the power spectrum of the linear transfer function filtered reconstructed field divided by the linear power spectrum. The ratio between



the two power spectra is larger than 0.5 at  $k \lesssim 0.5 h/\text{Mpc}$  for  $\mu$  bins around  $\mu = 0$  and  $k \lesssim 0.4 h/\text{Mpc}$  for  $\mu$  bins around  $\mu = 1$ . Again, we note a deficit in power along the line of sight direction. This is because the shift velocity divergence  $\nabla \cdot \mathbf{v}^s$  is more nonlinear than displacement and the reconstruction misses the  $B$ -mode shift velocity  $\mathbf{v}_B^s$  as we have discussed above. Figure 7 shows the power spectrum of the nonlinear transfer function filtered reconstructed field divided by the nonlinear power spectrum. The ratio between the two power spectra is larger than 0.5 at  $k \lesssim 0.6 h/\text{Mpc}$  for  $\mu$  bins around  $\mu \sim 0$  and  $k \lesssim 0.5 h/\text{Mpc}$  for  $\mu$  bins around  $\mu \sim 1$ . Here the missed power along the line of sight is solely due to the unrecovered  $B$ -mode shift velocity  $\mathbf{v}_B^s$ .

The transfer function introduced above is similar to the Wiener filter. The reconstructed density field can be written as

$$\delta_r^s(\mathbf{k}) = C_L(\mathbf{k})\delta_L^s(\mathbf{k}) + N_L(\mathbf{k}), \quad (23)$$

where  $C_L(\mathbf{k}) = P_{\delta_r^s \delta_L^s}(\mathbf{k})/P_{\delta_L^s}(\mathbf{k})$  is the propagator and  $N_L(\mathbf{k})$  is the noise with respect to the linear field. The Wiener filter for the linear signal in reconstructed density field is

$$W_L(\mathbf{k}) = \frac{C_L^2(\mathbf{k})P_{\delta_L^s}(\mathbf{k})}{C_L^2 P_{\delta_L^s}(\mathbf{k}) + P_{N_L}(\mathbf{k})}, \quad (24)$$

which differs from the linear transfer function  $t_L(\mathbf{k})$  by a propagator  $C_L(\mathbf{k})$ ,

$$t_L(\mathbf{k}) = \frac{W_L(\mathbf{k})}{C_L(\mathbf{k})}. \quad (25)$$

The transfer function also deconvolves the propagator from the density field in addition to apply the Wiener filter to recover the power spectrum shape. Figure 8 shows the two-dimension propagator with respect to the linear field. The propagator shows additional suppression along the  $z$ -direction. This is caused by the nonlinearity of the shift velocity  $\mathbf{v}^s$  and the Finger of God effect. Since the combination of displacement divergence  $\nabla \cdot \boldsymbol{\psi}$  and velocity divergence  $\nabla \cdot \mathbf{v}^s$  is more nonlinear than  $\nabla \cdot \boldsymbol{\psi}$  itself and most power of the shift velocity divergence  $\nabla \cdot \mathbf{v}^s$  is concentrated around  $\mu \sim 1$ , the  $\mu$  bins close to  $\mu = 1$  contain less information about the linear initial conditions. The small scale random velocity dispersion suppresses the density fluctuations along the line of sight, leading to larger separations along the  $z$ -axis between neighbouring grid lines. The nonlinear effects due to random motion on small scales not only cause the suppression of power along the line of sight, but also scramble the information about linear initial conditions along this direction.

We can also write the reconstructed density field as

$$\delta_r^s(\mathbf{k}) = C_E(\mathbf{k})\delta_E^s(\mathbf{k}) + N_E(\mathbf{k}), \quad (26)$$

where  $C_E(\mathbf{k}) = P_{\delta_r^s \delta_E^s}(\mathbf{k})/P_{\delta_E^s}(\mathbf{k})$  and  $N_E(\mathbf{k})$  are the propagator and noise with respect to the nonlinear field.

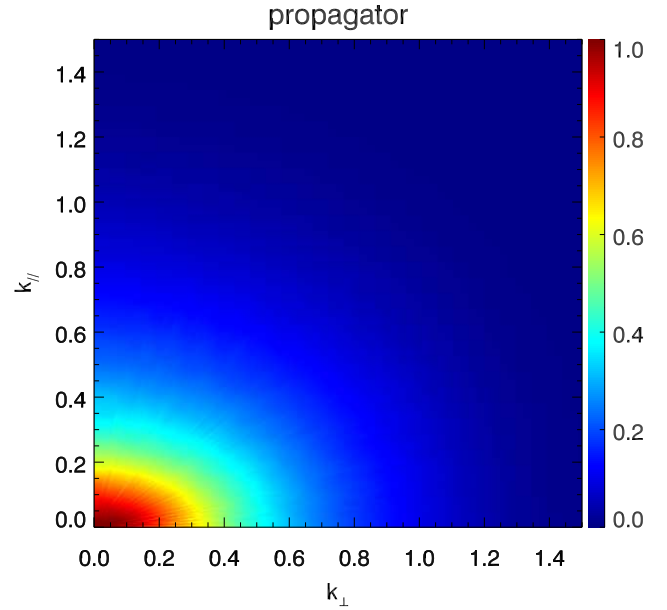


FIG. 8: The propagator  $C_L(\mathbf{k})$  for the linear field. The additional damping along the line of sight is caused by the nonlinearity of the shift velocity and the Finger of God effect.

