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It's not σ_8 : constraining the non-linear matter power spectrum with the Dark Energy Survey Year-5 supernova sample

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ABSTRACT

The weak gravitational lensing magnification of Type Ia supernovae (SNe Ia) is sensitive to the matter power spectrum on scales k > 1h Mpc⁻¹, making it unwise to interpret SNe Ia lensing in terms of power on linear scales. We compute the probability density function of SNe Ia magnification as a function of standard cosmological parameters, plus an empirical parameter A_{mod} which describes the suppression or enhancement of matter power on non-linear scales compared to a cold dark matter only model. While baryons are expected to enhance power on the scales relevant to SN Ia lensing, other physics such as neutrino masses or non-standard dark matter may suppress power. Using the Dark Energy Survey Year-5 sample, we find $A_{mod} = 0.77_{-0.40}^{+0.69}$ (68% credible interval around the median). Although the median is consistent with unity there are hints of power suppression, with $A_{mod} < 1.09$ at 68% credibility.

Key words: gravitational lensing: weak – transients: supernovae – cosmology: dark matter – galaxies: haloes – cosmology: cosmological parameters

1 INTRODUCTION

The Cosmic Microwave Background (CMB) together with the expansion history of the universe at redshifts z < 2, as resolved by SNe Ia or Baryon Acoustic Oscillations (BAO), suggest the universe is geometrically flat, and in the main consistent with cold dark matter (CDM) and a cosmological constant Λ as the dominant energy components at late times. While some hints have arisen recently of an evolving dark energy component (Rubin et al. 2023; DES Collaboration 2024; DESI Collaboration 2024; Camilleri et al. 2024) or physics that mimics it, if the universe is not Λ CDM it is at least very close to it in expansion history.

Nevertheless, measurements of the clustering of matter in Λ CDM differ between those measured in the late universe and those predicted from the spectrum of $O(10^{-5})$ CMB temperature and polarisation

fluctuations projected to the present day using the inferred expansion history and standard gravity (see Abdalla et al. 2022, and references therein). One measure of matter clustering is the dimensionless parameter σ_8 , which is the dispersion of the fractional fluctuation of the matter density $\delta = (\rho_m - \bar{\rho}_m)/\bar{\rho}_m$ at the present day in spheres of size $8h^{-1}$ Mpc, if structures had grown solely by the *linear* growth rate. Clustering in the late universe can be measured by weak gravitational lensing, which constrains a combination of this and the present day matter density Ω_m as $S_8 = \sigma_8 \sqrt{\Omega_m/0.3}$. In particular, a re-analysis of Planck satellite data (Efstathiou & Gratton 2021) gives

 $S_8 = 0.828 \pm 0.016$ (Planck TTTEEE) , (1)

whereas for the late universe the Dark Energy Survey (DES, Abbott et al. 2022) finds

$$S_8 = 0.779^{+0.014}_{-0.015}$$
 (DES Y3 3x2pt) . (2)

These are discrepant by ~ 6% (2.3 σ), and the results for other weak

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lensing surveys (Li et al. 2023; Dalal et al. 2023; Asgari et al. 2021) are consistent with the DES result.

One possible explanation for the difference is that the systematics of galaxy weak lensing measurements may have been underestimated: although different surveys use different analytical choices, some inherent commonality of pipeline remains between them. However, recent progress on photometric redshift calibration (Hildebrandt et al. 2021; Myles et al. 2021), shear measurements (Mandelbaum et al. 2018; Kannawadi et al. 2019; MacCrann et al. 2022) and intrinsic alignments (for example, see Paopiamsap et al. 2024) suggest systematics are not enough to account for the size of the difference, although different pipeline choices may lower the discrepancy to the 1.7σ level (Dark Energy Survey and Kilo-Degree Survey Collaboration 2023).

Another possibility is that the growth of structure on *linear scales* $k < 0.1h \text{ Mpc}^{-1}$ differs from the Λ CDM expectation, perhaps due to modified gravity. Structure on linear scales in the late universe is also measured by the four-point correlation function of CMB temperature fluctuations, induced by weak lensing. Results from the Planck satellite (Planck Collaboration 2020b), South Pole Telescope (Pan et al. 2023) and the Atacama Cosmology Telescope (ACT) (Darwish et al. 2021; Madhavacheril et al. 2024) are all consistent with the Planck TTTEEE power spectrum in a Λ CDM background, as stated in Equation 1. Although the CMB lensing sensitivity peaks between 1 < z < 3, this range is sufficiently close to the redshift range of galaxy surveys to disfavour non- Λ CDM linear growth as an explanation.

A remaining possibility is that structure on *non-linear scales* k > 0.1h Mpc⁻¹ is different from theoretical expectations. While galaxy weak lensing pipelines discard data at highly non-linear scales (the choice of cut varies from survey to survey), they retain data for some distance into the non-linear regime. While the growth of CDM-only matter fluctuations under standard gravity is well-understood, even down to very small scales (Liu et al. 2024), other physical processes may influence the power spectrum on these scales.

In particular, the presence of baryons modifies the CDM-only expectation. The influence of baryons is qualitatively understood as 1) the suppression of structure on *intermediate scales* $0.1 < k < 10^2 h$ Mpc⁻¹ by the pressure of "sub-grid" (meaning below the resolution of the hydrodynamic simulations) energetic outflows from active galactic nucleii and supernovae, and 2) the enhancement of structure on *small scales* $k > 10^2 h$ Mpc⁻¹ by condensation due to cooling. The strength of these two effects, collectively termed baryon feedback, differs between simulations (see next Section), and the methodology to make quantitative predictions of the changes they induce in the matter power spectrum remains an active subject of research (for example, see Mead et al. 2021; Eifler et al. 2015; Mohammed & Gnedin 2018; Schneider & Teyssier 2015; Schneider et al. 2019; Aricò et al. 2021; Lu & Haiman 2021; Lu et al. 2022).

