Testing dark energy paradigms with weak gravitational lensing

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Any theory invoked to explain cosmic acceleration predicts consistency relations between the expansion history, structure growth, and all related observables. Currently there exist high-quality measurements of the expansion history from Type Ia supernovae, the cosmic microwave background temperature and polarization spectra, and baryon acoustic oscillations. We can use constraints from these datasets to predict what future probes of structure growth should observe. We apply this method to predict what range of cosmic shear power spectra would be expected if we lived in a ΛCDM universe, with or without spatial curvature, and what results would be inconsistent and therefore falsify the model. Though predictions are relaxed if one allows for an arbitrary quintessence equation of state \( -1 \leq w(z) \leq 1 \), we find that any observation that rules out ΛCDM due to excess lensing will also rule out all quintessence models, with or without early dark energy. We further explore how uncertainties in the nonlinear matter power spectrum, e.g. from approximate fitting formulas such as Halofit, warm dark matter, or baryons, impact these limits.

I. INTRODUCTION

Consistency relations between cosmological observables exist for any underlying physical model class [1,2]. This means that the combination of observables pertaining to the expansion history and those pertaining to structure growth could potentially falsify a given dark energy paradigm, such as the standard ΛCDM model of cold dark matter and a cosmological constant with Gaussian initial conditions, or smooth dark energy models with equation of state parameter \(-1 \leq w(z) \leq 1\), known as quintessence models. For instance, even a single massive high-redshift cluster could falsify all ΛCDM and quintessence models, if its mass falls significantly outside of what we predict based on Type Ia supernovae (SNe), the cosmic microwave background (CMB), baryon acoustic oscillations (BAO), and the local measurement of the Hubble constant \((H_0)\) [3].

Weak gravitational lensing, whereby distant galaxy images are distorted due to the gravitational effects of matter lying along the line of sight, is another key probe of cosmic structure growth, with promising results in recent years, e.g. [4–8]. See Refs. [9–11] for recent reviews. However, current cosmological constraints are very limited; since we do not know the intrinsic shapes of galaxies, weak lensing is by its very nature a statistical measure and thus the power of a survey is directly related to its sky coverage [12–13]. Given future large-area surveys on the horizon, for instance from the ground with the Dark Energy Survey (DES) [14], and the Large Synoptic Survey Telescope (LSST) [15] or from space with Euclid [16] and the Wide-Field Infrared Survey Telescope (WFIRST) [17], it is particularly timely to determine our expectations [18].

Using the wealth of data in hand from distance measures and the CMB, we expand upon Ref. [3] to predict these weak lensing observables within the ΛCDM and quintessence paradigms. We focus on the shear power spectrum and show how the predictions relax as we generalize the model beyond flat ΛCDM and allow for curvature, an arbitrary dark energy equation of state, and early dark energy (EDE). Once the upcoming large-area weak lensing surveys are completed, we can compare their results to these predictions and possibly falsify the ΛCDM model or perhaps the entire quintessence paradigm. In this way weak lensing could provide a smoking gun for new physics such as primordial non-Gaussianity, phantom dark energy \((w(z) < -1)\), subhorizon dark energy clustering, an interacting dark sector, or even a modification to general relativity.

This paper is organized as follows. In [II] we describe our methodology, including the datasets we use, the parameter spaces we explore, our Markov Chain Monte Carlo (MCMC) likelihood analysis, and how we compute the ingredients needed to predict the weak lensing shear power spectrum and two point correlation functions (2PCFs) for source redshift distributions typical of current and future surveys. In [III] we present our weak lensing predictions and discuss how they depend on source redshift and dark energy model class. In [IV] we explore uncertainties related to the matter power spectrum and SN light-curve fitting. We discuss and conclude in [V].

II. METHODOLOGY

Our methodology is based on that of Refs. [1–3]. Briefly, we take current CMB constraints on the initial power spectrum plus current data related to the overall geometry and expansion history of the Universe, determine parameter constraints within a class of dark energy models using an MCMC analysis, and then compute the weak lensing predictions which can be used to falsify the class. The observational data we use come from local dis-
tance measures of $H_0$, the Type Ia SNe distance-redshift relation, BAO distance measures, and the CMB temperature and polarization power spectra.

### A. Data sets

The Type Ia SN sample we use is the compilation of 288 SNe from Ref. [19], consisting of data from the first season of the Sloan Digital Sky Survey-II (SDSS-II) Supernova Survey, the ESSENCE survey [20], the Supernova Legacy Survey [21], Hubble Space Telescope SN observations [22], and a collection of nearby SN data [23]. The light curves of these SNe have been uniformly analyzed by [19] using both the MLCS2k2 [23] and SALT2 [21] methods. We use the SALT2 method for our default analysis but discuss the impact of switching to MLCS2k2 in [1VD].

