

Observing relativistic features in large-scale structure surveys – II. Doppler magnification in an ensemble of relativistic simulations

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ABSTRACT

The standard cosmological model is inherently relativistic, and yet a wide range of cosmological observations can be predicted accurately from essentially Newtonian theory. This is not the case on ‘ultralarge’ distance scales, around the cosmic horizon size, however, where relativistic effects can no longer be neglected. In this paper, we present a novel suite of 53 fully relativistic simulations generated using the *gevolution* code, each covering the full sky out to $z \approx 0.85$, and approximately 1930 deg^2 out to $z \approx 3.55$. These include a relativistic treatment of massive neutrinos, as well as the gravitational potential that can be used to exactly calculate observables on the past light cone. The simulations are divided into two sets, the first being a set of 39 simulations of the same fiducial cosmology (based on the *Euclid* Flagship 2 cosmology) with different realizations of the initial conditions, and the second that fixes the initial conditions, but varies each of seven cosmological parameters in turn. Taken together, these simulations allow us to perform statistical studies and calculate derivatives of any relativistic observable with respect to cosmological parameters. As an example application, we compute the cross-correlation between the Doppler magnification term in the convergence, κ_v , and the CDM + baryon density contrast, δ_{cb} , which arises only in a (special) relativistic treatment. We are able to accurately recover this term as predicted by relativistic perturbation theory, and study its sample variance and derivatives with respect to cosmological parameters.

Key words: cosmological parameters – large-scale structure of Universe.

1 INTRODUCTION

The large-scale distribution of matter in our Universe will be studied at ever greater detail with upcoming astronomical surveys such as LSST (Abell et al. 2009), *Euclid* (Laureijs et al. 2011), DESI (Levi et al. 2013), and the *Roman Telescope* (Spergel et al. 2015). The complicated astrophysical feedback processes that plague the dynamics of galaxies and clusters become subdominant at cosmological distance scales, making large-scale structure an ideal laboratory for probing gravity, which is effectively the only important force at those scales. This is usually done by extracting summary statistics such as power spectra, bispectra, etc. (Yoo, Fitzpatrick & Zaldarriaga 2009; Bonvin & Durrer 2011; Leclercq, Pisani & Wandelt 2014). As the volume of surveys increases, these will be measured more accurately and for an increasing number of modes. As a result, previously unconstrained small effects can be detected with high significance.

Cosmological N -body simulations provide a powerful and versatile means to predict these summary statistics, given a cosmological model. These simulations commonly use Newtonian theory (Teyssier 2002; Springel 2005), which is sufficient for many purposes, in particular in the context of the Λ CDM concordance model. At extremely large distance scales, the interpretation of such simulations

becomes subtle, however (Chisari & Zaldarriaga 2011; Green & Wald 2012; Rigopoulos & Valkenburg 2015). One way of maintaining consistency with general relativity at leading order is by using so-called Newtonian motion gauges (Fidler et al. 2017), but the full machinery for analysing simulations in this context still needs to be developed.

Alternatively, general relativity can be implemented in the simulations explicitly. Employing techniques from numerical relativity, this has been explored e.g. in Giblin, Mertens & Starkman (2016) and Macpherson, Lasky & Price (2017). The main drawback of this formulation is the requirement to keep track of the wave-like solutions of the gravitational field, which needs extremely fine time resolution and thus leads to practical limitations. In the weak-field regime relevant for cosmology, however, one can easily perform a scalar-vector-tensor decomposition of the gravitational field in order to isolate these wave-like components. They can then be treated with fast approximate methods that completely remove the limitation on the time-stepping. Such an approach is implemented in the weak-field relativistic N -body code *gevolution* (Adamek et al. 2016a,b) that we employ in this work. Relativistic effects that appear at extremely large distance scales are naturally included in our numerical simulations. Physical quantities, whenever they are gauge-dependent, are computed in the Poisson gauge, which is widely employed in practical calculations. Furthermore, using ray tracing, we can compute physical observables exactly and directly.

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While the final observables (i.e. the observed summary statistics) in actual surveys contain the fully aggregated information of how matter is distributed *and* observed on our past light cone, in the weak-field regime, it can often be useful to study different effects separately. In the literature, one often finds that any effects that are not explained by weak lensing or specifically the Kaiser-type redshift-space distortions (Kaiser 1987) are called *relativistic* effects, even though confusingly some of them would still be captured in Newtonian simulations, and even though weak lensing and redshift-space distortions are both arguably relativistic in nature as well. Some examples include corrections in the two-point correlation function, which can be as large as 10 per cent (Bertacca et al. 2012; Yoo & Desjacques 2013; Bonvin 2014; Raccanelli et al. 2016; Lorenz, Alonso & Ferreira 2018; Tansella et al. 2018; Beutler & Di Dio 2020), or in bispectra where the signal-to-noise ratio for the relativistic part is ~ 10 for a survey like *Euclid* (Umeh et al. 2017; Bertacca et al. 2018; Jolicoeur et al. 2018; Clarkson et al. 2019; De Weerd et al. 2020; Jolicoeur et al. 2020; Maartens et al. 2020). Such corrections are also relevant for the study of non-Gaussianity and bias (Bruni et al. 2012; Alonso et al. 2015; Camera, Maartens & Santos 2015; Fonseca et al. 2015; Umeh et al. 2019; Wang, Beutler & Bacon 2020). The benefit of inherently relativistic simulations is that *all* such effects are transparently included, and so there is no ambiguity in the predictions for observable quantities.

In this paper, we present the UNITY simulations, a set of 53 fully relativistic N -body simulations for which we have retained ~ 35 TB of data in the form of HEALPIX maps of different fields as a function of comoving distance from a chosen observation point, and many power spectra, etc., derived from these fields. A large portion of our simulations are run using the same fiducial cosmology, but using a varying random seed to generate the initial conditions, so as to give us multiple random realizations of the same underlying cosmology. The rest of the simulation suite contains pairs of simulations with the same random initial conditions, but each of the cosmological parameters varied by some percentage around the fiducial value. This allows us to compute numerical derivatives to determine the effect of each parameter on various summary statistics and observables. A full ray-tracing procedure can be applied to the stored data in order to produce light cones, or more selective treatments can be used to study particular fields and observables in isolation. The data are available on request.¹

The layout of this paper is as follows. In Section 2, we describe the simulations in more detail, including the data products that are available. We then show some examples where the simulation data are used to extract a relativistic signal in Section 3, and then finally we conclude in Section 4.