In this work, we make no assumptions baryonic feedback is necessarily the only physics at work on the small-scale power spectrum. The nature of dark matter itself remains elusive. It may be "warm" rather than cold, may interact with itself, or may be made of ultralight particles whose de Broglie wavelength is so large that quantum effects influence their clustering properties. These non-standard dark matter models usually suppress power on small-scales. Furthermore, it is well-known that non-zero neutrino mass suppresses small-scale power by reducing CDM-only fluctuations below the scale of their free-streaming length. These effects are difficult to disentangle from baryon feedback at the scales used in current weak lensing surveys (but see Euclid Collaboration 2024, 2025, for future prospects).

Therefore, in this paper we use a parameter that is physics-agnostic

but encapsulates observational differences with the CDM-only nonlinear power spectrum. Amon & Efstathiou (2022) have proposed a simple empirical model for the power spectrum as

$$P(k,z) = P_{\rm L}(k,z) + A_{\rm mod} \left[P_{\rm NL}(k,z) - P_{\rm L}(k,z) \right] \quad . \tag{3}$$

Here $P_{\rm L}$ is that predicted by the spectrum of perturbations of the CMB evolved according to linear theory in a Flat ACDM background expansion, and $P_{\rm NL}$ is the dark matter only non-linear power spectrum enhanced by collapsed and virialised haloes. $A_{\rm mod,I}$ is a scalar that captures suppression (or enhancement) of structure formation compared to a CDM-only benchmark. It is expressed above as being independent of scale and redshift.

Using data from DES, priors on S_8 and Ω_m from Planck, and omitting the scale cuts used in the canonical analysis of DES data (see Eqn. 2), Preston et al. (2023) find

$$A_{\text{mod},\text{I}} = 0.858 \pm 0.052$$
, $S_8 = 0.811 \pm 0.01$, (4)

where we have used the subscript I to denote this result is derived from intermediate scales.

In isolation, this result represents a repack of the S_8 tension by the introduction of the A_{mod} nuisance parameter: that the value of S_8 is consistent with the CMB is unsurprising given the prior. It was also not possible to say the A_{mod} model was preferred by the data: the χ^2 fit is slightly worse compared to the same data analysed without the Planck prior and $A_{mod} = 1$ (see rows 4 and 6 of Table 2 of Preston et al. 2023). However, the value of $A_{mod,I}$ seems consistent with the median of a range of hydrodynamical simulations (see Figure 2 of Preston et al. 2023) for the scales the data applies to. Similar conclusions were previously reached with a reanalysis of data from the Kilo Degree Survey (KiDs) by Amon & Efstathiou (2022).

Our motivation in this paper is to examine the $A_{\rm mod}$ model using the weak lensing of SNe Ia. It has been recently detected at ~ 6σ significance that SNe Ia are weakly lensed by foreground matter (Shah et al. 2024b). The SNe Ia are dimmer if seen through voids, and brighter if seen on over-dense lines of sight (LOS). We will show in the next section that the size of this variation depends on the amplitude of the matter power spectrum on *non-linear scales* rather than the linear scales described by σ_8 or S_8 .

Weak lensing generates a non-Gaussian distribution of SN Ia magnitudes that is both generic in shape (that is, not strongly dependent on the modelling choices) and increasing in influence with redshift. As such, it may be distinguished from intrinsic non-Gaussian properties of SNe Ia (which are assumed not to vary with redshift). We forward model the probability density function (pdf) of lensing $p_{lens}(\Delta m)$ conditioned on the primordial power spectrum amplitude A_s , the matter density Ω_m (which is also constrained by the SN Ia Hubble diagram) and Amod. This pdf is then convolved with the intrinsic and observational noise of SN Ia data to generate a theoretical distribution of Hubble diagram residuals that may be compared to the data. We will impose priors from the CMB for A_s and Ω_m and marginalise over them, fixing the spectral index $n_s = 0.9665$ (Planck Collaboration 2020a) and the sum of neutrino masses $\Sigma m_{\nu} = 0.06$ eV (we discuss the sensitivity of our results to priors in Appendix A2). Our result is a posterior pdf for A_{mod} .

Our paper is organised as follows. In Section 2 we describe the construction of our model. In Section 3 we briefly describe the data, and in Section 4, we present the results of our analysis, and discuss them in Section 5.

By linear scales, we mean k < 0.1h Mpc⁻¹, and we distinguish non-linear scales as intermediate: $0.1 < k < 10^2 h$ Mpc⁻¹, and small: $k > 10^2 h$ Mpc⁻¹. The reduced Hubble constant is $h = H_0/100$ km sec⁻¹ Mpc⁻¹. We set c = 1 everywhere.

2 THEORY

2.1 What does the weak lensing of SNe Ia constrain?

The parameter σ_8 is an integral measure of the amplitude of the linear power spectrum¹, and therefore directly relatable to the amplitude A_s and shape of the primordial spectrum.

We express the change in magnitude Δm due to lensing of a given SN Ia as relative to a homogeneous universe of the same average matter density. Δm can then be positive (the LOS passes through an over-dense region) or negative (through a void). To linear order, averaging over sources we have $\langle \Delta m \rangle = 0$ as gravitational lensing conserves photons. The square of the dispersion $\sigma_{\Delta m}$ of Δm over many sources located at comoving distance χ_s may be written as an integral over the power spectrum (Frieman 1996):

$$\sigma_{\Delta m}^{2} = 9\pi (0.4 \log 10)^{2} \Omega_{\rm m}^{2} H_{0}^{4} \times \int_{0}^{\chi_{s}} d\chi \frac{\chi^{2} (\chi_{s} - \chi)^{2}}{\chi_{s}^{2}} (1 + z(\chi))^{2} \int_{0}^{k_{\rm max}} dk \frac{\Delta^{2}(k, z)}{k^{2}} , \quad (5)$$

where $\Delta^2(k, z) = k^3 P(k, z)/2\pi^2$ is the dimensionless matter power spectrum, and we have chosen to cut off the scale at k_{max} . The prefactors of (0.4 log 10) arise as $\sigma_{\Delta m}$ is the dispersion in magnitudes rather than flux amplification. The value of this integral depends considerably on the power spectrum at small scales.