For the CMB, we use the most recent, 7-year release of data from the WMAP satellite (WMAP7) [24] employing a modified version of the likelihood code available at the LAMBDA web site [25] which is substantially faster than the standard version while remaining sufficiently accurate [27]. We compute the CMB angular power spectra using the code CAMB [29, 30] modified to include models with general dark energy module [31, 32] to include models with general dark energy equation of state evolution where $w(z)$ may cross $w = -1$.

We use the BAO constraints from Ref. [33], which combines data from SDSS and the 2-degree Field Galaxy Redshift Survey that determine the ratio of the sound horizon at last scattering to the quantity $D_V(z) = [zD(z)/H(z)]^{1/3}$ at redshifts $z = 0.2$ and $z = 0.35$. Since these constraints actually come from galaxies spread over a range of redshifts, and our most general dark energy model classes allow the possibility of high frequency variations in the expansion rate $H(z)$ across this range, we implement the constraints by taking the volume average of $D_V$ over $0.1 < z < 0.26$ (for $z = 0.2$) and $0.2 < z < 0.45$ (for $z = 0.35$).

Finally, we include the recent Hubble constant measurement from the SHOES team [34], based on SN distances at 0.023 < $z$ < 0.1 that are linked to a maser-determined absolute distance using Cepheids observed in both the maser galaxy and nearby galaxies hosting Type Ia SNe. The SHOES measurement determines the absolute distance to a mean SN redshift of $z = 0.04$ which we implement as $D(z = 0.04) = 0.04c/(74.2 \pm 3.6$ km s$^{-1}$ Mpc$^{-1}$).

### B. Parameter constraints

For a given parameter set $\theta$ that defines the cosmological model class, we use a modified version of the CosmoMC code [35, 36] to sample from the joint posterior distribution,

$$P(\theta|x) = \frac{L(x|\theta)P(\theta)}{\int d\theta L(x|\theta)P(\theta)},$$

where $L(x|\theta)$ is the likelihood of the dataset $x$ given the model parameters $\theta$ and $P(\theta)$ is the prior probability density. We use the same priors as in [3], namely flat priors that are wide enough to not limit the MCMC constraints from the aforementioned datasets. The most restrictive parameter class we will consider is that of a flat $\Lambda$CDM model,

$$\theta_0 = \{\Omega_b h^2, \Omega_c h^2, \tau, \theta_A, n_s, \ln A_s\},$$

where $\Omega_b h^2$ and $\Omega_c h^2$ are the present physical baryon and cold dark matter densities relative to the critical density, $\tau$ is the reionization optical depth, $\theta_A$ is the angular size of the acoustic scale at last scattering, $n_s$ is the spectral index of the power spectrum of initial fluctuations, and $A_s$ is the amplitude of the initial curvature power spectrum at $k_p = 0.05$ Mpc$^{-1}$. All other parameters, including the Hubble constant $H_0 = 100 \ h$ km/s/Mpc, the present total matter density $\Omega_m$ and dark energy density $\Omega_{DE}$, and the amplitude of the matter power spectrum today $\sigma_8$, can be derived from this basic set.

We generalize this basic model, and consequently expand the aforementioned parameter set, in several ways. The first is to allow the spatial curvature $\Omega_K = 1 - \Omega_{tot}$ to vary. We further allow for quintessence by decomposing the dark energy equation of state into principal components (PCs) for $z < 1.7$,

$$w(z) + 1 = \sum_{i=1}^{N_{max}} \alpha_i e_i(z),$$

where the equation of state at higher redshift is a constant $w(z > 1.7) = w_{\infty}$, the $e_i$ are the PC amplitudes, the $e_i(z)$ are the PC functions, and $N_{max} = 10$ has been found to provide a complete basis for our purposes here. The PCs we use here are the eigenfunctions of the Fisher matrix for the combination of CMB constraints from Planck and WFIRST-like SN data, as described in Ref. [1]. We will fix $w_{\infty} = -1$ for the model classes which exclude EDE, as a constant equation of state dark energy component quickly becomes negligible at early times, and allow it to vary otherwise with a flat prior on $e^{-1}$ between $e^{-1}$ and 1. A detailed discussion of this EDE approach and its relation to CMB constraints can be found in Appendix B of Ref. [3].

Quintessence models describe dark energy as a scalar field with kinetic and potential contributions to energy and pressure. Barring models where large kinetic and (negative) potential contributions cancel (e.g., [27]), quintessence equations of state are restricted to $-1 \leq w(z) \leq 1$. Following Ref. [1], this bound is conservatively implemented with uncorrelated top-hat priors on the PC amplitudes $\alpha_i$. Any combination of PC amplitudes that is rejected by these priors must arise from an
equation of state $w(z)$ that violates the bound on $w(z)$, but not all models that are allowed by the priors strictly satisfy this bound.

In summary, the full set of parameters we will consider here are

$$ \theta = \{ \theta_0, \Omega_K, \alpha_1, \ldots, \alpha_{10}, e^{w_*} \} , \quad (4) $$

where we will look at models of increasing generality by exploring increasingly larger subsets of this parameter set.