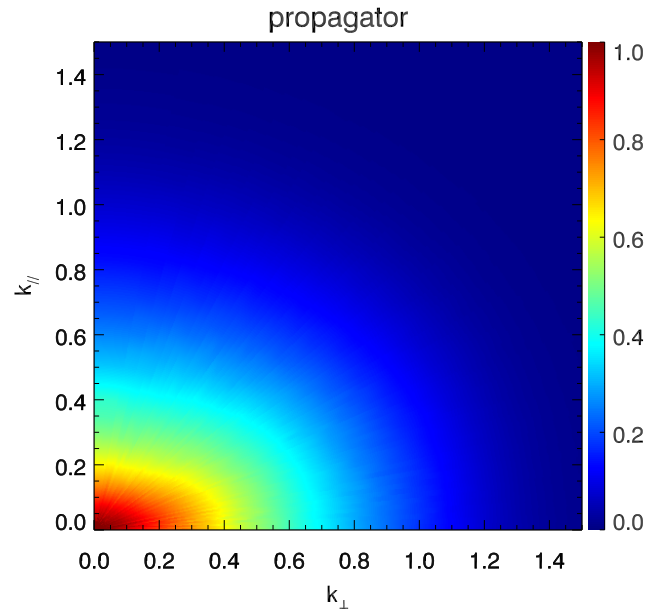


FIG. 9: The propagator  $C_E(\mathbf{k})$  for the nonlinear field. The additional damping along the line of sight is caused by the Finger of God effect only as we are correlating with the nonlinear field.

The similar relation  $t_E(\mathbf{k}) = W_E(\mathbf{k})/C_E(\mathbf{k})$  also holds for the Wiener filter for the nonlinear signal

$$W_E(\mathbf{k}) = \frac{C_E^2(\mathbf{k})P_{\delta_E^s}(\mathbf{k})}{C_E^2 P_{\delta_E^s}(\mathbf{k}) + P_{N_E}(\mathbf{k})}. \quad (27)$$

Figure 9 shows the two-dimension propagator with re-

spect to the nonlinear field. Here the additional damping along the line of sight is only due to the Finger of God effect as we are correlating with the nonlinear field.

To extract distinct features like BAO wiggles from the power spectrum of the reconstructed density field, we can directly apply the transfer function calibrated from simulations or mocks. In the BAO-only analysis, the transfer function is only used to match the power spectrum amplitude with the linear density field and itself does not contain any information about the BAO wiggles if calibrated to simulations with no-wiggle initial linear power spectra [31]. However, if we want to extract additional cosmological information about RSD from the density field, we need to model the broadband power spectrum of the reconstructed density field. The linear power spectrum  $P_{\delta_L^s}(\mathbf{k})$  or the nonlinear power  $P_{\delta_E^s}(\mathbf{k})$  does not suffice to model the reconstructed density field, as they only describe the coherent part of the reconstructed density field. We need to consider the small scale random velocity dispersion and the errors of reconstruction, i.e., the propagator  $C(\mathbf{k})$  and the noise term  $P_N(\mathbf{k})$ . For the linear field, we also need to consider nonlinearities of the coherent part of the reconstructed field for modeling the mode-coupling term  $P_N(\mathbf{k})$ . Probably these can be studied using higher-order Lagrangian perturbation theories or the effective field theories of large-scale structure [72]. Similar works for the standard reconstruction method have been conducted [16, 25, 26, 29, 77, 78]. Or we can directly calibrate them against simulations or mock catalogs but the computation cost would be expensive. We leave this for future work.

### E. Correlation with displacements and velocities

As we have reversed the nonlinear mapping from the initial Lagrangian positions to the final redshift space positions up to shell crossing during nonlinear reconstruction, the gradient of the solved displacement potential

$$\nabla\phi^s = (\phi_{,x}^s, \phi_{,y}^s, \phi_{,z}^s) \quad (28)$$

gives an estimate of the nonlinear displacement

$$\psi_E^s = (\psi_{Ex} + \frac{v_{Ex}^s}{aH}, \psi_{Ey} + \frac{v_{Ey}^s}{aH}, \psi_{Ez} + \frac{v_{Ez}^s}{aH}). \quad (29)$$

To examine the performance of reconstruction along different directions, we compute the correlation coefficient for each component separately. Figure 10 shows the cross correlation coefficients. For comparison we also plot the cross correlation coefficient between the reconstructed density field and the nonlinear displacement divergence. The cross correlation for the  $x$ - and  $y$ -component of the displacement is better than for the  $z$ -component as expected, since the missed  $z$ -component  $v_{Bz}^s$  of the  $B$ -mode shift velocity is much larger than  $v_{Bx}^s$  and  $v_{By}^s$ . Though we can not reconstruct these  $B$ -mode shift velocities, they still change the observed mass distribution used for

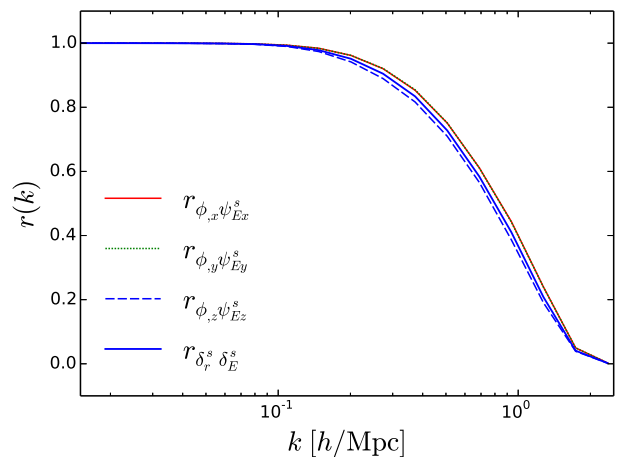


FIG. 10: The cross-correlation coefficients of the reconstructed displacement components with the nonlinear displacement components (solid, dotted, dashed lines are for  $x, y, z$ -components respectively) and the reconstructed density field with the nonlinear displacement divergence.