To illustrate this for the case of baryon feedback, we select a range of models from the HMCODE2020² emulation package³. We choose three options: 1) linear only, 2) the CDM-only non-linear model of Mead et al. (2020), and 3) the baryon feedback model of the same with $T_{AGN} = 8.0$. While HMCODE2020 is not designed to extrapolate to the k_{max} relevant for us, it nevertheless provides a useful illustration of uncertainty on small scales. To this, we add power spectrum data⁴ compiled from hydrodynamical simulations by van Daalen et al. (2020). The simulations we select are the Cosmic-OWLS suite (Le Brun et al. 2014) and the BAHAMAS suite (McCarthy et al. 2017), which encompass a reasonable range of outcomes. We plot the ratio of the baryon feedback power spectra at z = 0 to the relevant CDM-only reference model in Figure 1. The figure shows power is suppressed up to scales of $k \sim 30h$ Mpc $^{-1}$, then strongly enhanced with considerable difference between models. Equivalent figures for non-standard dark matter models and neutrino masses can be found in Euclid Collaboration (2024, 2025).

In Figure 2 we plot $\sigma_{\Delta m}$ as a function of the small scale cutoff k_{max} for the same selection of models. There are considerable differences between models: while the dark matter only model has little sensitivity to scales $k > 10^2 h$ Mpc $^{-1}$, results from baryonic feedback or neutrino prescriptions vary considerably. The apparent runaway behaviour of the model of Mead et al. (2020) is due to a flattening of the power spectrum at $k > 10^2 h$ Mpc $^{-1}$ because the stellar component is represented as a delta function (the authors make no claims their model can be extrapolated beyond k sim 20). The simulation



Figure 1. The suppression or enhancement of the power spectrum compared to the dark matter only model of Mead et al. (2020). The models used are HMCODE2020 with $T_{AGN} = 8.0$, the Cosmic-OWLS hydrodynamical simulations (Le Brun et al. 2014), and the BAHAMAS hydrodynamical simulations (McCarthy et al. 2017). Power is suppressed on scales 0.1 < k/h < 30 Mpc⁻¹ by AGN and supernovae feedback prescriptions, which differ from model to model. At scales k/h > 30 Mpc⁻¹ power is enhanced due to condensation from baryonic cooling, although again the extent of this depends considerably on the model.

results start to diverge at k > 250h Mpc⁻¹. This is likely due to the softening lengths (r = 4 kpc for Cosmic-OWLS and BAHAMAS) and particle masses ($4 \times 10^9 - 7 \times 10^8 M_{\odot}$) used in them which lead to spurious clumping on small scales. We estimate that to obtain adequate predictions of SN Ia lensing from N-body or hydrodynamical simulations would require particle masses of $m_p \sim 10^7 M_{\odot}$ and softening lengths of 1 kpc (corresponding to $k \sim 10^3 h$ Mpc⁻¹). This applies either to estimation from the power spectrum or directly through ray-tracing. We conclude that a reliable a-priori calculation of $\sigma_{\Delta m}$ is at present unavailable, and we emphasize that we do not make use of any of these models or Eqn. 5 in our analysis.

It is also clear from Fig. 2 that SN Ia weak lensing is insensitive to the linear regime and associated cosmological parameters such as σ_8 . Indeed, linear-scale correlation is undetectable with current datasets (Shah et al. 2024b). In this paper, we will fix the linear scales using a prior on A_s from the CMB, and use SNe Ia to constrain the non-linear empirical amplitude A_{mod} .

Before continuing, we note that the existence of compact objects (CO, for example primordial black holes, if they exist) produces an enhancement in $\sigma_{\Delta m}$ if they form a significant fraction $\alpha = \Omega_{\rm CO}/\Omega_{\rm m}$ of the matter density. In Shah et al. (2024a), it was shown that $\alpha < 0.12$ at 95% credibility, and $\alpha = 0$ was preferred (by the Bayes ratio of model probabilities) to $\alpha > 0$. Hence, while some compact objects (e.g. stars) certainly exist, they do not make a measurable contribution to SN Ia lensing. We therefore neglect them in this analysis.

2.2 SN Ia weak lensing as a function of A_{mod}

In the halo model description, matter power is the sum of linear low density-contrast perturbations, plus gravitationally bound high density-contrast haloes that have collapsed and virialised (Kaiser 1984). This is also taken as the starting point for the codes used in weak lensing shear analyses such as HMCODE2020, which adjusts the theoretical halo-model power spectrum empirically to better fit simulations (Mead et al. 2020). We write our power spectrum model

¹ The historical definition of σ_8 as the dispersion of density fluctuations in spheres of $8h^{-1}$ Mpc - as proxied by galaxy counts - was motivated by the earliest galaxy surveys. It was soon understood that this was difficult to express theoretically: fluctuations on this scale are enhanced by non-linear gravitational evolution and differ by the type of galaxy being counted due to galaxy bias. By re-defining σ_8 in terms of the *linear* power spectrum, the modern definition effectively refers to structure on scales larger than $8h^{-1}$ Mpc.