C. Observables

Once we have sampled the joint posterior distribution of the cosmological parameters, we can then compute the posterior probabilities for any derived statistic, in particular cosmic shear observables. As an intermediate step, it is useful to first consider the two basic ingredients that the two-point shear observables are constructed from: the comoving angular diameter distance $D$ and the nonlinear matter power spectrum $\Delta^2_{NL}$. The former is related to the cosmological parameters through

$$ D(\chi) = \frac{1}{\sqrt{\Omega_K H_0^2}} S_K \left[ \chi \sqrt{\Omega_K H_0^2} \right] , \quad (5) $$

where $S_K(x)$ equals $x$ in a flat universe ($\Omega_K = 0$), sinh $x$ in an open universe ($\Omega_K > 0$), and sin $x$ in a closed universe ($\Omega_K < 0$). Here the comoving radial coordinate is

$$ \chi(z) = \int_0^z \frac{dz'}{H(z')} , \quad (6) $$

where the Hubble expansion rate is

$$ H(z) = H_0 \left[ \Omega_m(1+z)^3 + \Omega_{\text{DE}}f(z) + \Omega_K(1+z)^2 \right]^{1/2} , \quad (7) $$

with

$$ f(z) = \exp \left[ 3 \int_0^z \frac{dz'}{1+z'} \frac{1+w(z')}{1+z'} \right] , \quad (8) $$

and the contribution from radiation is assumed to be negligible.

The linear matter power spectrum depends on the initial matter power spectrum and the growth function of linear density perturbations $\delta \propto G a$, where $a$ is the scale factor. Given the smooth dark energy paradigm, the growth function is given by

$$ G'' + \left( 4 + \frac{H'}{H} \right) G' + \left[ 3 + \frac{H'}{H} - \frac{3 \Omega_m H_0^2 (1+z)^3}{2 H^2(z)} \right] G = 0 , \quad (9) $$

where primes denote derivatives with respect to ln $a$ and the function is normalized so that $G = 1$ in the matter dominated limit. We then compute the linear power spectrum at redshift $z$ by rescaling the $z = 0$ spectrum $\Delta^2_{NL}(k; 0)$ (computed using CAMB with the PPF dark energy module) by the growth function:

$$ \Delta^2_{NL}(k; z) = \left[ \frac{G(z)}{(1+z)G(0)} \right]^2 \Delta^2_{NL}(k; 0) . \quad (10) $$

We shall see that current CMB constraints on $\Omega_m h^2$ still allow substantial variation in the matter radiation equality scale and hence the shape of the linear power spectrum.

Cosmic shear observables however depend on the full nonlinear power spectrum. We compute this quantity using the Halofit fitting function [38], which we will now describe. The halo model decomposes the nonlinear power spectrum into the sum of two contributions,

$$ \Delta^2_{NL}(k) = \Delta^2_{Q}(k) + \Delta^2_{H}(k) , \quad (11) $$

where $\Delta^2_{Q}(k)$ is the “two-halo” term, which encapsulates quasi-linear power from large-scale halo placement, and $\Delta^2_{H}(k)$ is the “one-halo” term, which arises due to correlations within the haloes themselves. The Halofit fitting functions depend on parameters based on Gaussian filtering of the linear power spectrum with variance

$$ \sigma^2(R) \equiv \int \Delta^2_{NL}(k) \exp \left( -k^2 R^2 \right) \ d \ln k : \quad (12) $$

the nonlinear scale $k_{NL}$ defined such that $\sigma_k(k_{NL}^{-1}) = 1$, the effective spectral index,

$$ n \equiv -3 - \frac{d \ln \sigma^2(R)}{d \ln R} \bigg|_{\sigma = 1} , \quad (13) $$

and the spectral curvature,

$$ C \equiv - \frac{d^2 \ln \sigma^2(R)}{d \ln R^2} \bigg|_{\sigma = 1} . \quad (14) $$

The nonlinear power spectrum is parameterized by a set of coefficients $(a_n, b_n, c_n, \gamma_n, \alpha_n, \beta_n, \mu_n, \nu_n)$ which are allowed to vary as a function of the aforementioned spectral index and curvature so as to fit N-body simulation data. In terms of these coefficients,

$$ \Delta^2_{Q}(k) = \Delta^2_{Q}(k) \left\{ \frac{[1 + \Delta^2_{H}(k)]^{\beta_n}}{1 + \alpha_n \Delta^2_{H}(k)} \right\} \exp [-f(y)] , \quad (15) $$

where $y \equiv k/k_{NL}$ and $f(y) = y/4 + y^2/8$ is a function introduced to truncate at high $k$, and

$$ \Delta^2_{H}(k) = \frac{\Delta^2_{H}(k)}{1 + \mu_n y^{-1} + \nu_n y^{-2}} , \quad (16) $$