2 SIMULATIONS

For our simulations, we use the N -body code *gevolution*. The code is described in full in Adamek et al. (2016a), but we will give a brief overview of its workings here. *gevolution* employs the Friedmann–Lemaître–Robertson–Walker (FLRW) metric with perturbations in Poisson gauge and a weak-field setting where all gravitational fields (ϕ , ψ , B_i , and h_{ij}) are small. These metric quantities are stored on a Cartesian grid and are evolved together with the N -body ensemble that describes the cold dark matter (CDM) in phase space. The joint evolution is therefore computed using a particle-mesh approach, keeping a fixed and uniform resolution on the mesh. The metric

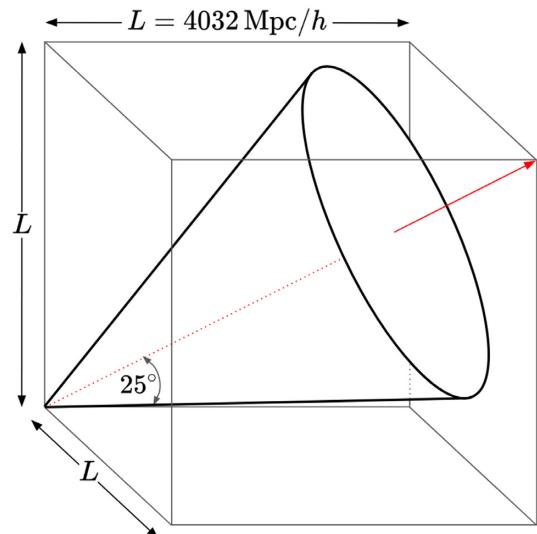


Figure 1. Schematic representation of the pencil-beam light cone construction. The observer is located at the vertex of the cone, with the line of sight along the diagonal of the simulation box (depicted by the red-dotted line and arrow). The opening half-angle of the pencil-beam light cone is 25° .

quantities can be saved on spatial hypersurfaces at any redshift alongside a full particle snapshot. The newest version of *gevolution*² also allows the data to be saved on a series of light cones, given a specific (or several) observers. In this case, the metric quantities are saved on a series of HEALPIX³ (Górski et al. 2005; Zonca et al. 2019) maps in an approach that is better adapted to the geometry of the problem.

2.1 The UNITY suite of simulations

Our simulations have a box volume of $(4032 \text{ Mpc } h^{-1})^3$, where all of the metric quantities are calculated on a Cartesian grid with 2304^3 grid points. This means that we have spatial and mass resolutions of $1.75 \text{ Mpc } h^{-1}$ and $\approx 4.6 \times 10^{11} \text{ M}_\odot h^{-1}$, respectively. We use 2304^3 particles to sample CDM and baryons, while the massive neutrino species are treated with a grid-based approach similar to the one described in Brandbyge & Hannestad (2009) that uses the linear transfer functions. For the case where we explore a non-standard equation-of-state parameter of dark energy, $w_0 = -0.9$ instead of $w_0 = -1$, we use a very similar approach to account for the perturbations of the dark energy fluid; see Dakin et al. (2019) and, in particular, Hassani et al. (2019) for details. As these perturbations are often neglected in Newtonian simulations of evolving dark energy, we also run a simulation for comparison where the dark energy is perfectly homogeneous (in Poisson gauge).

Each simulation has two light cones with the same observer at the corner of the box. The first light cone extends out to a distance of $2015 \text{ Mpc } h^{-1}$ and covers the full sky⁴ occupying a corner of the box, whereas the second one, illustrated in Fig. 1, extends further out to a distance of $4690 \text{ Mpc } h^{-1}$ with a disc-shaped survey area of approximately 1932 deg^2 . The pointing at the centre of the disc is towards the opposite corner of the box to allow us the

²<https://github.com/gevolution-code/gevolution-1.2>.

³<http://healpix.sourceforge.net>.

⁴To build the full-sky case, we use periodic boundary conditions, with each eighth of the sphere then.

¹<http://philbull.com/unity>.

Table 1. Details of the UNITY simulations.

No. sims	Varied parameter	Values	Gauge
34	Initial conditions	(random seed)	Poisson
5	Initial conditions	(random seed)	<i>N</i> -Body
2	n_s	0.69 ± 5 per cent	Poisson
2	A_s	$2.1 \times 10^{-9} \pm 5$ per cent	Poisson
2	h	0.67 ± 5 per cent	Poisson
2	ω_b	0.021996 ± 10 per cent	Poisson
2	ω_{cdm}	0.121203 ± 5 per cent	Poisson
2	M_ν	(0.1, 0.2) eV	Poisson
2	w_0	-0.9	Poisson

Notes. The first row is based on the fiducial cosmology, with multiple different realizations. This gives the ability to study the covariance. The second row is also run using the fiducial cosmology, but *gevolution* is run in the *N*-body gauge. These can be used to compare and determine gauge effects in our simulation. Finally, we list a series of pairs of simulations where we vary one of the parameters around the fiducial value. This allows us to calculate finite differences on different statistics.

maximum distance without repeating data, which could add spurious correlations (especially on large scales).

The UNITY simulations consist of a total of 53 simulations, the parameters of which are summarized in Table 1. Of these, 39 use the same fiducial cosmology of $n_s = 0.96$, $A_s = (2.1 \times 10)^{-9}$, $h = 0.67$, $\omega_b = 0.021996$, $\omega_{\text{cdm}} = 0.121203$, $M_\nu = 0.06$ eV, and $w_0 = -1$. These parameters match the fiducial cosmology of the *Euclid* Flagship 2 simulations (Knabenhans et al. 2019). For each of these simulations, we vary the random seed, which gives us a different realization of initial conditions for the same cosmology.

For the remaining simulations, we vary several of the parameters individually, in turn, adding or subtracting a small amount from the fiducial value in such a way that we can form finite-difference derivatives with respect to the parameters at a later stage,

$$\frac{\partial S}{\partial \theta} \simeq \frac{S(\theta + \Delta\theta) - S(\theta - \Delta\theta)}{2\Delta\theta}, \quad (1)$$

where S is an observable quantity of interest and θ is the respective cosmological parameter. We varied each parameter by 5 per cent, as if $\Delta\theta$ is too small, the results for the finite difference will be dominated by numerical noise, since the difference between the observables will in many cases be minimal. For ω_b and w_0 , we instead use a larger variation of 10 per cent. For the simulations where M_ν is varied, we take a slightly different approach, since the neutrino mass scale is less well constrained and we would therefore like to allow for larger excursions. We considered cases with a total neutrino mass of 0.1 and 0.2 eV (while ensuring the same squared-mass differences between the three mass eigenstates). Note that we did not keep Ω_m fixed in these cases.