² https://github.com/alexander-mead/HMcode

³ As implemented in CAMB, https://github.com/cmbant/CAMB

⁴ https://powerlib.strw.leidenuniv.nl



Figure 2. The dispersion of lensing magnification derived from Eqn. 5 as a function of the integral cutoff k_{max} with the lens at $z_l = 0.5$ and the source at $z_s = 1.0$, for a range of models and simulations of the matter power spectrum. The linear power spectrum is shown in yellow, the model of Mead et al. (2020) is shown without baryon feedback in magenta and with feedback in black Dotted and dashed magenta lines show the model adjusted for two choices of A_{mod} , as defined in Equation 3. Results from power spectra compiled from hydrodynamical simulations and published in van Daalen et al. (2020), for a selction of parameters, are shown in red and blue. It is clear that the value of $\sigma_{\Delta m}$ is sensitive to the power spectrum on intermediate and small scales. In particular, while baryonic feedback suppresses lensing on intermediate scales, it enhances it on the small scales relevant to SN Ia. It is not clear that all of the models converge at small scales: in the case of Mead et al. (2020) we have extrapolated the emulator far beyond the region it was designed to model. In the case of simulation data, it is likely that the softening length and particle size produce spurious additional power on small scales. This graph is for illustration only and does not form part of our analysis.

as denoted by subscripts L and H respectively:

$$P(k,z) = P_{\rm L}(k,z) + A_{\rm mod} P_{\rm H}(k,z) \quad . \tag{6}$$

 $P_{\rm L}$ is the same as that in Eqn. 3, while $P_{\rm H}$ is the contribution solely due to halos, and equivalent to the term $P_{\rm NL} - P_{\rm L}$ in Eqn. 3. For $P_{\rm H}$ we adopt the calibration of Sheth et al. (2001) refitted by Courtin et al. (2011), which describes the abundance and spectrum of haloes in a purely CDM universe very well over a huge range of halo masses (Zheng et al. 2024). The calibration of Sheth et al. (2001) was also used as the starting point for power spectrum emulator models such as Smith et al. (2003) and successors. We take the haloes to have the NFW profile described in Navarro et al. (1997), as it was shown in Shah et al. (2024b) that the NFW model is consistent with observations of SN Ia lensing. Although in general the profiles will not be spherically symmetric, this is an accurate approximation after averaging over many LOS (Mandelbaum et al. 2005).

Our aim is to model the statistics of magnitude fluctuations of SNe Ia in terms of A_{mod} . Before we proceed to the details below, we note that whilst our A_{mod} is equivalent in formulation to that of Amon & Efstathiou (2022), important practical differences arise when applied to SNe Ia data. The integral of Eqn. 5 does not define a window function of scale other than limits imposed by the size of the source and lens, and other statistics of SN Ia lensing operate in the same fashion. This is distinct from the scales introduced by spatial correlations functions of transverse separation used in galaxy surveys, which may then be used to set scale cuts. Fig. 2 indicates

that for intermediate scales, there is a broad expectation that baryonic feedback will result in $A_{\rm mod} < 1$ (the models considered generate less lensing than the CDM-only model of Mead et al. 2020). Conversely, continuing to small scales relevant for SN Ia lensing, it is likely that baryons alone would lead to $A_{\rm mod} > 1$ (more lensing is observed than expected in a CDM-only universe), although we noted above that the resolution limits of the simulations may generate spurious power enhancement on these scales. Alternatively, warm dark matter (if it exists) and neutrino masses are expected to suppress power on small scales, reducing $A_{\rm mod}$. We therefore have no a-priori expectations of whether the $A_{\rm mod,S}$ we measure from SNe Ia will indicate power suppression or enhancement.

Following the procedure of Zumalacárregui & Seljak (2018), revised in Shah et al. (2024a), we model the full-shape of the lensing pdf⁵ as a function of cosmological parameters, intrinsic skew and A_{mod} . We write the lensing pdf as a convolution of lensing due to linear scales and haloes as

$$p_{\text{lens}}(\Delta m) = p_{\text{L}}(\Delta m; A_s, \Omega_{\text{m}}, z) * p_{\text{H}}(\Delta m; A_s, \Omega_{\text{m}}, A_{\text{mod}}, z) \quad , \quad (7)$$

where * denotes the convolution operation. The 1-point distribution of weak lensing convergence on linear scales has been shown to be well-approximated by a log-normal distribution (Clerkin et al. 2017). We therefore take $p_{\rm L}$ as a log-normal distribution of zero mean and dispersion $\sigma_{\rm L}$ obtained from Eqn. (5) using the linear power spectrum. We obtain $p_{\rm H}$ from TurboGL⁶ (Kainulainen & Marra 2009, 2011) which uses semi-analytic integration to accurately model lensing by dark matter haloes (and thus avoids the resolution issues inherent in N-body or hydrodynamical simulations). The minimum halo mass has been set to be $10^7 M_{\odot}$. $A_{\rm mod}$ is then a simple scale parameter on this pdf such that $\sigma_{\rm H}(A_{\rm mod}) = A_{\rm mod}\sigma_{\rm H}(A_{\rm mod} = 1)$.

Intrinsic skew of SN Ia residuals may in principle be confused with lensing. However, while lensing skew is redshift-dependent, intrinsic skew is presumed not to be. We parametrise the intrinsic dispersion of SN Ia by the sin-arcsin distribution family (Jones & Pewsey 2009) where δ , ϵ of this family capture both skew and kurtosis with $\delta = 1$, $\epsilon = 0$ being a normal distribution. The distribution is defined as

$$p_{\text{Int}}(\Delta m) = \frac{1}{\sqrt{2\pi}E} \delta \sqrt{1 + x^2} \exp(-x^2/2) / \sqrt{1 + \Delta m^2} \quad , \tag{8}$$

where

$$x = (\sinh(\delta \operatorname{arcsinh}(\Delta m) - \epsilon) - D)/E \quad , \tag{9}$$

and the location and scale parameters D, E are determined by the constraints

$$\int \Delta m p_{\text{Int}}(\Delta m) d\Delta m = 0$$

$$\int \Delta m^2 p_{\text{Int}}(\Delta m) d\Delta m = \sigma_i^2 \quad .$$
(10)

Here $\sigma_i^2 = C_{ii}$ is the diagonal of the SN Ia covariance matrix (see Section 2.3 below), which is the statistical uncertainty in the SN Ia distance modulus. Convolving this with the lensing pdf gives

$$p_{\rm res}(\Delta m) = p_{\rm lens} * p_{\rm Int}$$
, (11)

with Δm now the Hubble diagram residual

$$\Delta m = m_i - \mathcal{M} - \mu_{i,\text{theory}}(\theta) \quad . \tag{12}$$

⁵ We avoid using methods based on higher order moments of the pdf (such as described in Marra et al. 2013; Quartin et al. 2014), which are vulnerable to the presence of outliers.

