Cosmological Tests of General Relativity with Future Tomographic Surveys

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<td>As Published</td>
<td><a href="http://dx.doi.org/10.1103/PhysRevLett.103.241301">http://dx.doi.org/10.1103/PhysRevLett.103.241301</a></td>
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<tr>
<td>Publisher</td>
<td>American Physical Society</td>
</tr>
<tr>
<td>Version</td>
<td>Final published version</td>
</tr>
<tr>
<td>Accessed</td>
<td>Mon Oct 15 12:27:25 EDT 2018</td>
</tr>
<tr>
<td>Citable Link</td>
<td><a href="http://hdl.handle.net/1721.1/52501">http://hdl.handle.net/1721.1/52501</a></td>
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Cosmological Tests of General Relativity with Future Tomographic Surveys

Gong-Bo Zhao,1 Levon Pogosian,1 Alessandra Silvestri,2 and Joel Zylberberg3

1Department of Physics, Simon Fraser University, Burnaby, British Columbia, V5A 1S6, Canada
2Kavli Institute for Astrophysics and Space Research, Massachusetts Institute of Technology, Cambridge, Massachusetts 02139, USA
3Department of Physics, University of California, Berkeley, California 94720, USA

(Received 13 May 2009; published 8 December 2009)

Future weak lensing surveys will map the evolution of matter perturbations and gravitational potentials, yielding a new test of general relativity on cosmic scales. They will probe the relations between matter overdensities, local curvature, and the Newtonian potential. These relations can be modified in alternative gravity theories or by the effects of massive neutrinos or exotic dark energy fluids. We introduce two functions of time and scale which account for any such modifications in the linear regime. We use a principal component analysis to find the eigenmodes of these functions that cosmological data will constrain. The number of constrained modes gives a model-independent forecast of how many parameters describing deviations from general relativity could be constrained, along with \( w(z) \). The modes’ scale and time dependence tell us which theoretical models will be better tested.

The observed acceleration of cosmic expansion poses a puzzle for modern cosmology. It may be evidence for dark energy (DE), a component with a negative equation of state, \( w \), making it gravitationally repulsive. It also warrants studying extensions of general relativity (GR) with extra degrees of freedom that can mimic the effects of DE. Modifications to GR are well constrained in dense regions like our Solar System [1]. On larger scales GR is less well tested. Several modifications to GR, capable of producing cosmic acceleration, have been proposed [2]. With the right parameter values, they can match the expansion history of a universe made of cold dark matter (CDM) and a cosmological constant \( \Lambda \)—the observationally favored \( \Lambda \)CDM model [3]. However, their predictions for the growth of structure can differ since the perturbation equations get modified. Future tomographic weak lensing surveys, like the Dark Energy Survey (DES) [4] and the Large Synoptic Survey Telescope (LSST) [5], will measure lensing shear and galaxy counts in many redshift slices (hence the term tomography), thus mapping the evolution of perturbations and offering a new test of GR on cosmological scales [2]. In this work, we use a two-dimensional principal component analysis (PCA) to forecast the constraints on modified growth (MG)—and thus our understanding of gravity—coming from these surveys. Unlike previous MG forecasts, ours is model independent and lets us determine how many parameters describing MG could be constrained, along with the regions in parameter space where we expect the most sensitivity to MG.

We consider linear scalar perturbations to the flat Friedmann-Robertson-Walker metric in Newtonian gauge

\[
ds^2 = -a^2(\eta)[(1 + 2\Psi(\vec{x}, \eta))d\eta^2 - (1 - 2\Phi(\vec{x}, \eta))d\vec{x}^2],
\]

where \( \eta \) is the conformal time and \( a(\eta) \) the scale factor. In Fourier space, one can write [6,7]

\[
\Phi/\Psi = \gamma(k, a), \quad k^2 \Psi = -\mu(k, a)4\pi G a^2 \rho \Delta,
\]

where \( \Delta \) is the comoving matter density perturbation. The function \( \gamma \) describes anisotropic stresses, while \( \mu \) describes a time- and scale-dependent rescaling of Newton’s constant \( G \), as well as the effects of DE clustering or massive neutrinos. In \( \Lambda \)CDM, the anisotropic stress due to radiation is negligible during matter domination, thus \( \gamma = 1 = \mu \). In this Letter, we determine how well the unknown functions \( \gamma \) and \( \mu \) can both be constrained by future data. We also address how well we can detect any departure from \( \gamma = 1 = \mu \) without distinguishing between them.