2.2 Data products

We retain several different types of data product from the simulations, giving us enough flexibility to calculate a wide range of observable quantities, but without needing to retain the full particle catalogue and metric perturbations on a grid for many snapshots, which would result in a very large data volume.

2.2.1 HEALPIX maps

Any calculations that we wish to do on the light cone are simplified by saving the data in spherical shells about a pre-specified observing location, instead of on a Cartesian grid for many snapshots in time.

We use the HEALPIX pixelization to store the data for each field and for each shell. These maps are saved at a spatial resolution in radial slices of comoving width $1.75 \text{ Mpc } h^{-1}$ and cover both the full-sky and the pencil-beam light cone. The N_{side} of the maps varies as a function of the distance from the observer, always maintaining a sampling of the fields close to the resolution of the simulation and reaching a maximum of $N_{\text{side}} = 2048$.

The quantities saved on these maps are the scalar potential ϕ , the peculiar velocity field (coarse-grained at the scale of the pixel) projected on to the line of sight, the CDM + baryon (cb) density, and the neutrino (ν) density. These quantities are calculated, as usual in *gevolution*, on a Cartesian grid, but are interpolated on to an appropriate set of HEALPIX maps around the observer at each time-step using trilinear interpolation, and then saved to disc.

In Fig. 2, we show examples of these HEALPIX maps for both the δ_{cb} and the δ_ν fields. In the left-hand panel, we have these saved for the full-sky light cone at a redshift of $z \approx 0.1$, and in the right-hand panel, we show the pencil-beam light cone at a redshift of $z \approx 1$.

2.2.2 Power spectra

For each simulation, we also store a set of power spectra for 14 different redshift slices: $z = 50, 30, 10, 4, 3, 2.5, 2, 1.5, 1, 0.75, 0.5, 0.25, 0.1, 0$. Thanks to *gevolution*'s ability to calculate all metric data for the entire simulation, we are able to store power spectra for a multitude of variables. In this case, we extract power spectra for both the scalar and vector gravitational potentials, as well as for the CDM + baryon density and the total matter density (i.e. including the massive neutrinos). This results in $14 \times 4 = 56$ power spectra per simulation, which can be used directly without needing to reanalyse the simulation data.

In Fig. 3, we plot the mean of all of the CDM + baryon power spectra over the 34 random realizations of the fiducial cosmology at $z \simeq 0$, plus confidence intervals showing the expected sample variance (estimated from the simulations) and the error on the mean. We also plot theoretical predictions for comparison, calculated using a linear power spectrum from CLASS (Blas, Lesgourges & Tram 2011) and a non-linear power spectrum model from HMcode (Mead et al. 2016). The deviations at large scales are due to the different gauges (Poisson gauge for *gevolution* and the linear prediction of CLASS, while HMcode uses synchronous gauge), while there is good agreement on intermediate and small scales (up to the expected non-linear corrections).

2.2.3 Angular power spectra

For completeness, in Fig. 4, we present the angular power spectra computed using full-sky light cones from the 34 random realizations of the fiducial cosmology. We employ the simple quadratic estimator,

$$\hat{C}_\ell = \frac{1}{2\ell + 1} \sum_{m=-\ell}^{\ell} a_{\ell m} a_{\ell m}^*, \quad (2)$$

where $a_{\ell m}$ are the spherical harmonic expansion coefficients of some scalar field A projected on to the unit sphere (Leistedt et al. 2013). For this example, we took $A = \delta_{\text{cb}}$ and performed the analysis using the HEALPIX *anafast* routine, which is suitable for the full-sky case. The shaded region in Fig. 4 corresponds to the standard deviation of

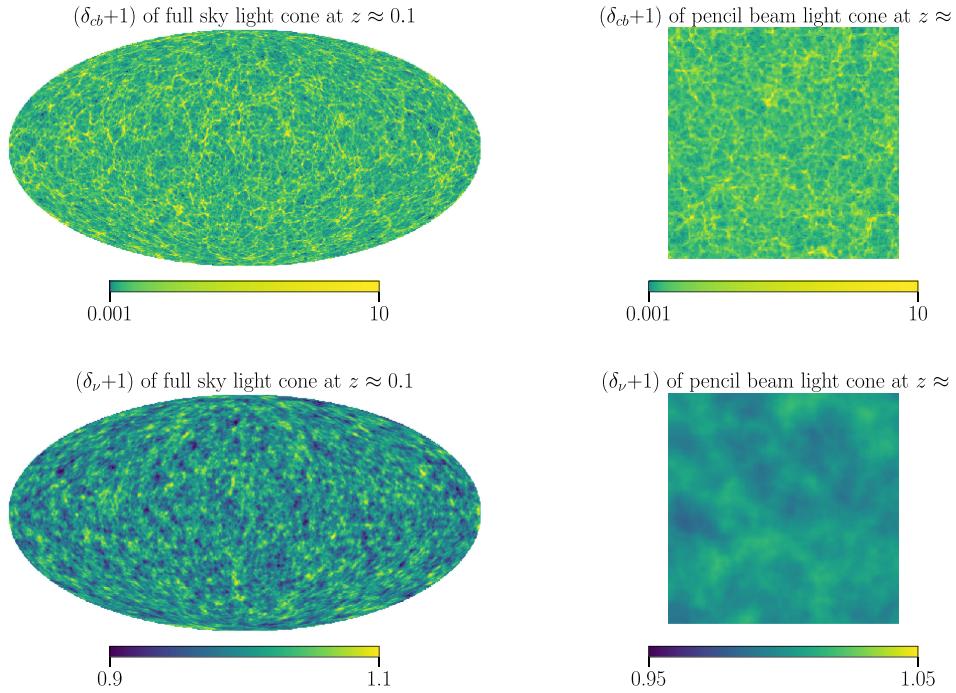


Figure 2. HEALPix maps for both the CDM + baryon density field and the neutrino density fields. We show these for the full-sky light cone at a redshift of $z \approx 0.1$, and also for the pencil-beam light cone at a redshift of $z \approx 1$. For the pencil beam, we plot a $10^\circ \times 10^\circ$ section.