⁶ https://github.com/valerio-marra/turboGL

Here $\mathcal{M} = M - 5 \log_{10} H_0$ is a degenerate combination of the Hubble constant H_0 and the fiducial SN Ia absolute magnitude M. The distance modulus is $\mu_{i,\text{theory}} = 5 \log_{10}(D_L(z_i, \Omega_M)H_0/c) + 25$ and we take D_L as the usual homogeneous cosmology luminosity distance in a Flat Λ CDM model.

2.3 The SN Ia likelihood adjusted for lensing

Denoting the model parameters collectively as θ , the SN Ia likelihood for a homogeneous cosmology using the DES SN Ia Year-5 sample (DES-SN5YR) is assigned to be the Gaussian

$$\ln \mathcal{L}_G(\Delta m|\theta) = -\frac{1}{2} \sum_{i,j} \Delta m_i C_{ij}^{-1} \Delta m_j$$
(13)

where C is the covariance matrix which is the sum of systematic and statistical errors (Vincenzi et al. 2024). In the presence of non-Gaussian lensing, we adjust this to

$$\log \mathcal{L} = \ln \mathcal{L}_G + \left(\sum_i \log p_{\text{res}} - \sum_i \log p_{\text{diag}}\right) \quad , \tag{14}$$

with

$$p_{\text{diag}}(\Delta m_i) = \frac{1}{\sqrt{2\pi\sigma_i}} \exp(-\frac{1}{2}(\Delta m_i/\sigma_i)^2) \quad , \tag{15}$$

where $\sigma_i^2 = C_{ii}$ as per Eqn. 10 and p_{res} is defined in Eqn. 11. The term in brackets of Eqn. 14 adjusts the likelihood of each individual SN Ia for the difference between a skewed (either by lensing or intrinsic skew) residual probability and a Gaussian probability. The first term retains covariance, which is important for correct error estimation. As explained in Shah et al. (2024a), this likelihood reduces correctly when there is no lensing or intrinsic skew as in this case $p_{res} = p_{diag}$, and also when the covariance matrix is diagonal as in this case $\log \mathcal{L}_G = \sum_i \log p_{diag}$. The key point is *C* is only weakly nondiagonal, so we expect our assignment to be accurate. We discuss the validation of this likelihood in Appendix A1.

We pre-compute the lensing pdfs for a grid of redshifts, cosmological parameters and $A_{\rm mod}$, and interpolate the log probabilities. The evaluation of the likelihood takes ~ 0.3 seconds for ~ 1,500 SN Ia on a typical laptop. We fix the optical depth of reionisation as $\tau = 0.0561$ and the power spectrum slope as $n_s = 0.9665$, and set the neutrino mass $m_{\nu} = 0.06$ eV. We take uniform priors as $A_{\rm mod} \in (0.2, 2.5)$ (motivated by consideration of Fig. 2), $\epsilon \in (-0.2, 0.2)$ and $\delta \in (0.6, 1.4)$. Runs are performed using Polychord⁷ (Handley et al. 2015), and plots and analysis are made using Anesthetic⁸ (Handley 2019).

3 DATA

3.1 SN Ia

We use the DES-SN5YR dataset as described in Sánchez et al. (2024), but with the modification not to exclude SNe Ia that are more than 4σ away from the best-fit Hubble diagram. Removing this cut avoids biasing our results by arbitrarily truncating the skewed and extended distribution of residuals that lensing produces. The SN Ia survey was conducted in four regions of the DES footprint with a total of ten fields, and the SN Ia range from 0.01 < z < 1.13. Supernova candidates are analysed using machine-learning classifiers (Möller & de Boissière 2020; Möller et al. 2022; Qu et al. 2021) whose inputs are the light curve observations, and whose output is the probability of being an SN Ia. The diagonal of the covariance is then adjusted for this probability, down-weighting likely contaminants but not discarding them altogether. The SN Ia redshift is set to be the spectroscopic redshift of the galaxy that is closest in directional light radius to the SN Ia (Sullivan et al. 2006; Qu et al. 2024).

There are 1,930 SN in the initial sample, of a similar redshift distribution to the original DES-SN5YR dataset, and we cut these to include only those between 0.2 < z < 1.0 in our analysis. The lower cut is because lensing will not materially affect low redshift SN Ia, and the lower redshift SN Ia are from older, heterogeneous surveys with uncertain selection functions. The upper cut is to reduce potential uncertainties due to larger bias corrections at high redshifts (for example, see Figure 7 of Vincenzi et al. 2024). We additionally cut likely contaminants or poorly measured SN Ia, by excluding data with $\sigma_m > 1.0$ mag and p(SN Ia) < 0.9. Our data input to the lensing likelihood therefore comprises 1,484 SN Ia of average redshift $z \sim 0.47$.

3.2 CMB

As input priors to H_0 , Ω_m and linear scales A_s , we use chains derived from the Python implementation of Planck's 2015 Plik_lite (Prince & Dunkley 2019) which may be found on the repository of the DES-SN5YR data⁹. These chains have somewhat wider constraints than the fiducial results of Planck Collaboration (2020a), but are perfectly adequate for our results.

4 RESULTS

0 60

Marginalising over all other parameters, and quoting the median, 16% and 84% quantiles, we find

$$A_{\text{mod},S} = 0.77^{+0.69}_{-0.40} \tag{16}$$

which is consistent within 1σ of the CDM-only value of $A_{\text{mod}} = 1$, and we have used the subscript S to denote our result is derived primarily from small scales. We show the triangle plot for the marginalised posterior and pair distributions in Figure 3.