We consider all the two-point correlations between galaxy counts (GC), weak lensing shear (WL), and cosmic microwave background (CMB) temperature anisotropy, plus the CMB E-mode polarization and its correlation with the CMB temperature. Detailed descriptions of our assumptions for each measurement are found in [8]. GC probe the distribution and growth of matter overdensities, thus giving an estimate of the Newtonian potential \( \Psi \), up to a bias factor. WL is sourced by the sum of the potentials \( \Psi + \Phi \). GC are also affected by “magnification bias” [9], where WL convergence magnifies some faint (thus otherwise undetected) galaxies, adding mild dependence on \( \Psi + \Phi \) to the GC. CMB data probe the integrated Sachs-Wolfe effect (ISW) which depends on \( d(\Phi + \Psi)/d\eta \). Thus, measuring GC and WL over multiple redshift bins, along with CMB data, yields information about the relation between \( \Psi \) and \( \Phi \) and their response to matter density fluctuations. Furthermore, supernovae (SN) redshift-luminosity measurements and CMB constrain the expansion history. For our forecasts, we assume the following probes: Planck [10] for CMB, DES and later LSST for GC and WL, and Joint Dark Energy Mission (JDEM) [11] for SN. To compare to current data, we use
large-scale structure (LSS) data compiled in [12] and the latest SN and CMB data from the “Constitution” SN sample [13] and WMAP5 [3]. For current data, we omit WL but include the GC-CMB cross correlations.

In full generality, we treat $\gamma(k, a)$ and $\mu(k, a)$ as unknown functions and perform a PCA [14,15] to determine how many of their DOFs can be constrained. We use redshift $z = a^{-1} - 1$ as a time variable and pixelize the late-time and large-scale universe ($z \in [0, 30]$, $k \in [10^{-5}, 0.2] h^{-1}$ Mpc$^{-1}$) into $M + 1$ scales and $N$ $k$ bins, with each of the $(M + 1) \times N$ pixels having independent values of $\mu_{ij}$ and $\gamma_{ij}$. We consider $w(z)$ as another unknown function, allowing each of the $M + 1$ $z$ bins to have an independent value $w_i$. Since the surveys we consider will not probe $z > 3$ in detail, we use $M$ bins linear in $z$ for $z \in [0, 3]$ and a single bin for $z \in [3, 30]$. We choose $M = N = 20$ and have checked that this pixelization is fine enough to ensure the convergence of the results. We use logarithmic $k$ bins on superhorizon scales and linear $k$ bins on subhorizon scales to optimize computational efficiency. As in [8], we only consider information from scales well described by linear perturbation theory, a fraction of the total $(k, z)$ volume probed by future surveys. Since the evolution equations [8] contain time derivatives of $\mu$, $\gamma$, and $w$, we follow [16] and use hyperbolic tangent functions to represent steps in these functions in the $z$ direction, while steps in the $k$ direction are left as step functions.

Our parameters are the $\gamma_{ij}$, $\mu_{ij}$, and $w_i$, along with the usual cosmology parameters: energy density of baryons $\Omega_b h^2$ and CDM $\Omega_c h^2$, Hubble constant $h$, optical depth $\tau$, spectral index $n_s$, and amplitude $A_s$. We include one bias parameter per GC $z$ bin and the intrinsic SN magnitude. Thus we have $(M + 1)(2N + 1) + 17 = 878$ parameters in total. For a given set of parameter values, we use MGCAMB [8,17] (a modification of CAMB [18] developed by us to study modified growth) to compute angular spectra for our observables. We generate numerical derivatives of observables with respect to parameters and use the specifications for the experiments to compute the Fisher information matrix, which defines the sensitivity of the experiments to these parameters (see [8] for computational details). Our fiducial values are in all cases $\Lambda$CDM: $\gamma_{ij} = \mu_{ij} = -w_i = 1 \forall i, j$, and the fiducial values of the other parameters are those of WMAP5 [3].