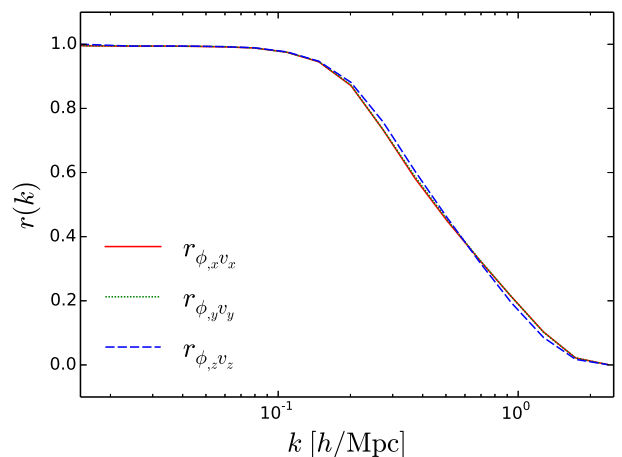


FIG. 11: The cross correlation coefficient of the reconstructed displacement with the the Lagrangian velocity.

nonlinear reconstruction; as a result, they will include noises for the reconstructed displacements respectively, reducing the correlations with the real  $E$ -mode displacements.

In linear theory, the Lagrangian displacement only differs from the Lagrangian velocity by a factor  $afH$ . Their relation becomes complicated in higher order Lagrangian perturbation theory. Nevertheless, it is still interesting to study the cross correlation of the reconstructed displacement with the real Lagrangian velocity. Figure 11 shows the cross correlation coefficients. The cross correlation coefficient for each component is larger than 0.5 at  $k \lesssim 0.43 \text{ h/Mpc}$ , though the behavior of  $z$ -direction is a little different from the other two directions. This scale is much smaller than the scale where the nonlinear density field correlates with the linear density field



( $k \lesssim 0.17 h/\text{Mpc}$ ). Therefore, we expect the new velocity reconstruction scheme based on the relation between the reconstructed displacement and the Lagrangian velocity would be better than the method based on the linearized continuity equation.

#### IV. APPLICATIONS

The prime application of nonlinear reconstruction is to reverse the large scale bulk flows and improve the precision measurement of the BAO scale. The density field reconstructed with the dark matter density in real space has a comparable fidelity as the linear density field for measuring the BAO scale, since the BAO wiggles are washed out at  $k > 0.6 h/\text{Mpc}$  in both fields. The new reconstruction methods can restore the BAO signal of the linear density perfectly in the idealized situation [31, 79]. The shot noise will limit the performance of reconstruction when we apply nonlinear reconstruction to real space halo fields [80]. The new reconstruction method requires a higher number density  $\bar{n} \geq 10^{-3} (h/\text{Mpc})^3$  to have a better performance than the standard reconstruction method [80]. In this paper we show RSD will further degrade the performance due to the nonlinearity of the shift velocity and the small scale random velocity dispersion. Nevertheless, we still expect the new method to substantially improve the precision of the measured BAO scale of dense low redshift large scale structure surveys, including SDSS main sample, DESI BGS, and 21 cm intensity mapping initiatives, while the improvement over standard reconstruction for less dense high redshift samples like BOSS LOWZ and CMASS will be moderate.

To apply nonlinear reconstruction to observations, we need a continuous galaxy density field without any hole or masked region. A practical issue related to this is the boundary of a survey region, which is not the periodic simulation box. This problem can be solved by embedding the observed galaxy density field into a uniform background. We have demonstrated a simple test about this solution in Appendix D. We have applied reconstruction to halo density fields in redshift space. The two-dimensional correlation coefficients of the reconstructed density field with the linear initial conditions are presented in Appendix E. The practical application of nonlinear reconstruction to SDSS galaxy samples will be studied in a future paper [81].

The most innovative result of this paper is that the nonlinear reconstruction method can improve not only the BAO measurement but also the RSD measurement. The nonlinear mapping solved from the anisotropic redshift space density field includes the  $E$ -mode shift velocity  $\mathbf{v}_E^s$ , which contains cosmological information about the growth of structure. The reconstructed density field shares many similar features as the redshift space density field like the large scale squashing effect and small scale Finger of God effect, but much more linear than the original redshift space density field. A large amount

of shift nonlinearities has been removed in the nonlinear reconstruction. The number of linear modes can be used to measure the growth rate is increased by about  $30 \sim 40$  times after reconstruction. Therefore, we expect this will significantly reduce uncertainties of the measured structure growth rate since the constraining power of galaxy surveys scales steeply with the number of modes included in the RSD analysis. In contrast to the BAO reconstruction, where nonlinear reconstruction still needs a higher number density to gain better performance than standard reconstruction, the RSD nonlinear reconstruction can always obtain better performance than without reconstruction, no matter the number density of galaxy samples. This has significant implications for constraining modified gravity theories using the measured linear structure growth rate [82]. However, we still expect substantial improvements for the dense low-redshift surveys as discussed above while the performance for less dense samples will be limited by the shot noise.