with

$$ \Delta^2_{H}(k) = \frac{a_n y^{3 f_1(\Omega_m)}}{1 + b_n y f_2(\Omega_m) + c_n f_3(\Omega_m) y^{4 - \gamma_n}} . \quad (17) $$
The coefficients are
\[
\begin{align*}
\log_{10} a_n &= 1.4861 + 1.8369n + 1.6762n^2 + 0.7940n^3 + 0.1670n^4 - 0.6206C , \\
\log_{10} b_n &= 0.9463 + 0.9466n + 0.3084n^2 - 0.9400C , \\
\log_{10} c_n &= -0.2807 + 0.6669n + 0.3214n^2 - 0.0793C , \\
\gamma_n &= 0.8649 + 0.2989n + 0.1631C , \\
\alpha_n &= 1.3884 + 0.3700n - 0.1452n^2 , \\
\beta_n &= 0.8291 + 0.9854n + 0.3401n^2 , \\
\log_{10} \mu_n &= -3.5442 + 0.1908n , \\
\log_{10} \nu_n &= 0.9589 + 1.2857n .
\end{align*}
\]

The functions \( f_1, f_2, f_3 \) are power laws, with the exponents fit to N-body data for either matter-only open models
\[
\begin{align*}
f_1(\Omega_m) &= \Omega_m^{-0.0732} , \\
f_2(\Omega_m) &= \Omega_m^{0.1423} , \\
f_3(\Omega_m) &= \Omega_m^{0.0725} 
\end{align*}
\]
or flat \( \Lambda \)CDM models
\[
\begin{align*}
f_1(\Omega_m) &= \Omega_m^{-0.0307} , \\
f_2(\Omega_m) &= \Omega_m^{0.0585} , \\
f_3(\Omega_m) &= \Omega_m^{0.0743} 
\end{align*}
\]

The efficiency factor is weighted according to the source distribution in a given survey, \( n(z) \), normalized such that \( \int_0^\infty n(z)dz = 1 \). For this paper we will use the simple model
\[
n(z) \propto \left( \frac{z}{z_0} \right)^{\alpha} \exp \left[ -\left( \frac{z}{z_0} \right)^\beta \right] ,
\]
with parameters \( z_0, \alpha, \) and \( \beta \). This parameterization is similar to what has been used in both the COSMOS \cite{6} and CFHTLS \cite{43} analyses, with \( (z_0, \alpha, \beta) = (0.894, 2.0, 1.5) \) for COSMOS and \( (z_0, \alpha, \beta) = (0.555, 1.197, 1.193) \) for CFHTLS. This leads to approximate median redshifts of 1.3 and 0.8, respectively, for our simplified COSMOS and CFHTLS surveys. Note that these distributions closely approximate the ones expected for WFIRST \cite{13, 15} and DES. In the work that follows we will plot results for both of these simple source distributions, but will specialize to CFHTLS (as in \cite{46}) when results for the two are similar.

Finally, we also compute the 2PCFs
\[
\xi_{+/-}(\theta) = \frac{1}{2\pi} \int_0^\infty dl \int J_0(l\theta)P_n(l)
\]
of the shear components from the power spectrum. Being the real-space correlation and the easier quantity to measure on small scales, the 2PCFs are better suited to compare against current measurements than \( P_n \) itself.

### III. Weak Lensing Predictions

Current distance constraints are highly predictive for growth of structure statistics such as the weak lensing power spectrum in the flat \( \Lambda \)CDM context. Since this is the standard model of cosmology and our baseline case, we shall use it to illustrate the steps in our chain of inference in \[III A\]. We then discuss observations that would falsify \( \Lambda \)CDM in favor of quintessence and those that would falsify the whole quintessence class of \(-1 \leq w(z) \leq 1\) smooth dark energy models in \[III B\].

#### A. \( \Lambda \)CDM

In Fig. 1 we start with the growth function prediction. Unless stated otherwise, henceforth shaded regions will correspond to 68% confidence level (CL) regions and curves will bound 95% CL regions defined to have equal posterior probabilities in the two tails. We will often plot results as fractional differences from the prediction for the maximum likelihood (ML) model. The growth predictions are most precise at high redshifts, but even
at $z = 0$ the allowed range in the growth function is only 1–2% [2].

As described in the previous section, we can combine the growth function with the initial power spectrum and transfer function to predict the linear matter power spectrum at $z = 0$; see the upper and lower panels of Fig. 2. For plotting purposes we follow the usual convention of taking

$$P(k) = \left(\frac{2\pi^2}{k^3}\right)\Delta^2(k)$$

with $k$ defined in $h\text{Mpc}^{-1}$ for both linear and nonlinear power spectra. Here the predictions carry $\sim 10\%$ errors in spite of the precise growth results. The dominant source of uncertainty is from the measurements of the matter density and shape of the initial power spectrum from WMAP7. In particular, uncertainties in $\Omega_m h^2$ correspond to shifts in matter-radiation equality which cause left-right shifts in the power spectrum. There are also contributions from tilt ($n_s$) uncertainties that pivot the spectrum around the best constrained WMAP scale of $k \sim 0.02 \text{Mpc}^{-1} \approx 0.03 h \text{Mpc}^{-1}$. Both of these types of uncertainties should be reduced by a factor of a few with Planck CMB data and allow the full precision of the growth predictions to be utilized (see IV A and Fig. 7).