Let us first study the expected errors on $\mu(k, z)$. The error on any $\mu_{ij}$ is large, and the pixels have highly correlated errors. PCA finds the linear combinations of pixels with uncorrelated errors. We take only the $\mu_{ij}$ block of the covariance matrix, thus marginalizing over all other parameters, including the $w_i$ and $\gamma_{ij}$. We invert this block to obtain the Fisher matrix for our $\mu$ values, $F(\mu)$, and diagonalize $F(\mu)$ by writing $F(\mu) = W^T \Lambda W$. The rows of matrix $W$ are the eigenvectors, or the principal components (PCs) [14], while the diagonal elements of $\Lambda$ are the eigenvalues $\lambda_m$. Each eigenvector, $e_\mu(k, z)$, is a linear combination of the original pixels $\mu_{ij}$, forming a surface in $(k, z)$ space. The eigenvectors are orthogonal. We normalize them to unity, rescaling the eigenvalues accordingly. \{ $e_\mu(k, z)$ \} forms an orthonormal basis in which we can expand $\mu$: $\mu(k, z) = \sum_\alpha \lambda_\alpha e_\alpha(k, z)$, where $\alpha_\alpha$ are the new uncorrelated parameters with variances given by $\lambda_\alpha = \langle \sigma^2(\alpha_\alpha) \rangle^{-1}$. We expect, from existing data, that variations in $\mu$ larger than $\mathcal{O}(1)$ are unlikely. We enforce this by applying a prior $\lambda_\alpha > 1$ to $F(\mu)$. This procedure, analogous to the treatment of $w(z)$ in [19], does not affect the well-measured modes but gives a reference point with respect to which we define poorly constrained modes. The worst-measured modes have variances approaching the prior, while those with smaller variances are the well-measured ones. Since we compute the full covariance matrix, then marginalize over all but the parameter(s) of interest, our procedure yields the results that we would get for $\mu$ if we simultaneously measured $w$, $\gamma$, and $\mu$. This analysis can be repeated for $\gamma$ or $w$. Given the PCA results, one can convert uncertainties in the expansion parameters $\alpha_\alpha$ into uncertainties in any other parametrization of $\mu$ ($\gamma$) without recalculating the Fisher matrices [15,16] by projecting the PC Fisher matrix onto a new basis.

Measurements probe combinations of $\Phi$ and $\Psi$, so the effects of $\gamma$, which affects only $\Phi$, are mixed with those of $\mu$, which affects both potentials. This yields degeneracy between $\mu$ and $\gamma$. By varying both, then marginalizing over one, we lose information common to both functions. This is necessary when separately constraining $\mu$ and $\gamma$. While this degeneracy impedes their ability to do so, DES

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**FIG. 1 (color online).** Uncertainties in the eigenmodes for current data and for future data sets including LSST (DES). The two upper panels show modes of $\mu$ and $\gamma$ with mutual marginalization. The lower panel shows uncertainties in the combined modes. The solid lines and the shaded region denote the thresholds $T_1$, $T_2$, and $T_3$. The filled symbols denote the redshift-dependent modes. In the lower panel, the dashed lines show the uncertainties for $\mu$ with $\gamma$ fixed.
and LSST will yield nontrivial constraints on \( \mu \) and \( \gamma \) with mutual marginalization (marginalizing over \( \mu \) when measuring \( \gamma \) and vice versa). The uncertainties associated with PCs of \( \mu \) and \( \gamma \) are shown in the top two panels of Fig. 1 for LSST, DES, and current data. To quantify the sensitivity of the surveys to MG, we introduce three thresholds: well constrained \([T_1, \sigma_\alpha \leq 0.01]\), constrained \([T_2, 0.01 \leq \sigma_\alpha \leq 0.1]\), and informative \([T_3, 0.1 \leq \sigma_\alpha \leq 0.5]\). From Fig. 1, we see that DES could constrain two \( \mu \) parameters and no \( \gamma \) parameters. LSST could constrain many modes, as it will have a superior sky coverage and resolution, wider \( z \) span, and more precise photometric redshift measurements. Current data effectively cannot constrain either \( \mu \) or \( \gamma \). The constraints on \( \mu \) are generally stronger than those on \( \gamma \): \( \mu \) affects GC, WL, and CMB, while \( \gamma \) primarily affects WL and CMB.