The current BAO reconstruction involves displacing particles according to the large-scale linear displacement field computed from the observed galaxy density field with some certain model assumptions like the smoothing scale, galaxy bias and growth rate etc [14]. The results depend on the assumed fiducial model and have to be tested against different parameter choices, which are computationally expensive [18]. The new reconstruction method directly solves the displacement potential from the observed density field, which is a purely mathematical problem without any cosmological dynamics involved. The parameters like the halo bias and structure growth rate do affect the reconstructed density field, but they are to be jointly fitted when we model the power spectrum after reconstruction, while there is no cosmological parameters involved in the reconstruction process itself. This saves a large amount of computation costs when applying reconstruction to galaxy surveys to improve BAO and RSD measurements.

The fact that the reconstructed displacements correlate with the Lagrangian velocities to nonlinear scales in Lagrangian space implies we can construct new peculiar velocity reconstruction schemes based on this relation. The current peculiar velocity reconstruction methods are often based on the linearized continuity equation like the one used to measure the kSZ effect through cross correlation [83]. The linearized continuity equation is

$$\dot{\delta} + \nabla \cdot \mathbf{v} = 0. \quad (30)$$

Assuming the velocity is curl-free, the velocity can be inferred from the density through

$$\mathbf{v}(\mathbf{k}) = afH \frac{i\mathbf{k}}{k^2} \delta(\mathbf{k}) W_R(k), \quad (31)$$

where  $W_R(k)$  is a smoothing function with the characteristic scale  $R$ . The smoothing scale should be chosen close to the scale where linear theory breaks down [16, 29, 84]. To make the above linear approximation valid, the nonlinear density field usually needs to be smoothed on scales

$\sim 10 \text{ Mpc}/h$  to remove small scale nonlinearities. This is because all quantities defined in Eulerian space like  $\delta(\mathbf{x})$  and  $\mathbf{v}(\mathbf{x})$  involve very strong shift nonlinearities, which have to be reduced in order to apply the linear relation. However, a lot of cosmological information about small-scale velocities is lost in this process. The quantities defined in Lagrangian space like Lagrangian displacement  $\psi(\mathbf{q})$  and Lagrangian velocity  $\mathbf{v}(\mathbf{q})$  are free of any nonlinear shift terms [72]. This motivates us to develop new velocity reconstruction methods using the relation between Lagrangian quantities instead of Eulerian quantities.

In the Lagrangian perturbation theory, the Lagrangian displacement is solved as

$$\psi(\mathbf{q}, t) = \psi^{(1)}(\mathbf{q}, t) + \psi^{(2)}(\mathbf{q}, t) + \psi^{(3)}(\mathbf{q}, t) + \dots, \quad (32)$$

where  $\psi^{(n)} \propto (\psi^{(1)})^n$  is the  $n$ th-order solution. The Lagrangian velocity of mass element with initial coordinate  $\mathbf{q}$  is the time derivative of the Lagrangian displacement

$$\mathbf{v}(\mathbf{q}, t) = a\dot{\mathbf{x}}(\mathbf{q}, t) = a\dot{\psi}(\mathbf{q}, t). \quad (33)$$

The time dependence of the  $n$ th-order solution  $\psi^{(n)}$  can be well approximated as  $\psi^{(n)} \propto D^n$ . Therefore, the Lagrangian velocity can be expressed as

$$\mathbf{v}(\mathbf{q}, t) = afH\psi^{(1)}(\mathbf{q}, t) + 2afH\psi^{(2)}(\mathbf{q}, t) + \dots. \quad (34)$$

The most direct and simplest idea is to filter the reconstructed displacement field with the linear transfer function to minimize the difference with the linear displacement field. Then compute the Lagrangian velocity using the linear relation  $\mathbf{v}^{(1)}(\mathbf{q}) = afH\psi^{(1)}(\mathbf{q})$  and next generate the Eulerian space velocity field using a velocity assignment method [85, 86]. This  $\mathcal{O}(1)$  velocity reconstruction is the Lagrangian space version of the current velocity reconstruction in Eulerian space. However, since much less shift nonlinearities are involved in Lagrangian space and the displacement and velocity correlate to a much smaller scale in Lagrangian space than the density and velocity in Eulerian space, we thus expect the new method will have a better performance than the previous methods operated in Eulerian space. We can also compute the second-order solution  $2afH\psi^{(2)}$  using the linear transfer function filtered displacement since the displacement is much more linear than the Lagrangian velocity. Then we can use the second order transfer functions to optimally combine  $afH\psi^{(1)}$  and  $2afH\psi^{(2)}$  to obtain the reconstructed Lagrangian velocity [31]. The  $\mathcal{O}(2)$  velocity reconstruction scheme is expected to better recover small scale velocities. One more possibility is to directly correlate the reconstructed displacement with the nonlinear Lagrangian velocity and solve the optimal transfer function. The detailed tests of the new velocity reconstruction schemes and comparisons against the current velocity reconstruction method will be investigated in a future paper [87].