Also shown in Fig. 2 is the full nonlinear matter power spectrum as computed with Halofit (grey hatched curve in the upper panel), along with its deviation from the ML prediction in the middle panel. For the nonlinear power spectrum, the uncertainties are the same as the linear one for $k < k_{NL} \sim 0.3 \text{h Mpc}^{-1}$. For smaller scales, the nonlinear power spectrum uncertainties no longer reflect the linear uncertainties. Whereas the tilt $n_s$ makes the linear uncertainties continue to grow larger, the nonlinear ones become saturated reflecting the fixed nature of the one-halo piece of the Halofit prescription. We shall see that uncertainties here are dominated by the accuracy of Halofit and the ability of adjustments in its parameters to model baryonic physics (see IV B).

Uncertainties in $P_{NL}(k,z)$ propagate into the shear statistics. We show the shear power spectrum as computed from Eq. (22) in Fig. 2 for both the COSMOS and CFHTLS source distributions. The confidence intervals mainly reflect those of the nonlinear matter power spectrum. The interval is actually slightly narrower at low $l$ than at low $k$ in Fig. 2. Significant contributions to $P_\kappa$ come from $z > 0.5$, where both linear growth function uncertainties diminish and the linear power spectrum of the gravitational potential is better fixed by the CMB (in the relevant units of Mpc$^{-1}$ rather than in $h \text{Mpc}^{-1}$). Likewise, at a fixed angular scale CFHTLS predictions tend to be slightly weaker than those from COSMOS due to the lower source redshifts typical of ground-based surveys.

We can make this redshift dependence explicit by replacing the realistic redshift distributions with idealized single source planes. Figure 4 shows the 95% CL region widths for sources at $z = 0.5$, 2, and 3.5. Note that the well-constrained pivot in the power spectrum projects to higher multipole at higher redshift leading to tighter predictions at a fixed multipole. Given the tighter predictions, high redshift cosmic shear measurements provide an interesting opportunity to falsify the flat ΛCDM model.
FIG. 3. Upper panel: Flat ΛCDM predictions for the shear power spectrum, showing the 68% and 95% CL regions as in Fig. 1 for COSMOS (upper, grey hatched) and CFHTLS (lower, blue solid). Lower panel: CFHTLS shear power spectrum prediction plotted with respect to the ML flat ΛCDM model.

In Fig. 5 we show the 2PCF $\xi_+$ which is more useful for comparison with the relevant observations from COSMOS [8] and CFHTLS [43]. The displayed 1σ error bars are computed from the full covariance matrices estimated for each survey, as described in Refs. [8, 43]. The predicted range of flat ΛCDM models appears to be consistent with the observations. However, the error bars at different angular scales are heavily correlated, and therefore do not represent the actual uncertainty at any individual scale. Further note that these small-volume surveys are not well-suited for making statements for or against ruling out the ΛCDM model; COSMOS results use a 1.64 deg$^2$ field containing 76 galaxies per arcmin$^2$, and CFHTLS results use 22 deg$^2$ containing 12 galaxies per arcmin$^2$.

If future observations falsify these predictions, then one would need to generalize the cosmological model class. The next simplest class of models retains Λ as the dark energy but allows for non-vanishing spatial curvature $\Omega_K$ in the ΛCDM context. We find a minimal error increase in this class, as shown in the upper panel of Fig. 6. This is because spatial curvature is well constrained in the ΛCDM paradigm. Thus a measurement that falsifies the flat ΛCDM model would also falsify the ΛCDM assumption itself, indicating that the dark sector is more complicated.

FIG. 4. Single source plane, 95% CL full-width extent for the shear power spectrum $\Delta P_{\kappa} / P_{\kappa}$ (as plotted in the lower panel of Fig. 3) as a function of $l$, for sources at $z = 0.5$ (top), 2 (middle), and 3.5 (bottom).

FIG. 5. Flat ΛCDM predictions for $\xi_+$, showing the 68% and 95% CL regions as in Fig. 3 for both the COSMOS (grey hatched) and CFHTLS (blue solid) source redshift distributions. Also shown are the data points from Refs. [8] (black points) and [43] (magenta points) with 1σ error bars. Note that the error bars at different angular separations are correlated.

B. Quintessence

Measurements of shear observables outside the bounds shown in the previous subsection would be in statistical conflict with ΛCDM. Barring systematic errors and un-
known astrophysical effects, some of which we address in §IV, such a measurement would indicate that the true cosmology belongs to a wider class of models.