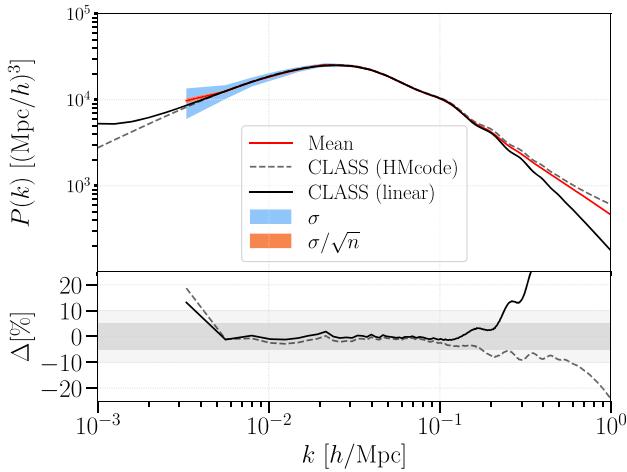


Figure 3. Top panel: power spectrum of the CDM + baryon distribution at $z \approx 0$. The solid red line shows the mean value of the 34 random realizations of the fiducial cosmology, with the standard deviation (and error on the mean) shown as blue and orange 68 per cent confidence intervals, respectively. The black solid and dashed lines show theoretical predictions using the CLASS linear power spectrum, and HMcode non-linear model, respectively. HMcode uses the synchronous gauge while the other cases are in Poisson gauge, which explains the different behaviour as one approaches the horizon scale. Bottom panel: fractional difference between the theoretical predictions and the mean. The light and dark grey regions show 10 and 5 per cent differences between the theory and the mean, respectively.

\hat{C}_ℓ over the 34 samples, with the theoretical angular power spectrum C_ℓ computed as

$$C_\ell(z) = \frac{2}{\pi} \int k^2 dk P_{\text{cb}}^s(k, z) j_\ell^2(kr(z)), \quad (3)$$

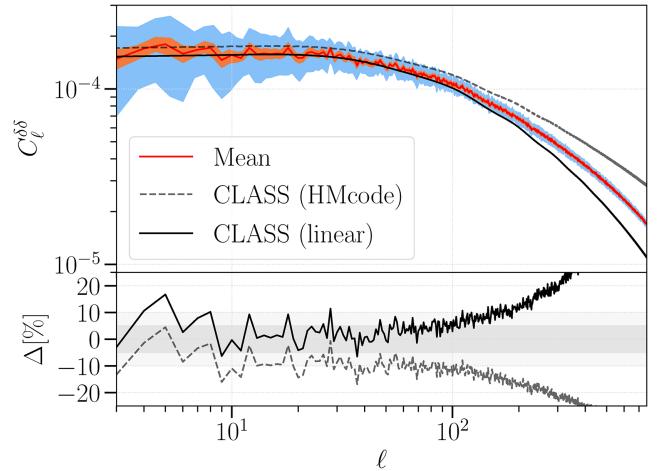


Figure 4. Top panel: angular power spectrum of the CDM + baryon field evaluated on the past light cone at $z \approx 0.58$. The mean and standard deviation over the 34 random realizations of the fiducial cosmology are shown as the solid red line and blue shaded region respectively, while the orange region shows the corresponding standard error of the mean (as in Fig. 3). We also show theoretical predictions based on the linear power spectrum for CDM and baryons from CLASS, and the non-linear power spectrum from HMcode, convolved with the CIC kernel (see Section 3.2). Bottom panel: fractional difference between the theoretical predictions and the mean. The light and dark grey regions show 10 and 5 per cent differences between the theory and the mean respectively.

where $P_{\text{cb}}^s(k, z)$ is the power spectrum accounting for baryons and CDM only, smoothed by the cloud-in-cell (CIC) kernel, as discussed in Section 3.2. We projected each map at $z \approx 0.58$, which is equivalent to a comoving distance of $r(z) = 1500 \text{ Mpc } h^{-1}$.

In this case, the theoretical calculations are broadly consistent with the measured angular power spectrum, but a slight amplitude offset can be seen, particularly for the non-linear theoretical curve at $\ell \lesssim 100$. From several consistency checks, we have found a contribution coming from the mildly and fully non-linear scales that leaks to large angular scales from the projection integral (equation 3). By comparing the output angular power spectrum from CLASS, for CDM + baryons, the same offset was observed. However, as we can see, both linear and non-linear predictions match the mean value of the simulations at a level of 10 per cent for $\ell \lesssim 100$. On the other hand, the discrepancies at higher ℓ are primarily caused by resolution effects.

2.2.4 Ray-traced quantities

Given a set of HEALPIX maps of the gravitational potential ϕ on the light cone, it is straightforward to construct other interesting quantities that are linearly related to ϕ or its time derivative. Examples include the integrated Sachs–Wolfe effect and the weak-lensing potential. Using the Born approximation, HEALPIX maps of these quantities can be computed directly in pixel space by adding together maps of ϕ with appropriate weights. For instance, the weak-lensing potential, which is defined in Lewis & Challinor (2006) as

$$\Psi(\theta, z) \equiv - \int_0^{r(z)} dr' \frac{r(z) - r'}{r(z)r'} (\phi + \psi), \quad (4)$$

can be constructed by such a procedure, explained in more detail in Lepori et al. (2020). Here we make the assumption that $\psi \approx \phi$, which is an excellent approximation here, and neglect the effect of frame dragging. The HEALPIX maps can then be easily converted into linear weak-lensing convergence and shear maps by using

$$\kappa_g = -\frac{1}{2} \Delta \Psi, \quad (5)$$

$$\gamma_1 + i\gamma_2 = -\frac{1}{2} (\nabla_1 \nabla_1 - \nabla_2 \nabla_2) \Psi - i \nabla_1 \nabla_2 \Psi, \quad (6)$$

where Δ is the Laplacian, and the derivatives are taken on the map. The convergence κ_g (we use the subscript ‘g’ to distinguish this term from Doppler magnification, as discussed below) and the shear γ parametrize the amplification matrix,

$$A = \begin{pmatrix} 1 - \kappa_g - \gamma_1 & -\gamma_2 \\ -\gamma_2 & 1 - \kappa_g + \gamma_1 \end{pmatrix}, \quad (7)$$

which relates lensed images to unlensed ones if lensing is treated linearly (Bartelmann & Schneider 2001). It is worth pointing out that non-linear ray tracing is also possible with the data, e.g. using the full machinery developed in Lepori et al. (2020), although we do not pursue this here.