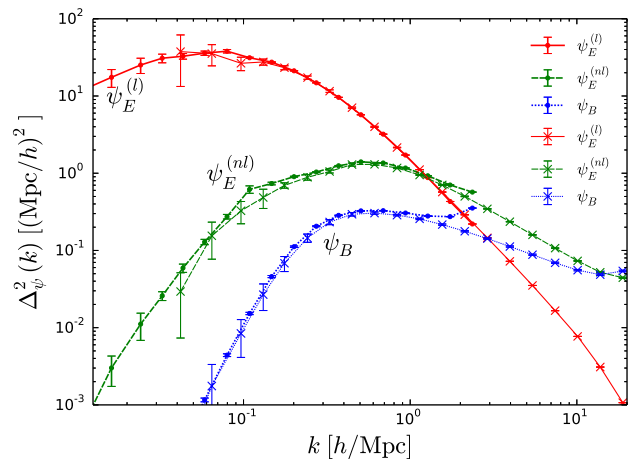


FIG. 12: The power spectra for the decomposed Lagrangian displacements. The large scale nonlinear displacement field is well described by the Zeldovich approximation for  $k \lesssim 1 \text{ h/Mpc}$ . The nonlinear  $E$ -mode displacement becomes important at  $k \gtrsim 1 \text{ h/Mpc}$ . The  $B$ -mode displacement only matters on very small scales for  $k \gtrsim 10 \text{ h/Mpc}$ .

## V. CONCLUSIONS

Reconstruction errors/noises...

## VI. ACKNOWLEDGEMENT

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## Appendix A: Lagrangian displacement field

Any vector field can be separate into a gradient part and a curl part. Hence we can decompose the real space nonlinear displacement as

$$\psi(\mathbf{q}) = \psi_E(\mathbf{q}) + \psi_B(\mathbf{q}), \quad (\text{A1})$$

where  $\psi_E = (\mathbf{k} \cdot \psi)\mathbf{k}/k^2$  and  $\psi_B = \psi - \psi_E$ . The  $E$ -mode displacement can be completely described by its

divergence  $\delta_E = -\nabla \cdot \psi_E$ . To see how the  $E$ -mode displacement correlates with the initial linear displacement, we carry out a further decomposition,

$$\psi_E(\mathbf{q}) = \psi_E^{(l)}(\mathbf{q}) + \psi_E^{(nl)}(\mathbf{q}), \quad (\text{A2})$$

where

$$\psi_E^{(l)}(\mathbf{k}) = W(k)\psi_L(\mathbf{k}), \quad (\text{A3})$$

and  $W(k) = P_{\delta_E \delta_L}(k)/P_{\delta_L}(k)$ . Here, the superscript “ $l$ ” denotes the part completely correlated with the linear displacement and “ $nl$ ” denotes the part generated in nonlinear evolution. Then the nonlinear displacement is composed by three components

$$\psi(\mathbf{q}) = \psi_E^{(l)}(\mathbf{q}) + \psi_E^{(nl)}(\mathbf{q}) + \psi_B(\mathbf{q}). \quad (\text{A4})$$

To measure the power spectra of these three components, we use two sets of simulation. The large box simulation involves  $1024^3$  dark matter particles in a cubic box of length  $600 h/\text{Mpc}$  and the small box simulation involves  $1024^3$  dark matter particles in a cubic box of length  $150 h/\text{Mpc}$ . Each set of simulation have ten independent realizations with independent random initial conditions. The power spectrum of displacement is defined as

$$P_\psi(k) = \sum_i P_{\psi_i}(k), \quad (\text{A5})$$

where  $i$  denotes  $x$ ,  $y$ , or  $z$  components. Figure 12 shows the power spectra measured from the two sets of simulations. The displacement power spectra are averaged over ten realizations. The nonlinear displacement field on large scales is well described by the Zeldovich displacement for  $k \lesssim 1 h/\text{Mpc}$ . The nonlinear  $E$ -mode displacement dominates over the linear  $E$ -mode displacement at  $k = 1 h/\text{Mpc}$ . The  $B$ -mode displacement is negligible on scales smaller than  $k = 10 h/\text{Mpc}$ .

## Appendix B: Lagrangian velocity field

The Lagrangian velocity describes the growth of the Lagrangian displacement,

$$\mathbf{v}(\mathbf{q}) = \dot{\psi}(\mathbf{q}), \quad (\text{B1})$$

where the dot denotes partial derivative with respect to time. Since both fields are defined in Lagrangian space, we can decompose the Lagrangian velocity similarly,

$$\mathbf{v}(\mathbf{q}) = \mathbf{v}_E(\mathbf{q}) + \mathbf{v}_B(\mathbf{q}), \quad (\text{B2})$$

where  $\mathbf{v}_E = \nabla(\nabla \cdot \mathbf{v})$  and  $\mathbf{v}_B = \mathbf{v} - \mathbf{v}_E$ . The  $E$ -mode displacement can be decomposed into

$$\mathbf{v}_E(\mathbf{q}) = \mathbf{v}_E^{(l)}(\mathbf{q}) + \mathbf{v}_E^{(nl)}(\mathbf{q}) \quad (\text{B3})$$

where

$$\mathbf{v}_E^{(l)}(\mathbf{k}) = W_v(k)\mathbf{v}_L(\mathbf{k}). \quad (\text{B4})$$