If we relax the equation of state to allow all 10 PC amplitudes to vary but revert to the flatness assumption, we find that the contours shift toward lower power as shown in the middle panel of Fig. 6. Nevertheless, the ML quintessence model is quite similar to the flat ΛCDM ML model and is near the 68% CL upper limit of the quintessence model class. This is an artifact of the parameter priors: given the freedom in \( w(z) \) there are many ways to reduce the growth in ways unconstrained by distance measures (see Fig. 3 of [2]). The converse is not true as there are no models with good likelihood values in the quintessence class with more power than already allowed in flat ΛCDM, an effect also seen in Ref. [3]. The effective data-driven upper bound on shear statistics for this quintessence class remains that of flat ΛCDM in spite of the downward shift in posterior contours.

The lower panel of Fig. 6 overlays the results for flat quintessence with \( w_\infty = -1 \), and quintessence with both curvature and \( w_\infty \) allowed to vary (the latter generalization potentially allowing for EDE). This additional freedom further reduces growth, in accordance with the results of Ref. [3]. This is one of our main results, namely that generalizing from ΛCDM to quintessence models only serves to reduce cosmic shear. We will show below in §IV that warm dark matter could also only explain a power deficit: likewise, massive neutrinos would only decrease power. It is possible for baryonic effects to increase power in either ΛCDM or quintessence, but only at high multipoles. In general, observations which rule out the ΛCDM model by finding a shear excess that cannot be explained by astrophysical uncertainties also falsify the entire quintessence paradigm.

IVA. Linear Matter Power Spectrum

Much of the residual statistical uncertainty in the flat ΛCDM predictions comes from uncertainties in the linear matter power spectrum from WMAP7. The main sources are the matter density \( \Omega_m h^2 \), tilt \( n_s \) and normalization \( \ln A_s \). These uncertainties should soon be improved by Planck. We show how the predictions change from importance sampling to Planck-like priors on the parameters \( \Omega_b h^2, \Omega_m h^2, n_s, \) and \( \ln A_s \), using the DETF projected covariance matrix [47] and assuming the mean corresponds to the ML flat ΛCDM model.

In Fig. 7 we show the improvement in the fractional constraints (grey hatched curves) on the linear power spectrum (upper panel) and on the shear power spectrum of the CFHTLS source redshift distribution (lower panel). One typically should expect Planck to improve the precision of predictions by more than a factor of 2 to the level of ±3–5%, which will greatly enhance the prospects for falsifying the standard ΛCDM paradigm.

IV. SYSTEMATIC AND OTHER UNCERTAINTIES

We now turn to exploring various factors that can impact the statistical predictions shown in the previous section. These uncertainties largely stem from uncertainties in the matter power spectrum and systematic uncertainties in the SNe data. In this section we will discuss the impact of an improved initial linear matter power spectrum from Planck ([IV A]), uncertainties in the nonlinear matter power spectrum from small-scale physics and the Halofit fitting function ([IV B]), tomography ([IV C]), and supernova light-curve fitting ([IV D]).

A. Linear Matter Power Spectrum

1 Massive neutrinos would also shift parameter values due to their effect on the CMB, so accurately quantifying the suppression of the predicted shear power requires including neutrino mass as a parameter in the initial MCMC analysis, which we leave for future work.
FIG. 7. Expected improvements from Planck (grey hatched) over WMAP7 (blue solid) for flat ΛCDM predictions. Upper panel shows the linear matter power spectrum and the lower panel shows the shear power spectrum for the CFHTLS source redshift distribution, with the 68% and 95% CL regions relative to the ML flat ΛCDM model.

B. Nonlinear Matter Power Spectrum

We now explore the impact of a few sources of uncertainty in the nonlinear matter power spectrum on our predictions for the shear power spectrum — the Halofit fitting function \[38\], warm dark matter \[48\], and baryonic physics \[49, 50\]. We will find it most convenient to parameterize the latter two (physics-based) uncertainties in terms of the former. For these studies we employ our CFHTLS source redshift distribution and the flat ΛCDM model class.

Halofit itself has been found to be only accurate at up to the 5–10% level in the nonlinear regime for the ΛCDM cosmological models for which it was designed \[40–42\], even with the correction factor Eq. \((21)\). While these inaccuracies are smaller than our current statistical errors in the same regime, they are comparable to the expected Planck errors.

Furthermore, we have employed the ΛCDM-calibrated Halofit parameters in our predictions for the more general quintessence \(w(z)\) classes.

In Fig. 8 we show once again our flat ΛCDM predictions for the CFHTLS shear power spectrum where in the upper panel we show what happens when \(c_n\) is amplified and in the lower panel we show what happens when \(a_n\) is amplified. In each case we are only altering the nonlinear matter power spectrum, and thus only the high multipole regime of the shear power spectrum. In the case of amplifying \(a_n\), we are simply enhancing the one-halo term. We conclude that \(a_n\) and \(c_n\) must be calibrated in ΛCDM to the ∼10% level to make use of current statistical predictions for \(\ell < 1000\) whereas in quintessence models with EDE and curvature creating large power deficits the tolerances can be up to double that.