Figure 2 shows selected eigenmodes of \( \mu \) and \( \gamma \) for LSST and DES. The \( i \)th \( \mu \) \( \gamma \) mode represents the \( i \)th best-constrained independent \( \mu(k, z) \gamma(k, z) \) surface. Models that predict \( \mu \) and \( \gamma \) similar to our “best” modes, then, will be better constrained. We observe no degeneracy in the \( k \) and \( z \) dependences of the modes. This is counterintuitive, since changing \( \mu \) at some point \( (k, z) \) should have the same impact on the observables as a change at a larger scale but later time. However, since we allow for simultaneous variation of \( \gamma \), the change in \( \mu \) is more efficiently offset by adjusting \( \gamma \), eliminating the \( k-z \) degeneracy. There is a clear pattern to the modes; the best-constrained modes have no \( z \) nodes, but apparent \( k \) nodes, and, approximately, the \( m \)th mode has \( m \) \( k \) nodes. For LSST, at roughly the 10th mode, the first \( z \) node appears, followed by another period of scale-dependent patterns. The alternating \( (k, z) \) patterns repeat until mixed \( (k, z) \) modes appear, after which there are no clear patterns. The best modes are mainly functions of \( k \) and not \( z \). This is partly because the total observable volume in the radial \( z \) direction is limited by the dimming of distant objects and, ultimately, the fact that structures only exist at relatively low \( z \). Also, it is related to us considering only linear perturbations in our analysis, since at small \( z \) the observable volume is too small to fit the small \( k \) modes that are still in the linear regime. Hence, there is more volume available for studying the spatial distribution of structure than the radial distribution. The number of nodes in the \( z \) and \( k \) directions tell us, respectively, the number of \( z \)- and \( k \)-dependent parameters that surveys could constrain. For example, for LSST, there are three clear \( z \)-dependent patterns of \( \mu \) and \( \gamma \) whose eigenvalues fall within \( T_2 \) and \( T_3 \) ranges. For DES, the only constrained modes are those of \( \mu \), and they exhibit only one type of \( z \) dependence, with one node.

The LSST modes for \( \mu \) and \( \gamma \) have similar shapes except that the \( \mu \) modes have a deeper \( z \) span than the \( \gamma \) modes. This is due to an accumulation effect on \( \mu \). On subhorizon scales, the density contrast evolves via \( \Delta + \mathcal{H} \Delta = 4\pi G \mu \rho a^2 \Delta \). Perturbing \( \mu \) at one pixel, e.g., enhancing it at some \( k \) and \( z = 3 \), enhances \( \Delta \) (and thus \( \Psi \)) for that \( k \) mode \( \forall z < 3 \). On the other hand, changes in \( \gamma \) at high \( z \), which primarily affect WL through changes in \( \Phi \) [Eq. (1)], do not affect WL at low \( z \). Hence, the \( z \)-sensitivity range of \( \gamma \) is primarily determined by the redshift range of the WL kernel. The \( z \) dependence of \( \gamma \) also affects the ISW contribution to the CMB at small \( k \) and low \( z \), but its contribution to the Fisher matrix is small due to a large cosmic variance.

In most models of modified gravity, \( \mu \) and/or \( \gamma \) evolve in a time- and scale-dependent way [2]. For example, in scalar-tensor theories, they undergo a steplike transition at the Compton scale of the scalar field. The peaks in the eigensurfaces indicate the “sweet spots” in the \((k, z)\) space where such a transition scale can be detected by the survey, while the frequencies of the modes tell us how well a transition can be resolved. For LSST, the best-measured modes peak in the region \( 0.04 < k < 0.16 h^{-1} \) Mpc and \( 0.5 < z < 2 \), indicating a sensitivity to a Compton scale today of \( 50 \lesssim \lambda_0 \lesssim 1500 \) Mpc, where we have allowed for a range of possible time evolutions of the mass scale [8]. Massive neutrinos also introduce a transition in \( \mu \) due to freestreaming. From the expression for the freestreaming length in terms of \( z \) and the neutrino mass \( m_\nu \) (see, e.g., [20]), we find that the transition scale is within the LSST sensitivity window for \( 0.1 \approx m_\nu \approx 0.7 \) eV. Smaller masses induce an overall suppression of growth with no scale-dependent signatures. While observable to some extent, this suppression is largely degenerate with \( w(z) \), especially if one allows for an arbitrary evolution of \( w(z) \), as we have done.