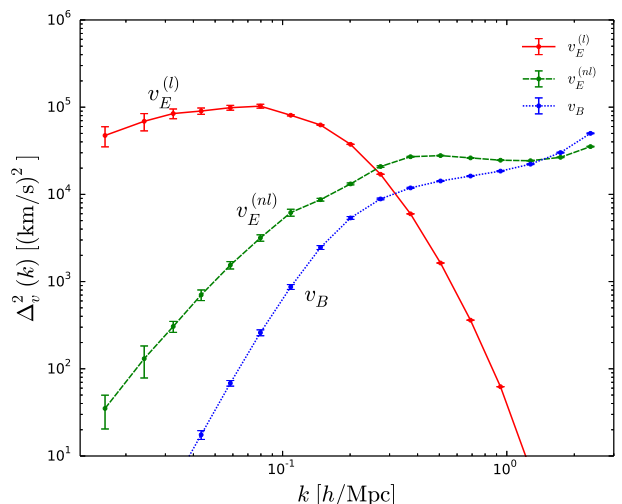


FIG. 13: The power spectra for the decomposed Lagrangian velocity. The linear part of the velocity is dominant at  $k \lesssim 0.25 h/\text{Mpc}$ . The nonlinear contribution becomes important at  $k = 0.25 h/\text{Mpc}$ . The curl part is already non-negligible on scales where the displacement can still be described by the gradient.

Here,  $\mathbf{v}_L(\mathbf{k})$  is derived from the linear density field,

$$\mathbf{v}_L(\mathbf{k}) = afH \frac{i\mathbf{k}\delta_L(\mathbf{k})}{k^2}, \quad (\text{B5})$$

and the velocity window function is

$$W_v(\mathbf{k}) = \frac{P_{\theta_E \delta_L}(k)}{P_{\delta_L}(k)}, \quad (\text{B6})$$

where  $\theta_E = -\nabla \cdot \mathbf{v}/(afH)$ . Now we can write the Lagrangian velocity as

$$\mathbf{v}(\mathbf{q}) = \mathbf{v}_E^{(l)}(\mathbf{q}) + \mathbf{v}_E^{(nl)}(\mathbf{q}) + \mathbf{v}_B(\mathbf{q}), \quad (\text{B7})$$

where different velocities describe the growth of the corresponding Lagrangian displacements. The velocity power spectrum is defined as

$$P_v(k) = \sum_i P_{v_i}(k), \quad (\text{B8})$$

In Fig. 13, we show the velocity power spectra for different velocities. We only present the results from the large box simulations mentioned above since the Lagrangian velocity is much more nonlinear than the displacement. The linear theory can only describe the Lagrangian velocity up to  $k \lesssim 0.25 h/\text{Mpc}$ . The nonlinearities dominate on scales where the displacement can still be described by linear theory. The  $B$ -mode Lagrangian velocity is already non-negligible on scales  $k \gtrsim 1.4 h/\text{Mpc}$ . This is simply because that the Lagrangian velocity corresponds to the most nonlinear contribution to the Lagrangian displacement as we discussed above.

The fact that the Lagrangian velocity is more nonlinear than the displacement implies that it is more important to study the nonlinearity of the velocity rather

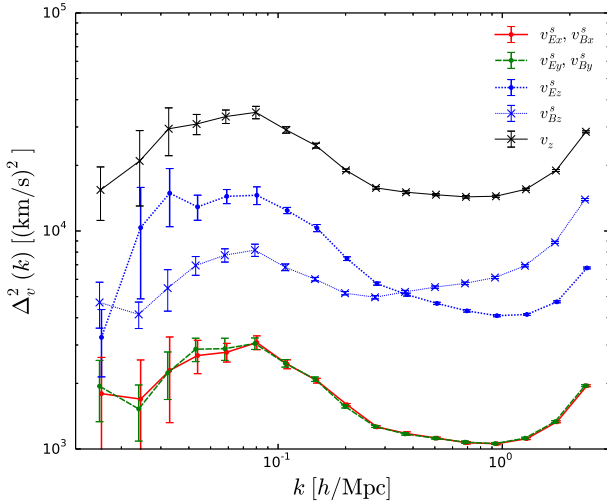


FIG. 14: The power spectra for the different shift velocity components. The power spectrum of  $v_{Bz}^s$  matches the linear theory prediction at  $k \lesssim 0.1 h/\text{Mpc}$ , while the other power spectra can be well described by linear perturbation theory at  $k \lesssim 0.15 h/\text{Mpc}$ . The power of  $v_{Bz}^s$  becomes larger on smaller scales and dominates over  $v_{Ez}^s$  at  $k = 0.36 h/\text{Mpc}$ .

than the displacement to model the reconstructed density field. This also implies we can use the much more linear displacement to calculate the higher order terms of the velocity field using perturbation theories, in order to construct  $\mathcal{O}(2)$  peculiar velocity reconstruction schemes.

### Appendix C: Shift velocity field

The shift velocity corresponds to the  $z$ -component of the Lagrangian velocity,

$$\mathbf{v}^s(\mathbf{q}) = v_z(\mathbf{q})\hat{z}, \quad (\text{C1})$$

which describes the shift of a particle from the real space position to the redshift space position. We can write the shift velocity as

$$\mathbf{v}^s(\mathbf{q}) = \mathbf{v}_E^s(\mathbf{q}) + \mathbf{v}_B^s(\mathbf{q}), \quad (\text{C2})$$

where

$$\mathbf{v}_E^s(\mathbf{k}) = \left( \frac{k_x k_z v_z(\mathbf{k})}{k^2}, \frac{k_y k_z v_z(\mathbf{k})}{k^2}, \frac{k_z k_z v_z(\mathbf{k})}{k^2} \right) v_z(\mathbf{k}) \quad (\text{C3})$$

and

$$\mathbf{v}_B^s(\mathbf{k}) = \left( -\frac{k_x k_z v_z(\mathbf{k})}{k^2}, -\frac{k_y k_z v_z(\mathbf{k})}{k^2}, \frac{(k^2 - k_z^2) v_z(\mathbf{k})}{k^2} \right) \quad (\text{C4})$$