While Halofit was constructed to fit N-body simulations with cold dark matter and no baryonic effects, we find that in particular \(c_n\) is a useful parameter for describing typical deviations from these assumptions.

Warm dark matter (WDM) reduces the small-scale matter power spectrum in a way that increases with decreasing WDM particle mass. By using high-resolution N-body/hydrodynamic simulations, Ref. \[48\] constructs a fitting function to describe the modified matter power spectrum \(P_{\Lambda WDM}(k)\) in terms of that of the corresponding ΛCDM model \(P_{\Lambda CD M}(k)\),

\[
P_{\Lambda WDM}(k) = \left[1 + (ak)^{1.8}\right]^{-2/15} P_{\Lambda CD M}(k),
\]
where

\[ \alpha(m_{\text{WDM}}, z) = 0.0476 \left( \frac{1 \text{ keV}}{m_{\text{WDM}}} \right)^{1.85} \left( 1 + \frac{z}{2} \right)^{1.3} \]  

(27)

in units of \( h^{-1} \text{ Mpc} \) and \( m_{\text{WDM}} \) is the mass of the warm dark matter particle. This fitting function is accurate at the \(~2\%\) level below \( z = 3 \) and for masses \( m_{\text{WDM}} > 0.5 \) keV. On the other hand, this form assumes a fixed set of \( \Lambda \text{CDM} \) parameters associated with the simulations and thus effects should be taken as illustrative of the magnitude of the effect. We find that the WDM effects corresponding to 0.5 keV dark matter can be mimicked by the Halofit parameter \( c_n \) with \( c_n \to 1.01c_n \) (Figure 9 upper panel). We thus see that, even for this fairly extreme particle mass, warm dark matter contributes negligibly to the error budget for this particular statistic, in agreement with Ref. [18].

Baryonic effects, such as radiative heating and cooling, star formation, and supernova and AGN feedback, affect our predictions through their impact on the small-scale matter power spectrum. Given the inherent difficulties in adding baryons to large-scale structure simulations, the degree to which these various processes impact structure growth remains highly uncertain. However we can make a conservative assessment by using the most extreme example, namely the results of Ref. [50] which found a significant decrease in power ranging from \((z = 0)\) 1% at \( k \approx 0.3 h \text{ Mpc}^{-1} \) to 10% at \( k \approx 1 h \text{ Mpc}^{-1} \) and 30% at \( k \approx 10 h \text{ Mpc}^{-1} \). To find the rough impact on the nonlinear matter power spectrum, we interpolate Fig. 8 of Ref. [47]. Again we find that the Halofit \( c_n \) accurately mimics this effect with \( c_n \to 1.15c_n \) corresponding to the simulations of Ref. [50] which include AGN feedback (Figure 9 lower panel). A more detailed study of the effect on weak lensing observations is presented in Ref. [51].

Thus current uncertainties in baryonic physics at their most extreme are as large as current statistical ranges in the predictions. Hence the modeling of baryonic effects on structure formation must improve if we are to make use of future constraints above \( l = 10^8 \), as found in Refs. [49, 50]. Note also that, although the particular model we have shown here suppresses scale-scale power, baryons can have the opposite effect of potentially the same magnitude, by increasing small-scale power via cooling [49]. In the future, one approach to accounting for these systematic errors is to marginalize \( a_n \) and \( c_n \) as model parameters given a prior range appropriate to the state-of-the-art simulations including baryonic and other effects.

C. Tomography and Ratio Tests

One means of reducing both the statistical and systematic uncertainty is to employ tomographic probes that compare the shear at different source redshifts [52]. Uncertainties remaining from the primordial power spectrum discussed in IV A largely drop out and some of the effects of baryonic physics may be “self-calibrated” through jointly determining the concentration-mass relation [49, 53].

As a simple demonstration and a proxy for tomography, we can define the statistic \( f(l) \) (not to be confused with \( f(z) \) from Eq. 8) for a given model class as the ratio of the shear power spectrum \( P_\kappa(l) \) predicted for \( \Omega_m \text{CDM} \) (median redshift of 1.3) to that predicted for CFHTLS (median redshift of 0.8). The resulting prediction is shown in Fig. 10 for our most restrictive (flat \( \Lambda \text{CDM} \), upper panel) and most general (non-flat quintessence with EDE, lower panel) model classes, where now the \(~1\sigma\) uncertainties are reduced to the 1% level for the flat \( \Lambda \text{CDM} \) model. This sharp prediction offers another opportunity to falsify flat \( \Lambda \text{CDM} \) if systematics in the measurement and ratio prediction can be made comparably precise. In the lower panel of Fig. 10 the features at small scales are related to the asymmetry of the quintessence growth predictions and the difference in the nonlinear scales for the source distributions of COSMOS (\( l \gtrsim 200 \)) and CFHTLS (\( l \gtrsim 500 \)). However, present uncertainties in the nonlinear power spectrum arising from Halofit render predictions in the nonlinear regime unreliable. In the upper panel of Fig. 10 we show how the uncertainties in \( c_n \) (green, upper curves)
and $a_n$ (magenta, lower curves), respectively, affect the flat $\Lambda$CDM predictions for $f(l)$. These uncertainties affect the quintessence predictions similarly. We see that $\sim 10\%$ level calibration of these Halofit parameters is not sufficient to exploit the sub-percent level $\Lambda$CDM predictions in the nonlinear regime. On the other hand, observed deviations in $f(l)$ of $\gtrsim 3\%$ would falsify $\Lambda$CDM and $\gtrsim 10\%$ our most general quintessence class even considering these uncertainties.