3 RESULTS

In this section, we show one of the many ways these simulations can be used to construct relativistic observables by using the *Doppler magnification* effect as an example. This was first highlighted and investigated as an observable in its own right by Bonvin, Durrer & Gasparini (2006) and Bonvin (2008). Later works have shown that this signal should be detectable with modern day optical and radio surveys (e.g. Bonvin et al. 2017; Andrianomena et al. 2019).

A perturbative expression was derived for the cross-correlation between the Doppler magnification signal and the matter density in Bacon et al. (2014). Doppler magnification is a relativistic effect that

is caused by the relative motion of the source and the observer, which correlates with matter density as sources will tend to fall towards areas of high density. It is an inherently (special) relativistic effect, and while it is also possible to derive it from Newtonian simulations, this requires a more careful handling of gauge issues, etc., than is needed here. To illustrate their utility, we will go through the steps in calculating this signal within our suite of simulations and then compare with the perturbative results from Bacon et al. (2014).

3.1 Doppler magnification

In the linear weak-lensing regime, the true shape of the source is related to the observed image through the Jacobi map,

$$\mathcal{J} = \bar{D}_A(r) A, \quad (8)$$

where \bar{D}_A is the angular diameter distance to the source in the background metric, and A is the amplification matrix given in equation (7). Doppler magnification appears if one uses the observed redshift as a distance indicator, i.e. the Jacobi map is written at fixed observed redshift as

$$\begin{aligned} \mathcal{J} &= \bar{D}_A(z_s) \left(1 - \frac{\partial \ln \bar{D}_A}{\partial z} \delta z \right) A \\ &\simeq \bar{D}_A(z_s) \begin{pmatrix} 1 - \kappa_g - \kappa_v - \gamma_1 & -\gamma_2 \\ -\gamma_2 & 1 - \kappa_g - \kappa_v + \gamma_1 \end{pmatrix}, \end{aligned} \quad (9)$$

where one defines

$$\kappa_v = \frac{\partial \ln \bar{D}_A}{\partial z} \delta z = \left(\frac{1 + z_s}{H r_s} - 1 \right) \mathbf{v}_s \cdot \mathbf{n}. \quad (10)$$

Here \mathbf{v}_s is the peculiar velocity of the source, and z_s and r_s are the redshift and the comoving radial distance to the source, respectively. We define the direction of \mathbf{n} to be pointed from the observer to the source.

It can be seen that galaxies that have a velocity vector directed towards the observer will create a negative κ_v at low redshift, which means that they will appear smaller in angular size and dimmer when compared to a typical source at the same observed redshift. In contrast, if the source is moving away from the observer, the κ_v will be positive, which means it will appear brighter with a larger angular size. This effect comes about because a surface of fixed observed redshift does not coincide with a surface of fixed comoving distance. Here and in the following, we assume that the Doppler contribution from the source peculiar motion is the only relevant redshift perturbation, i.e. we neglect gravitational redshift and other subdominant corrections. Note that all such corrections *can* be included by combining the appropriate fields when ray-tracing, however.

Another important point to note from equation (10) is that at high redshifts the term in brackets will decrease, causing a lower amplitude of Doppler magnification. Using this fact and the estimate $|\mathbf{v}| \sim H_0 \delta/k$, one expects that the Doppler magnification is only important on large scales and at low redshift (Bolejko et al. 2013; Bacon et al. 2014). Fig. 2 of Bacon et al. (2014) does indeed show that the Doppler magnification term is dominant at medium-to-low redshifts and wavenumbers ($\ell \lesssim 1000$ at $z = 0.2$, and $l \lesssim 100$ at $z = 0.4$).

The relative importance of the different contributions to the observed convergence can also be judged from Fig. 5, where we show the Doppler magnification signal and the weak lensing signal within our simulation. To calculate the Doppler magnification signal, we use equation (10) together with maps of the redshift space distortion

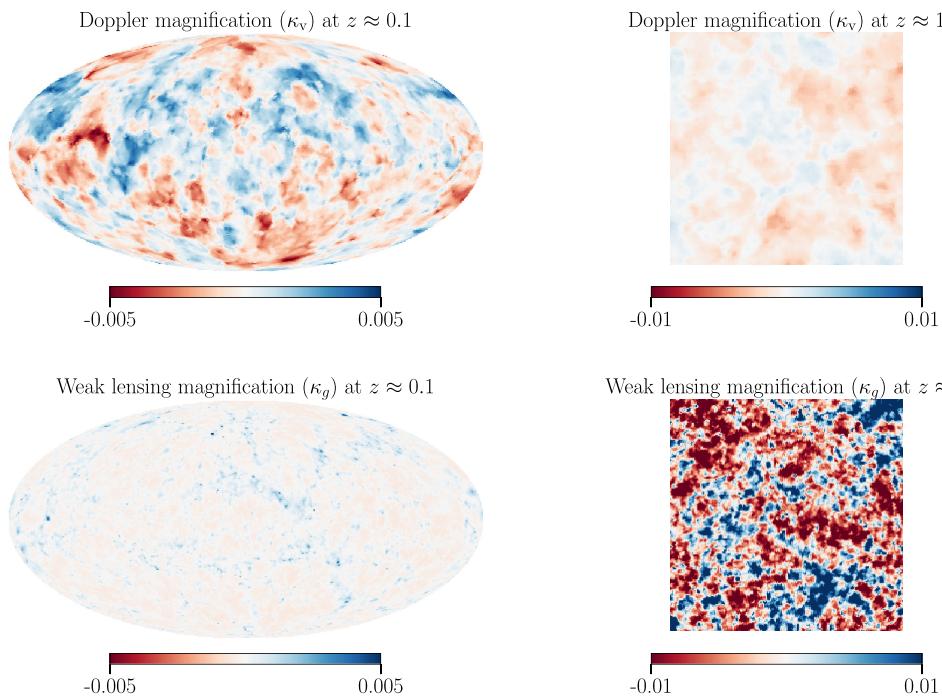


Figure 5. Heat maps for both the Doppler convergence and the weak-lensing convergence. We show the full-sky signal at redshift $z \approx 0.1$ and the pencil beam at $z \approx 1$, plotting a $10^\circ \times 10^\circ$ section. For each redshift, we keep the scale of the colour bars the same to show the difference between the two signals.

field that are included in our data products. For the full-sky maps, we show the signal at $z \approx 0.1$, where it can be seen that the Doppler magnification is much stronger than the weak lensing signal. When we look at the pencil beam at $z \approx 2$, we see that the weak-lensing effect dominates instead, as this integrated effect becomes stronger with increasing distance.