In linear theory, we can compute the power spectra of these velocities from the linear initial density. From the linear continuity equation, we have

$$v_z(\mathbf{k}) = \frac{ik_z}{k^2} afH\delta_L(\mathbf{k}). \quad (\text{C5})$$

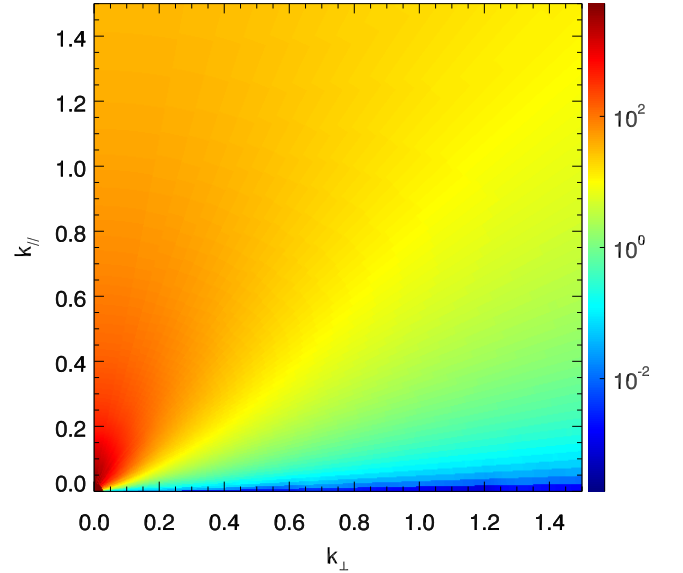


FIG. 15: The two-dimensional power spectrum of the shift velocity divergence. Most power is concentrated in the  $\mu$  bins close to  $\mu \sim 1$  as expected. The values of power spectrum are shown in the logarithmic scale.

Then the angular averaged power spectra of the  $v_z(\mathbf{k})$  is

$$P_{v_z}(k) = \frac{1}{3}(afH)^2 P_{\delta_L}(k)/k^2. \quad (\text{C6})$$

In linear theory, the power spectra of different velocity components are

$$P_{v_{Ex}^s}(k) = P_{v_{Ey}^s}(k) = \frac{3}{35} P_{v_z}(k), \quad (\text{C7})$$

$$P_{v_{Ez}^s}(k) = \frac{15}{35} P_{v_z}(k), \quad (\text{C8})$$

$$P_{v_{Bx}^s}(k) = P_{v_{By}^s}(k) = \frac{3}{35} P_{v_z}(k), \quad (\text{C9})$$

and

$$P_{v_{Bz}^s}(k) = \frac{8}{35} P_{v_z}(k). \quad (\text{C10})$$

Figure 14 shows the velocity power spectra measured from the large box simulations. The effects of the  $x$ - and  $y$ -components of the  $E$ - and  $B$ -mode shift velocities are much smaller than the  $z$ -components. The power spectrum of  $v_{Bz}^s$  matches the linear theory prediction at  $k \lesssim 0.1 h/\text{Mpc}$ , while the other power spectra can be well described by linear theory at  $k \lesssim 0.15 h/\text{Mpc}$ . The power of  $v_{Bz}^s$  becomes larger on smaller scales and dominates over  $v_{Ez}^s$  at  $k = 0.36 h/\text{Mpc}$ . This behavior is not predicted within the linear perturbation theory. The strong nonlinearities in the shift velocity field cause the



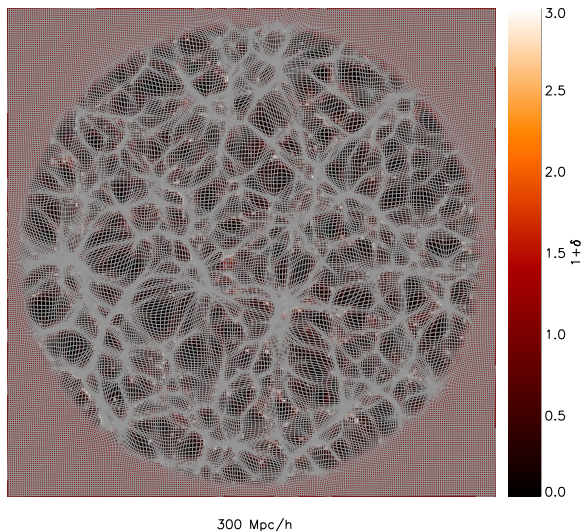


FIG. 16: A slice of the density field from the high-resolution simulation. The curvilinear grid shows the Eulerian coordinate of each grid point of the potential isobaric coordinate. The reconstruction works very well within the spherical region though near the spherical boundary there are some artificial features caused by the discontinuities.

divergence of  $P_{v_{Bz}^s}(k)$ . This implies that the random velocity dispersion is large on small scales, which leads to the so-called Finger of God effect.

Figure 15 shows the two-dimensional power spectrum of the negative divergence of the shift velocity  $-\nabla \cdot \mathbf{v}^s$ . We plot the logarithmic value of the power spectrum for display purpose. Most power of  $-\nabla \cdot \mathbf{v}^s$  is concentrated in the  $\mu$  bins close to  $\mu \sim 1$  as expected. This causes part of the degradation of reconstruction performance along the line of sight direction.