D. Supernova Light-Curve Fitting

The analyses of the datasets outlined in §II A contribute further systematic uncertainties in our predictions, the largest of which arise from the fitting of SN light-curves. The two most widely used light-curve fitters are SALT2 and MLCS2k2. Assuming a flat cosmological model with a constant dark energy equation of state $w$, the SALT2 and MLCS2k2 methods yield $(\Omega_m, w) = (0.26 \pm 0.03, -0.96 \pm 0.13)$ and $(\Omega_m, w) = (0.31 \pm 0.03, -0.76 \pm 0.13)$, respectively [19]. Thus these two methods lead to a discrepancy in the dark energy equation of state of $\Delta w \sim 0.2$. This might be due to a possible mismatch in the UV spectra between low-redshift and intermediate-redshift SNe [53] and efforts are underway to reduce the resulting error. We have chosen here to use SN data analyzed using the SALT2 technique, but we now briefly discuss what happens when we instead use the MLCS2k2 data.

Even for the flat $\Lambda$CDM model class, the differences between SALT2 and MLCS2k2 estimates of $\Omega_m$ and $w$ significantly alter our predictions for the shear power spectrum, as can be seen in the upper panel of Fig. 11. In the lower panel we show the comparison for flat quintessence; in both cases, the CL regions shift by $\sim 0.5–1\sigma$. This shift to higher shear with MLCS2k2 is driven mainly by the preference for higher $\Omega_m$ when using this method; this higher $\Omega_m$ also increases the present day matter power spectrum normalization for a fixed amplitude of initial curvature fluctuations $A_s$. The corresponding effect for cluster abundance was found in Ref. [3]. With improvements in the calibration of SN data and modeling of the effects of dust extinction, it is likely that these systematics can be reduced so that they will not be a limiting uncertainty for future predictions of shear statistics.

Note that Ref. [3] made the opposite choice, using the MLCS2k2 SN constraints for their main results.
V. DISCUSSION

In this paper we have provided robust statistical predictions for weak lensing observables in a variety of cosmological contexts. Given existing local distance measures of $H_0$, the Type Ia SNe distance-redshift relation, BAO distance measures, and the CMB temperature and polarization power spectra, we have constrained the expected cosmic shear power spectrum in the $\Lambda$CDM model. We then generalized this model class to show how the predictions widen when curvature, EDE, and a generalized dark energy equation of state are added.

In a similar fashion as in the clusters study of Ref. [3], we find that any observation that claims to rule out the $\Lambda$CDM model via a shear excess generically also rules out the entire quintessence paradigm, as the extra freedom allowed by the free function $w(z) \geq -1$ can only serve to reduce the relative growth if we normalize to the CMB and constrain the distance-redshift relation to the CMB and SNe at opposite ends of the expansion history. Adding EDE, warm dark matter, massive neutrinos, or baryonic AGN feedback tends to exacerbate the reduction, as they too only reduce growth and therefore suppress the cosmic shear power spectrum. Therefore a measured shear excess would require updating our models to include significant baryonic cooling, primordial non-Gaussianity, non-smooth dark energy, phantom dark energy, an interacting dark sector, or even a modification to the gravitational force law.

In our analysis we have focused only on those uncertainties pertaining to the predictions for the shear statistics presented here, as opposed to uncertainties in the measurement of the shear itself. The latter becomes important when comparing the data to these predictions. For example in comparing our predictions to current data in Fig. 5 we have employed the statistical and systematic error estimates in the literature. Uncertainties relating to shape measurement, photometric redshifts, and intrinsic alignments feed into the data error covariance. Accounting for these effects, the flat $\Lambda$CDM model is consistent with the current data.

Future imaging surveys, both from the ground (e.g. DES and LSST) and from space (Euclid and/or WFIRST) will provide the necessary number of galaxy shear measurements to enable precise tests of dark energy model classes, as long as the various shear systematics can be either eliminated or understood sufficiently well. Moreover, improved CMB constraints from Planck will soon reduce the uncertainties in the cosmic shear predictions by a factor of a few, further enhancing the ability of weak lensing observations to detect deviations from the concordance cosmological model but also requiring more stringent control on astrophysical systematic errors in both the predictions and the measurements.

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