There are hints the data prefers lower values with the maximum of the full posterior (as determined by a kernel density estimate) at $A_{\text{mod},\text{S}} = 0.30$, and the highest-density 68% credible interval is $A_{\text{mod},\text{S}} < 1.09$. We expect this upper bound to be conservative, as Figure 3 shows our posterior is truncated by our prior and it is likely that more probability mass might be found below our prior range rather than above it. Conversely, our credibile interval around the median will be somewhat over-optimistic for the same reason.

Notably, our method demonstrates that SN Ia can distinguish the effects of the linear and non-linear parts of the power spectrum, as S_8 and $A_{\text{mod},\text{S}}$ show little correlation in their posteriors. The credible intervals for the intrinsic skew parameters $\epsilon = -0.07^{+0.04}_{-0.04}$ and $\delta = 0.91^{+0.06}_{-0.05}$ are moderately discrepant from the values (0, 1) of a Gaussian distribution, but at no great significance. There is little covariance between intrinsic skew and matter-power parameters.

Our maximum likelihood improves the fit over the Gaussian likelihood by $\delta\chi^2 = -5.6$. Subtracting out the fit improvement due to the

⁷ https://github.com/PolyChord/PolyChordLite

⁸ https://github.com/handley-lab/anesthetic

introduction of intrinsic skew, we find $A_{\text{mod},\text{S}} = 0.3$ improves the fit by $\delta \chi^2 = -1.2$ compared to $A_{\text{mod},\text{S}} = 1$.

We describe our estimation of systematics in Appendix A, with the dominant contribution being the assignment of the likelihood. While there is some inaccuracy in our posterior, examining the expected coverage probability we see that the 68% highest-density credible interval we quote above is likely to be conservative (over and above the effect of truncation by the prior noted above) by $\delta A_{\text{mod},\text{S}} \sim 0.19$. Other credible intervals of interest have systematics of a similar amount or smaller. We have tested the effect of the redshift and SN Ia probability cuts used for our data and find they are small. Our total systematic error estimate is $\delta A_{\text{mod},\text{S}} \sim 0.2$. Therefore, we judge that systematics are smaller than our statistical error by a factor of at least two.

The literature has to date quoted results in terms of σ_8 which (ignoring for now the caveats we noted in Section 2) this be interpreted as ~ $A_{\text{mod,S}}\sigma_{8,\text{CMB}}$. The first measurement was done by Castro & Quartin (2014), who found $\sigma_8 = 0.84^{+0.28}_{-0.65}$ using 706 SN Ia from the Joint Lightcurve (JLA) catalogue (Betoule et al. 2014). Re-analysing the JLA data using a model that also incorporated SN Ia peculiar velocities (via their redshifts), Macaulay et al. (2017) found similar results but also that the systematics in Castro & Quartin (2014) were underestimated. In Macaulay et al. (2020), 196 SN Ia from the DES Year 3 release were used in the same methodology to find $\sigma_8 = 1.2^{+0.9}_{-0.8}$. Calibrating a halo model using the observed correlation between SN Ia residuals and foreground galaxy positions, Shah et al. (2024b) found $\sigma_8 = 0.9 \pm 0.13$ by comparing the dispersion $\sigma_{\Delta m}$ of lensing in this model along random lines of sight to a fitting formula given in Marra et al. (2013). In general, although these results do not marginalise over cosmological parameters and intrinsic skew as we have done, they all indicate a weak preference for $A_{\text{mod,S}} > 1$. However, they are consistent with our results within 1σ .

5 CONCLUSIONS

In this paper, we have shown that the non-Gaussian distribution of residuals of SNe Ia to the Hubble diagram induced by weak gravitational lensing carries statistical information about the matter power spectrum on scales k > 1h Mpc⁻¹. In particular, SNe Ia data provides access to scales that cannot be probed by galaxy-sized sources. Due to the theoretical uncertainties of modelling the power spectrum on these scales, we have constrained an empirical parameter $A_{\text{mod},S}$, which describes the suppression (or enhancement) of matter power on small scales compared to a benchmark of a dark matter only Λ CDM universe with cosmological parameters derived from the CMB. We find hints of suppression with $A_{\text{mod},S} = 0.77^{+0.69}_{-0.40}$ (median and 68% credible interval), with the posterior peaking at low values of $A_{\text{mod},S}$.

While our results are weaker than those of Preston et al. (2023) (quoted as Eqn. 4 here), who find almost 3σ preference for $A_{\text{mod}} < 1$, they are independent of data and the typical systematics of galaxy shear surveys such as photometric redshifts, shape blending and intrinsic alignments. By contrast, our systematics arise primarily from assignment of the likelihood.

Our results are not precise enough to distinguish between competing models of dark matter or baryonic physics. The pathway to doing so lies in improvements to statistical error, systematics and theoretical modelling. Regarding statistical error, in Quartin et al. (2014), it was forecast that a sample of 3,000 SN Ia from DES would be able to constrain σ_8 to within ~ 35%; this is consistent with our ~ 55% constraint for A_{mod} using 1,484 SN Ia. The authors also forecast that

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a sample of 500,000 SN Ia from the Rubin LSST survey would result in a constraint of ~ 3%. In this case, the statistical error would be below the systematics, and the current likelihood assignment would need to be improved. One way this could be achieved is the use of simulation-based inference.

Finally, regarding theoretical modelling it would be desirable to interpret our results in the context of parameters of physical processes, which requires extending these models to small scales. As noted in Section 2, the apparent non-convergence of the baryonic models illustrated in Fig. 2 is likely due to an extrapolation of emulators beyond the scales they were trained on, because of the resolution limits of hydrodynamical simulations. This is in principle surmountable: promising work has been done on "nesting" dark matter only simulations to increase resolution (Wang et al. 2020; Zheng et al. 2024) arbitrarily, and it seems plausible that this methodology could be applied to other simulations. Even if this were not possible, we note that certain ranges of A_{mod} would provide severe constraints on the underlying physics: for example, $A_{\text{mod}} \leq 0.5$ is not expected from any of the models considered here with cold dark matter as the dominant component.