3.2 Density-convergence cross-correlation

Since matter tends to collapse on to massive structures, there is an obvious correlation between the Doppler magnification and the density field. At low redshift, the Doppler magnification of sources on the far side of a large concentration of matter tends to be negative, while the opposite is true for sources on the near side. In order to measure this correlation, we take the average density over a small interval in distance from the observer, $[r, r + \Delta r]$, and compute the angular cross-power with the Doppler convergence, κ_v , evaluated at the far end of the interval. We then compute the average of the resulting cross-power over a larger distance range in order to accumulate a larger total signal.

A full derivation of the angular cross-correlation from perturbation theory is presented in Bacon et al. (2014). We present only the final result in Poisson gauge here. We obtain

$$\begin{aligned} C_{\ell}^{\delta\kappa_v}(r') &= \frac{16\pi^2}{\Delta r} \left(H(r') - \frac{1+z(r')}{r'} \right) \frac{\partial D(r')}{\partial z} \\ &\times \int_0^{\infty} dk P_{cb}(k) k j_{\ell}'(kr') \\ &\times \int_{r'-\Delta r}^{r'} dr \left(D(r) - \left(\frac{3H(r)^2}{(1+z(r))k^2} \right) \frac{\partial D(r)}{\partial z} \right) j_{\ell}(kr), \end{aligned} \quad (11)$$

where $P_{cb}(k)$ is the CDM + baryon power spectrum at redshift zero and $D(r)$ is the linear growth factor,⁵ defined as $\delta(k, r) = D(r)\delta(k, 0)$. Note that, in a slight variation to Bacon et al. (2014), we employ weights that are uniform in comoving distance r instead of weights that are uniform in redshift.

Matter overdensities only have an appreciable gravitational influence over short distances, expected to be somewhere in the region of tens of megaparsecs. Without trying to make an optimal choice, we set $\Delta r = 52.5 \text{ Mpc } h^{-1}$ for the distance window in which the density is computed, ignoring longer range correlations. The cross-correlation signal $C_{\ell}^{\delta\kappa_v}(r')$ can then be averaged over a broader bin to get the average angular cross-correlation within the bin, which we will denote as $C_{\ell}^{\delta\kappa_v}$.

To compute this value in our simulations, we use the HEALPIX maps of the line-of-sight peculiar velocity field $\mathbf{v}_s \cdot \mathbf{n}$ and the CDM + baryon density as described in Section 2.2. Specifically, for each radial shell we use maps from the two consecutive simulation time-steps that together enclose the light cone at the given distance and per cent. For each of these radial shells, we save two additional shells at time-steps on either side. This allows us to calculate the data on the null hypersurface by linear interpolation in conformal time. From these data, we create a thin bin of $\Delta r = 52.5 \text{ Mpc } h^{-1}$ where we sum up the density maps, and then cross-correlate with the Doppler magnification map directly on the far side of the bin, which we compute from equation (10). The cross-correlation is calculated using the HEALPIX `anafast` function. We then repeat

⁵In equation (11), scale-independent growth has been assumed for simplicity. This is a good approximation on scales much smaller than the neutrino free-streaming scale (and in the case of $w_0 < -1$, the sound horizon scale of dark energy perturbations), where D is computed neglecting any perturbations in components other than CDM and baryons. Scale-dependent growth is fully taken into account in the simulations themselves.

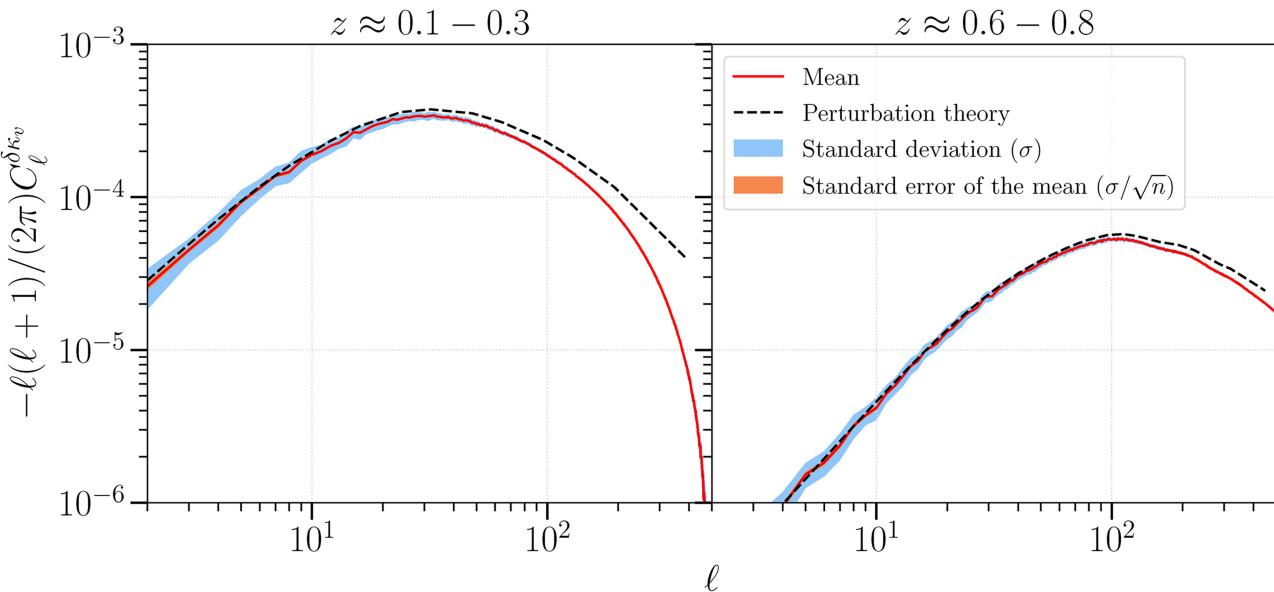


Figure 6. Plot of the mean $C_\ell^{\delta\kappa_v}$ across all of the different random realizations of the fiducial cosmology for two different redshift bins. We also plot the standard deviation and the standard error of the mean that are represented by the shaded areas. The dashed line on the plot shows the theoretical prediction of the signal from equation (11).

for all thin bins within the thick bin and take the average to obtain $C_\ell^{\delta\kappa_v}$.