#### Appendix D: Reconstruction with survey mask

To apply nonlinear reconstruction to real observation data, we need to consider the boundary of a survey volume. In this Appendix we present a simple test of reconstruction without the periodic boundary conditions. We take a spherical region with radius  $140.625 h/\text{Mpc}$  from the high-precision simulation and set the density fluctuations  $\rho(\mathbf{x})/\bar{\rho}$  in the outer space to 1. Then we perform reconstruction with this density field. Figure 16 shows a slice through the density field. We also plot the Eulerian positions of a uniform grid of potential isobaric coordinates on the density field slice. The reconstruction

works very well within the spherical region though near the spherical boundary there are some artificial features caused by the discontinuities. However, this will cause little degradation of reconstruction as the measure of a 2D boundary is nearly zero in the 3D space.

#### Appendix E: Reconstruction with halos

The dark matter field is nearly constant in Lagrangian space, while the halo field is not uniform in Lagrangian space. However, when we apply reconstruction, we still map the halo density field into a nearly uniform distribution as

$$\rho_h(\mathbf{x})d^3x = \rho(\mathbf{q})d^3q, \quad (\text{E1})$$

where  $\rho_h(\mathbf{x})$  is the halo density field in Eulerian space. As a result, the reconstructed displacement is biased,

$$\psi_h(\mathbf{k}) = b(k)\psi(\mathbf{k}), \quad (\text{E2})$$

where  $b(k)$  is the bias factor of the displacement reconstructed from the halo field. At linear order the displacement bias is equal to the constant Eulerian halo bias,

$$\delta_h(\mathbf{x}) = b_1\delta(\mathbf{x}), \quad (\text{E3})$$

where  $b_1$  is the linear halo bias. However, the mapping from real to redshift space for the halo density field is still defined as the dark matter field

$$\rho_h(\mathbf{s})d^3s = \rho_h(\mathbf{x})d^3x, \quad (\text{E4})$$

where  $\rho_h(\mathbf{s})$  is the redshift space halo density field. Thus the displacement reconstructed from halos in redshift space is

$$\psi_h^s(\mathbf{k}) = b(k)\psi(\mathbf{k}) + \mathbf{v}^s(\mathbf{k})/aH. \quad (\text{E5})$$

To quantify the performance of reconstruction with halos in redshift space, we should correlate the reconstructed field with  $(b_1 + f\mu^2)\delta_L(\mathbf{k})$  instead of  $(1 + f\mu^2)\delta_L(\mathbf{k})$ .

Figure 17 show the two-dimensional cross correlation coefficients of the halo reconstructed density field with the linear field  $(b_1 + f\mu^2)\delta_L(\mathbf{k})$ . Three halo samples with number densities of  $2.77 \times 10^{-2}$ ,  $2.77 \times 10^{-3}$ , and  $2.77 \times 10^{-4} (h/\text{Mpc})^3$  are used for reconstruction. The bias factors are  $b_1 = 0.68, 0.92$ , and  $1.44$ , respectively. We find that the correlation coefficients have the same anisotropic features as the result obtained with the dark matter density field, but the performance degrades with the lower number density. And also the less dense sample suffers less from the RSD effect.

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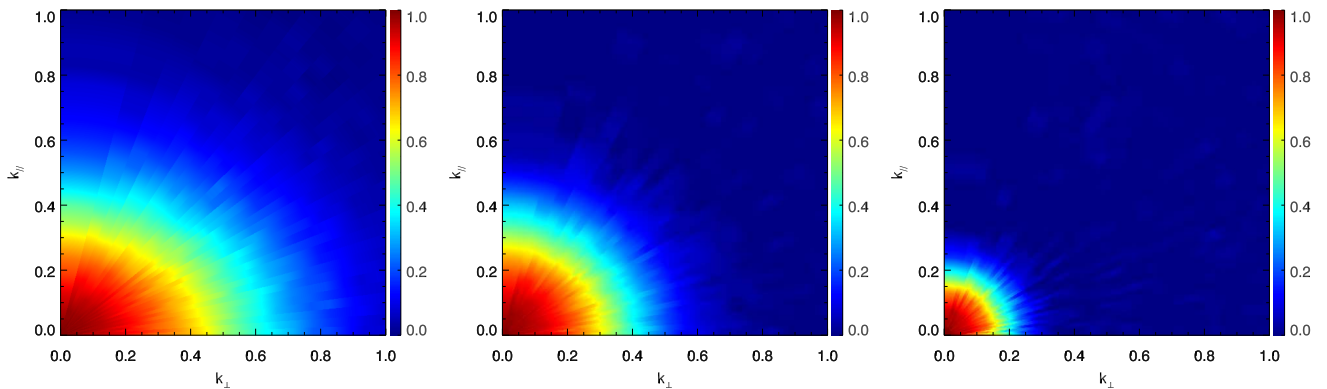


FIG. 17: The two-dimensional cross correlation coefficients of the halo reconstructed field with the linear field  $(b_1 + f\mu^2)\delta_L(\mathbf{k})$  for  $\bar{n} = 2.77 \times 10^{-2}$ ,  $2.77 \times 10^{-3}$ ,  $2.77 \times 10^{-4}$   $(h/\text{Mpc})^3$  (from left to right), respectively.

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