We conclude that for future datasets, SN Ia offer a unique window into the power spectrum on small scales, and the pathway to improve the control of systematics and theoretical modelling is clear.

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P.S. devised the project, performed the analysis and drafted the manuscript; M.V. re-analysed the DES-SN5YR data with cuts removed; T.M.D., R.C., D.H., L.G., N.J., J.L., C.L., M.Su., L.W. advised on the analysis and commented on the manuscript; P.A. assisted with the Pippin pipeline used for the simulations. T.M.D. and M.Su. were also internal reviewers and R.M. was the final reader. The remaining authors have made contributions to this paper that include, but are not limited to, the construction of DECam and other aspects of collecting the data; data processing and calibration; developing broadly used methods, codes, and simulations; running the pipelines and validation tests; and promoting the science analysis.

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Figure 3. A triangle plot of the posteriors for relevant model parameters, with the medians and 68% quantiles shown along the diagonal. The constraints on S_8 and Ω_m arise from the fit of the CMB power spectrum and SN Ia luminosity distances. The CMB priors we use are shown in red. As noted in the text, the Planck-lite-py likelihood used to generate the chains has moderately wider constraints than the full likelihood used in Planck Collaboration (2020a), however this does not impact our results. In contrast, the constraint on $A_{mod,S}$ arises from the detail of the distribution of SN Ia residuals around the mean and is not degenerate with S_8 . ϵ and δ represent intrinsic skew and kurtosis of the distribution; it is evident from the posterior that there is little degeneracy between this and lensing. For comparison, we also plot a combination (in green) of the DES 3x2pt analysis of Abbott et al. (2022) and BAO measurements from the Sloan Digital Sky Survey (SDSS), as summarised in Alam et al. (2021). We remind the reader that we have defined $A_{mod} = 1$ as small-scale power that is compatible with the growth of CDM-only fluctuations in a Flat Λ CDM universe with cosmological parameters derived from the CMB.

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Software: numpy (Harris et al. 2020), CAMB (Lewis & Bridle 2002), matplotlib (Hunter 2007), scipy (Virtanen et al. 2020), SNANA (Kessler et al. 2009), Pippin (Hinton & Brout 2020), Polychord (Handley et al. 2015), Anesthetic (Handley 2019), PlikLite Python implementation (Prince & Dunkley 2019), HMCODE2020 (Mead et al. 2020), TurboGL (Kainulainen & Marra 2009)

DATA AVAILABILITY

The data and Python code used to generate the results and plots in this paper are available on reasonable request from the authors. The lensing pdfs used will be made available upon publication at https://github.com/paulshah/SNLensing.

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APPENDIX A: SYSTEMATIC ERROR ESTIMATION

A1 The lensing likelihood

Although we have given plausible reasons for our assignment of the likelihood, doubts may still arise on its validity. One may worry about the ability of the sin/arcsin distribution to accurately represent the non-Gaussianity of intrinsic effects and residual contamination by non SN Ia. Also, bias corrections adjust the mean of the residual distribution in bins to account for selection effects.

The Bayesian framework of estimated credible intervals allows us to exploit rigorous theorems to establish whether a posterior or specified credible interval is accurate. We note that while the homogeneous Gaussian SN Ia likelihood \mathcal{L}_G has been extensively tested against simulations (see for example Camilleri et al. 2024), even this commonly-used likelihood remains (in the sense we explain below) unproven.

Constraints on parameters are normally quoted as credible intervals, defined to be the range in which the integrated probability mass takes the specified value. A credible interval is located arbitrarily: common choices are to center it on the mean, the median, or define it as the region of highest probability density (such that the size of the interval in parameter space is minimized). An inference pipeline such as the one we have described in this paper is a process to assign an *estimated* posterior distribution $\hat{p}(\theta|d)$ to model parameters θ given the data *d*. Intuitively, one may anticipate that if the estimated posterior correctly reproduces *all* of the credible intervals (with arbitrary location) of the true distribution *p*, then the posterior estimation is correct and $\hat{p} = p$. This was made rigorous in Lemos et al. (2023), who also give efficient methods for the computation.

Their method can be summarised as follows:

• choose a credibile level $1 - \alpha$, a proposal function g for the centre of the credibile interval θ_r , and a distance metric on the space of model parameters $s(\theta_1, \theta_2)$

• generate $i = 1 \dots N_{sim}$ simulations of data vectors d_i drawing truth model parameters θ_i^* from the prior

• for each simulation *i*, construct the estimated posterior \hat{p}_i and draw $j = 1 \dots N_{\text{sample}}$ samples θ_{ij} from it

• the coverage probability for each simulation is $f_i = (1/N_{\text{sample}}) \sum_j \mathcal{I}[s(\theta_{ij}, \theta_r) < s(\theta_{i,*}, \theta_r)]$ where \mathcal{I} is the indicator function

• the expected coverage probability (ECP) for the estimated posterior method is then ECP = $(1/N_{sim}) \sum_i I(f_i < 1 - \alpha)$

If the ECP = $1 - \alpha$ for all α and proposal functions *g*, then Theorem 3 of Lemos et al. (2023) states that $\hat{p} = p$. In our context, this procedure represents a comparison of our analytic likelihood to an implicitly-defined simulation-based likelihood that is deemed to be the truth.