To accurately compare the simulations to the perturbative prediction, it is necessary to convolve the power spectrum in the perturbative calculation with the CIC kernel (Hockney & Eastwood 1981). This effectively gives a smoothed power spectrum which accounts for the fact that the density field in the simulations is coarse-grained at a finite resolution. The expression for the monopole of this smoothed power spectrum is

$$P_0^s(k) = P(k) \frac{1}{4\pi} \int d^2n \, W_{\text{CIC}}^2(k, n), \quad (12)$$

where the CIC kernel $W_{\text{CIC}}(k)$ is defined as

$$W_{\text{CIC}}(k) = \text{sinc}^2\left(\frac{\pi k_1}{2k_N}\right) \text{sinc}^2\left(\frac{\pi k_2}{2k_N}\right) \text{sinc}^2\left(\frac{\pi k_3}{2k_N}\right), \quad (13)$$

where k_i is the i th component of k and k_N is the Nyquist wavenumber. The smoothed monopole power spectrum of equation (12) is then substituted for P_{cb} in equation (11).

In Fig. 6, we show the distribution of $C_\ell^{\delta\kappa_v}$ measured from all realizations of the fiducial cosmology, with the mean shown as a solid red line and the standard deviation shown as a blue shaded region. We also plot the perturbative prediction calculated from equation (11) as a black dashed line. The results are shown for two redshift bins, the first at $z \approx 0.1\text{--}0.3$, and the second at $z \approx 0.6\text{--}0.8$.

In the lower redshift bin, perturbation theory overestimates the signal beyond $\ell \gtrsim 30$ or so. This difference is mostly due to non-linear effects on small and intermediate scales that are not included in the linear power spectrum model, caused by orbit crossings that generate both vorticity and velocity dispersion and at the same time reduce the power in the velocity divergence (Pueblas & Scoccimarro 2009; Hahn, Angulo & Abel 2015; Jelic-Cizmek et al. 2018). As expected, this effect is also present in the density–velocity cross-correlation. Since this bin is at relatively low redshift, non-linear effects are important even at quite low values of ℓ . In the higher redshift bin, on the other hand, the value measured from our simulations fits more closely to the perturbative prediction, although the strength of

the signal has decreased by around an order of magnitude by this point. Also, note the shift of the peak in the cross-correlation to correspondingly smaller angular scales.

In Fig. 7, we show the numerical derivatives of $C_\ell^{\delta\kappa_v}$ by using finite differences (equation (1)) of this quantity in the two redshift bins, $z \approx 0.1\text{--}0.3$ (upper panels) and $\approx 0.6\text{--}0.8$ (lower panels), for seven pairs of simulations with the same initial conditions but different values of the cosmological parameters A_s , n_s , h , ω_{cdm} , ω_b , M_v , and w_0 as described in Table 1. In the case of varying w_0 , the pair of simulations used are the baseline cosmology and the one where we include perturbations in the dark energy field (see Section 2.1). Derivatives of this kind are useful for Fisher forecasting studies, and give a direct measure of the sensitivity of an observable to a given parameter.

Theoretical predictions of the derivatives from perturbation theory are also shown as dashed lines in Fig. 7. As in previous figures, we see differences of a few per cent between the perturbation theory and simulated quantities in the lowest redshift bin (upper panels), which is consistent with the growing importance of non-linear effects at these redshifts, as discussed above. At higher redshift (lower panels), these effects are subdominant however, and the agreement between perturbation theory and the simulated quantities is good to within a couple of percent even on smaller angular scales.

A notable feature of Fig. 7 is the wiggle-like features in some of the curves for the higher redshift bin. These are due to shifts in the location of the baryon acoustic oscillations, which could also be seen (albeit at quite a low level) in Fig. 6. While the signal-to-noise ratio of any practical measurement of the Doppler magnification signal is unlikely to be sufficient to detect the BAO feature for the foreseeable future, it is interesting that they can, in principle, be picked up by this observable.

4 CONCLUSIONS

In this paper, we have presented a suite of novel general relativistic cosmological simulations. These were run using the *gevolution*