In addition to constraining \( \mu \) or \( \gamma \) individually, a less ambitious yet equally interesting question is how sensitive data are to any departure from standard growth. Namely, one may ask if either function deviates from unity without specifying which. For this purpose, we want to save the information common to both functions, which we previously lost by mutual marginalization. Hence, we consider the combined principal components of \( \mu \) and \( \gamma \). We follow the same procedure as before, except now we diagonalize the block of the Fisher matrix containing \( \mu \) and \( \gamma \) pixels. The eigenvalues of these combined PCs are shown in the

![FIG. 2 (color online). Eigensurfaces for \( \mu \) and \( \gamma \), with mutual marginalization, for LSST (DES) along with Planck and JDEM.](241301-3)
lower panel of Fig. 1. Even today’s data can provide around 15 “constrained” modes. This is unsurprising since the LSS power spectrum, \( P(k) \), is known to better than 10% precision over an order of magnitude in \( k \). Changing \( \mu \) at any \( k \) directly affects \( P(k) \), which means that large variations in \( \mu \) are disallowed in the range where \( P(k) \) is well measured. After marginalizing over \( \gamma \), the direct impact on \( P(k) \) is lost, since one can now offset changes in \( \mu \) by adjusting \( \gamma \). The dashed lines in Fig. 1 represent the eigenvalues of \( \mu \) without marginalizing over \( \gamma \). They are comparable to the eigenvalues of combined PCs, supporting the above notion. From Fig. 1, we see that DES will provide around 20, while LSST will provide up to 100, “well-constrained” combined modes.

So far, our analysis has neglected systematic errors, which are model dependent and hard to predict. While we address systematics thoroughly in an upcoming publication, we report here the outcome of a preliminary analysis based on the assumptions in [21,22]. We repeat our PCA with extra parameters describing likely sources of systematic error: shifts in the centroids of the \( z \) bins, distortions of the \( z \)-bin distribution functions, and additive and multiplicative errors on the WL signal due to point-spread-function contributions, as in [21]. This adds 58 (80) parameters to our analysis for DES (LSST). We assume no “catastrophic” photo-\( z \) misestimation and apply a conservative set of priors [22] to these parameters and marginalize over them. The systematics result in a noticeable but not dramatic dilution of constraints on MG from DES; photo-\( z \) errors mostly affect the \( z \) dependence of MG, to which DES was only weakly sensitive. As discussed above, constraints from DES will be primarily on the scale dependence of \( \mu \) and \( \gamma \), and that information is mostly preserved. The impact of the systematics on LSST forecasts is more significant because LSST has more potential to resolve \( z \)-dependent features. There, too, we find that inclusion of systematics preserves most of the scale-dependent information but reduces our ability to measure eigenmodes of \( \mu \) with \( z \)-dependent features, underscoring the need to study and control systematic errors in lensing surveys. Even after accounting for systematics, LSST and DES are powerful probes of MG.

The Dark Energy Task Force [23] recently analyzed constraints on DE from future surveys, without considering MG, and found that \( w(z) = 0 \) and \( (dw/dz)|_{z=0} \) could both be constrained. A time-varying \( w(z) \) alters the growth dynamics in a scale-independent way, so the scale dependence of \( \mu \) and \( \gamma \) cannot be duplicated by a choice of \( w \). Furthermore, since we consider linear scales, the dominant portion of the information on MG comes from higher \( z \) (\( z > 0.5 \)), at which DE effects are not as important. Thus, in addition to measuring \( w(z) \), future surveys will tightly constrain scale-dependent departures from \( \Lambda \)CDM and those occurring at high redshifts. Note that including the nonlinear growth data from lower \( z \) requires a model-dependent treatment of MG, in which case \( w(z) \) and MG would be related by the same theory.

To recap, we have forecasted the constraints on modifications to GR from future data sets, in comparison with existing data. We find that data from Planck, JDEM, and LSST (DES) can tightly constrain around 100 (20) parameters of MG [corresponding to the 100 (20) eigenvalues in region \( T_1 \) in the combined \( \mu \) and \( \gamma \) analysis] if the systematics are negligible. Current data constrain only one parameter to this level. We have identified the regions in parameter space to which future data are most sensitive. Our technique can be used in survey design to move the sweet spots to the most interesting parts of parameter space. Our results use only linear-scale data, being a conservative “proof of concept” that upcoming surveys can rigorously test GR over cosmic distances.

We thank R. Crittenden, C. Shapiro, E. Bertschinger, T. Giannantonio, and K. Koyama for useful comments and discussion. L. P. and G. B. Z. are supported by the NSERC and SFU, A. S. by the NSF Grant No. AST-0708501, and J. Z. by the Fulbright Foundation, NSERC, and UCB.