We adopt three choices for g. We validate the regions of highest probability density (HPD) of the 1-d marginalised distribution of A_{mod} , a uniform $g = \mathcal{U}(0.2, 2.5)$, and a normal $g = \mathcal{N}(0.75, 0.5)$ (i.e. in the vicinity of the median). The distance metric is $s(\theta_1, \theta_2) =$ $|\theta_1 - \theta_2|$. We generate 240 simulations using the SNANA software (Kessler et al. 2009) with pipeline coordinator PIPPIN (Hinton & Brout 2020), 10 for each uniformly spaced A_{mod} value between the limits of the prior $A_{\text{mod}} \in (0.2, 2.5)$. The simulations mimic the observing conditions and selection functions of the DES SN Ia survey, and are processed using the same pipeline as used for the real data, including adjustments for bias corrections. For each simulation, we sample 10^4 times from the posterior.

Figure A11 shows the results. Although the HPD credible intervals (our preferred statistic) are all under-confident, the likelihood shows



Figure A1. Graphs of the expected coverage probability (ECP) (blue) against the specified credible interval for three choices of the credible interval location proposal function g. The x-axis is the integrated probability mass within the credible interval, and the y-axis are the results of the validation procedure outlined in the text above. The correct case is the black dashed line. An under-confident likelihood would result in the blue line above the black dashed line: the truthful credible interval level $1 - \alpha$ is larger than the desired input. Conversely, an over-confident likelihood would lie under. A biased likelihood would mix under-confidence in some intervals with over-confidence elsewhere. *Upper panel*. The estimation of the highest-density credible interval is under-confident by between 0% – 20%. *Middle panel*. The intervals concentrated around the median are over-confident by up to ~ 7%. *Lower panel*. The intervals randomly located across the prior show the likelihood is biased to a similar degree, confirming the results of the upper two panels.

evidence of inaccuracy at the ~ 10% level, and credible intervals in the vicinity of the median are somewhat over-confident. We can approximately convert this inaccuracy in the credible interval to a systematic error in A_{mod} in the following way : for a given location and desired credible interval $1 - \alpha$, interpolate the ECP until it matches $1 - \alpha$, and then re-evaluate the statistic at the credible interval matching this ECP. For example, for the 68% HDR constraint, we see the ECP (the y-axis on the upper panel of Figure A1) that matches this corresponds to a ~ 59% credible interval (the x-axis) for our estimated posterior. Re-evaluating the posterior with this revised interval leads to a constraint of $A_{\text{mod}} < 0.90$, indicating our result was conservative by $\Delta A_{\text{mod}} \sim 0.19$. Evaluations of other intervals at such intermediate levels of credibility lead to similar, or smaller, systematic error estimates.

We therefore assign $\Delta A_{\text{mod}} = 0.19$ as the systematic error due to our likelihood assignment.

A2 Cosmological priors

It has been noted in the literature that assigning CMB priors to the amplitude of linear fluctuations can result in galaxy shear analyses preferring greater power spectrum suppression compared to adopting a wider prior (see Figure 6 of García-García et al. (2024) and Figure 11 of Terasawa et al. (2024)). This is likely due the combined influence of linear and intermediate scales on galaxy shear data, which we do not expect to occur for SN Ia lensing constraints (see Figure 2). To check this, we have repeated our analysis using priors derived from combining the DES 3x2pt (Abbott et al. 2022) and baryon acoustic oscillations derived from the Sloan Digital Sky Survey (SDSS) (as summarized in Alam et al. 2021) (but retaining the Planck constraint on n_s). Note, we do not alter the reference benchmark where $A_{mod} = 1$ refers to fluctuations on all scales being consistent with the CMB in dark-matter only ΛCDM . This would represent a change of our baseline unit rather than any physical effect, and inhibit comparison of the values derived here to each other or the literature.

The change of prior affects our $A_{\rm mod}$ mean value at the level of $\Delta A_{\rm mod} \sim 0.01$. This is confirmed by Figure 3, as the posteriors for large and small scale fluctuation amplitudes are largely independent. Additionally varying n_s within the constraints allowed by the CMB results in a similar very small variation. We do not consider variations in the sum of neutrino masses as this is one of the physical effects expected to be captured in the parameter $A_{\rm mod}$.

A3 Variation between TurboGL and power spectrum emulators

We have used TurboGL to construct our lensing pdf, and this forms our benchmark with respect to our $A_{\text{mod},\text{S}}$ result is obtained. Nevertheless, for the purposes of comparing with literature results, it is worthwhile to examine differences in lensing predictions from TurboGL to power spectrum emulators like HMCODE2020.

Following the arguments of Marra et al. (2013), we examine the second moment of the lensing pdf. This directly relates to the A_{mod} parameter, and the shape of the lensing pdf is broadly universal, so the higher moments that drive our constraints scale in proportion to the standard deviation. We assign error budgets to the following potential inaccuracies : 1) the use of linearisation in weak lensing 2) the halo mass function calibration of Sheth et al. (2001) 3) the halo concentration relation of Zhao et al. (2009) 4) uncertainties in the emulation of the power spectrum.

The first three were determined in Marra et al. (2013) as 5%, 3%



Figure A2. The standard deviation of the lensing pdf (in magnitudes) by redshift. The literature models are as described in Mead et al. (2020, 2016) and Takahashi et al. (2012) respectively (coloured lines), and the calculation of $\sigma_{\Delta m}$ is made with Equation 5 with $k_{\text{max}} = 10^3$. We compare this to the pdf we calculate using TurboGL (black line).

and 3% respectively and we adopt these estimates. For the last, we plot the variation in $\sigma_{\Delta m}$ (as determined from Equation 5 for $k_{\text{max}} = 10^3$) for the power spectrum models of Mead et al. (2020, 2016); Takahashi et al. (2012) alongside our results derived from TurboGL in Figure A2. We adopt the median variation of 6% as our systematic error for this component. Adding these errors in quadrature gives a total of 9%.

A4 Summary

To recap, we estimate systematic errors of 0.19 for the likelihood, 0.01 for cosmological priors, and 0.09 for our lensing pdf. These add in quadrature to a total systematic error of $\Delta A_{mod} = 0.21$.

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