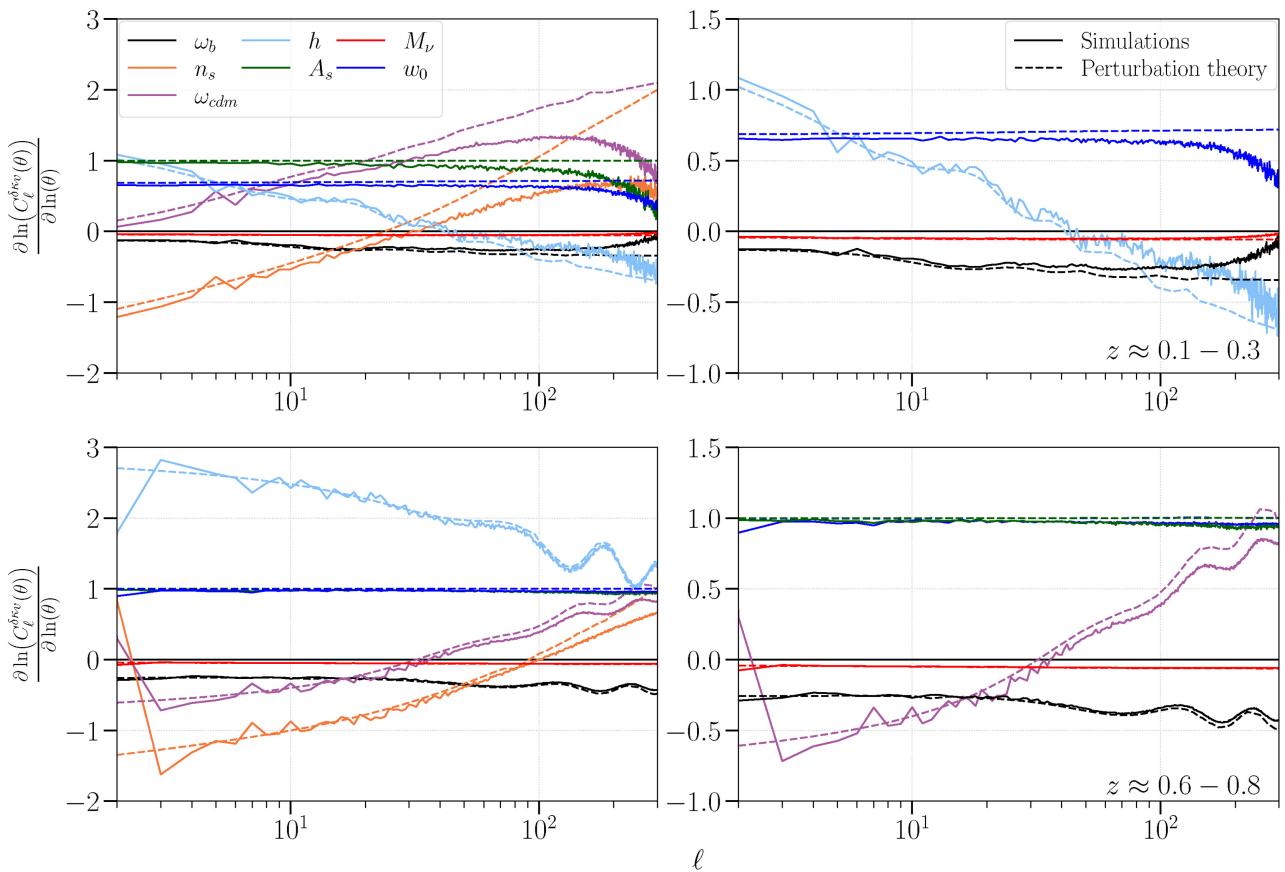


Figure 7. Plots of the numerical derivatives of $C_\ell^{\delta\kappa_v}$ using finite differences of the density Doppler magnification cross-correlation signal in the redshift bin $z \approx 0.1-0.3$ (upper panels) and $\approx 0.6-0.8$ (lower panels). This is done for seven pairs of simulations where $\theta = A_s, n_s, h, \omega_{\text{cdm}}, \omega_b, M_\nu$, and w_0 are varied. The solid lines show the results measured from the simulations, whereas the dashed lines show the perturbative predictions. The right-hand panels are zoomed-in sections of the left-hand panels so the difference in the selected quantities can be seen more clearly.

N-body code (Adamek et al. 2016a), which takes into account all relevant relativistic effects, including frame dragging and relativistic neutrino effects. While most of those effects are small in comparison with the conventional ‘Newtonian’ terms in most large-scale observables, the rapidly increasing precision of upcoming surveys will soon make it impractical to ignore them without risking biases in cosmological parameter estimates.

The suite of simulations contains a total of 53 runs, divided into subsets that are envisioned to have two main uses. The first subset contains 39 simulations that all use the same fiducial cosmology, but vary the random seed used to generate the initial conditions, which essentially gives us different realizations of the same underlying cosmology. Possible applications of this subset include statistical studies of observables, estimators, and data extraction methods, plus some rudimentary kinds of simulation-based covariance estimation.

The second part of this suite consists of seven pairs of simulations with cosmological parameters that are systematically varied around the fiducial cosmology (which matches the *Euclid* Flagship 2 cosmology), while maintaining the same (random) initial conditions. These allow us to study the derivatives of any observable that we can calculate with respect to a set of cosmological parameters. Possible applications of this suite include Fisher forecasting, where derivatives of observables are used to estimate the uncertainties on measurements that can be achieved by future experiments.

We have stored a range of data products for each simulation, including all of the metric degrees of freedom and other fields needed

to reconstruct any cosmological observable on large scales. These fields have been determined in a spherical coordinate system about a fiducial observer, and can be fed into a ray-tracing algorithm to produce precise predictions of observables on the past light cone. The geometry of the simulations has been chosen to maximize the sky area and depth of the light cones that can be simulated, with the full sky accessible out to $z = 0.85$ and a large area (1930 deg^2) available out to $z = 3.55$. These specifications are well matched to a variety of current and near-future large-scale structure surveys, including the ESA *Euclid* mission, the *Roman Space Telescope*, the VRO Legacy Survey of Space and Time, and the Square Kilometre Array. While the simulations do not have sufficient resolution to produce suitable dark matter halo catalogues for these surveys, biased tracers can be painted on to the simulations using other means (e.g. Borzyszkowski, Bertacca & Porciani 2017; Bull 2017; Witzemann et al. 2019; Yip et al. 2019; Farr et al. 2020; Ramanah et al. 2020).

To showcase the potential use-cases of our suite of simulations, we calculate the cross-correlation of the inherently special relativistic Doppler magnification signal, κ_v , and the matter density contrast, δ_{cb} , from our 34 random realization simulations and compare it to the perturbation theory result from Bacon et al. (2014). We find good agreement with the perturbation theory calculation in a relatively high-redshift bin of $z \approx 0.6-0.8$, but find non-negligible corrections in a lower redshift bin of $z \approx 0.1-0.3$, where linear theory overestimates the signal. This is due to non-linear effects that appear on small scales that are not included in the linear

prediction, and which (due to projection effects) affect most of the relevant angular scales at the low redshifts where the Doppler magnification signal is largest. While an improvement, replacing the linear matter power spectrum in the perturbation theory calculation with a non-linear power spectrum model does not fully capture these effects (cf. Figs 3 and 4). This highlights the value of having fully relativistic cosmological simulations on hand to make predictions of such observables.

We also calculated the Doppler magnification cross-correlation signal for matched pairs of simulations with the same initial conditions but different values for the cosmological parameters. This allowed us to approximate the derivatives of the observable $C_{\ell}^{\delta_{kv}}$ and therefore see which parameters it is most sensitive to. While we again saw generally good agreement with predictions from perturbation theory, especially in the higher redshift bin, non-negligible corrections remained. Since reasonably any large-scale structure observable can be constructed from our suite of simulations, including many different combinations of cross-correlations and even high-order statistics, it should be possible to make Fisher matrix-type forecasts for a very wide range of surveys and observables using these data.

In conclusion, in this paper, we have described a suite of fully relativistic N -body simulations, and shown a particular example (the Doppler magnification term in the density-convergence cross-correlation) in which such simulations are needed in order to make accurate predictions for next-generation surveys. While perturbation theory calculations were able to capture most features of the target signal, non-linear effects made few per cent differences at low redshifts. To accurately model these in perturbation theory, one would likely have to include higher order corrections in redshift-space are difficult to compute.

CARBON FOOTPRINT

The numerical simulations presented in this paper used about 3900 kWh of electrical energy. Using a conversion factor of 0.681 kg CO₂ kWh⁻¹ (typical for the UK grid according to myclimate.org, c. 2021 February 12), this gives a carbon footprint of approximately 2.7 t CO₂.

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DATA AVAILABILITY

The data that support the findings of this research are available on request from <http://philbull.com/unity>.

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