Cosmological Constraints from the SDSS Luminous Red Galaxies


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We measure the large-scale real-space power spectrum $P(k)$ using luminous red galaxies (LRGs) in the Sloan Digital Sky Survey (SDSS) and use this measurement to sharpen constraints on cosmological parameters from the Wilkinson Microwave Anisotropy Probe (WMAP). We employ a matrix-based power spectrum estimation method using Pseudo-Karhunen-Loève eigenmodes, producing uncorrelated minimum-variance measurements in 20 $k$-bands of both the clustering power and its anisotropy due to redshift-space distortions, with narrow and well-behaved window functions in the range $0.01h$/$\text{Mpc} < k < 0.2h$/$\text{Mpc}$. Results from the LRG and main galaxy samples are consistent, with the former providing higher signal-to-noise. Our results are robust to omitting angular and radial density fluctuations and are consistent between different parts of the sky. They provide a striking confirmation of the predicted large-scale $\Lambda$CDM power spectrum. Combining only SDSS LRG and WMAP data places robust constraints on many cosmological parameters that complement prior analyses of multiple data sets. The LRGs provide independent cross-checks on $\Omega_m$ and the baryon fraction in good agreement with WMAP. Within the context of flat $\Lambda$CDM models, our LRG measurements complement WMAP by sharpening the constraints on the matter density, the neutrino density and the tensor amplitude by about a factor of two, giving $\Omega_m = 0.24 \pm 0.02$ (1σ), $\sum m_\nu \lesssim 0.9$ eV (95%) and $r < 0.3$ (95%). Baryon oscillations are clearly detected and provide a robust measurement of the comoving distance to the median survey redshift $z = 0.35$ independent of curvature and dark energy properties. Within the $\Lambda$CDM framework, our power spectrum measurement improves the evidence for spatial flatness, sharpening the curvature constraint $\Omega_{\text{tot}} = 1.05 \pm 0.05$ from WMAP alone to $\Omega_{\text{tot}} = 1.003 \pm 0.010$. Assuming $\Omega_{\text{tot}} = 1$, the equation of state parameter is constrained to $w = -0.94 \pm 0.09$, indicating the potential for more ambitious future LRG measurements to provide precision tests of the nature of dark energy. All these constraints are essentially independent of scales $k > 0.1h$/Mpc and associated nonlinear complications, yet agree well with more aggressive published analyses where nonlinear modeling is crucial.

I. INTRODUCTION

The dramatic recent progress by the Wilkinson Microwave Anisotropy Probe (WMAP) and other experiments [1–4] measuring the cosmic microwave background (CMB) has made non-CMB experiments even more important in the quest to constrain cosmological models and their free parameters. These non-CMB constraints are crucially needed for breaking CMB degeneracies [5, 6]; for instance, WMAP alone is consistent with a closed universe with Hubble parameter $h = 0.3$ and no cosmological constant [7]. As long as the non-CMB constraints are less reliable and precise than the CMB, they will be the limiting factor and weakest link in the precision cosmology endeavor. Much of the near-term progress in cosmology will therefore be driven by reductions in statistical and systematic uncertainties of non-CMB probes of the cosmic expansion history (e.g., SN Ia) and the matter power spectrum (e.g., Lyman α Forest, galaxy clustering and motions, gravitational lensing, cluster studies and 21 cm tomography).

The cosmological constraining power of three-dimensional maps of the Universe provided by galaxy redshift surveys has motivated ever more ambitious observational efforts such as the CfA/UZC [8, 9], LCRS [10], PSCz [11], DEEP [12], 2dFGRS [13] and SDSS [14] projects, resulting in progressively more accurate measurements of the galaxy power spectrum $P(k)$ [15–30]. Constraints on cosmological models from these data sets have been most robust when the galaxy clustering could be measured on large scales where one has confidence in the modeling of nonlinear clustering and biasing (e.g., [7, 31–42]).

Our goal in this paper therefore is to measure $P(k)$ on large scales using the SDSS galaxy redshift survey in a way that is maximally useful for cosmological parameter estimation, and to explore the resulting constraints on cosmological models. The emphasis of our cosmological analysis will be on elucidating the links between cosmological parameters and observable features of the WMAP and SDSS power spectra, and on how these two data sets alone provide tight and robust constraints on many parameters that complement more aggressive but more systematics-prone analyses of multiple data sets.

In a parallel paper, Percival et al. [43] present a power spectrum analysis of the Main Galaxy and LRG samples from the SDSS DR5 data set [44], which is a superset of the data used here. There are a number of differences in the analysis methods. Percival et al. use an FFT-based method to estimate the angle-averaged (monopole) redshift-space galaxy power spectrum. We use a Pseudo-Karhunen-Loève method [45, 46] (see further discussion and references below) to estimate the real space (as opposed to redshift space) galaxy power spectrum, using finger-of-god compression and linear theory to remove redshift-space distortion effects. In addition, the many technical decisions that go into these analyses, regarding completeness corrections, angular masks, K-corrections and so forth, were made independently for the two papers, and they present different tests for systematic uncertainties. Despite these many differences of detail, our conclusions agree to the extent that they overlap (as discussed in Section III F and Appendix A 1), a reassuring indication of the robustness of the results.
A. Relation between different samples

The amount of information in a galaxy redshift survey about the galaxy power spectrum $P_g(k)$ and cosmological parameters depends not on the number of galaxies per se, but on the effective volume of the survey, defined by \[ \text{V}_{\text{eff}}(k) \] as

$$\text{V}_{\text{eff}}(k) = \int \frac{n(r)P_g(k)}{1+n(r)P_g(k)} \, d^3r,$$  \hspace{1cm} (1)

where $n(r)$ is the expected number density of galaxies in the survey in the absence of clustering, and the FKP approximation of [19] has been used. The power spectrum error bars scale approximately as $\Delta P_g(k) \propto \text{V}_{\text{eff}}(k)^{-1/2}$, which for a fixed power $P_g$ is minimized if a fixed total number of galaxies are spaced with density $\bar{n} \sim P_g^{-1}$ [48]. The SDSS Luminous Red Galaxy (LRG) sample was designed [49, 50] to contain such “Goldilocks” galaxies with a just-right number density for probing the power around the baryon wiggle scale $k \sim (0.05 - 0.1) h/\text{Mpc}$. For comparison, the SDSS main galaxy sample [50] is much denser and is dominated by sample variance on these scales, whereas the SDSS quasar sample [51] is much sparser and is dominated by Poisson shot noise. As shown in [36], the effective volume of the LRG sample is about six times larger than that of the SDSS main galaxies even though the number of LRGs is an order of magnitude lower, and the LRG volume is over ten times larger than that of the 2dFGRS. These scalings are confirmed by our results below, which show that $(\Delta P_g/P_g)^2$ on large scales is about six times smaller for the SDSS LRGs than for the main sample galaxies. This gain results both from sampling a larger volume, and from the fact that the LRG are more strongly clustered (biased) than are ordinary galaxies; $P_g$ for LRGs is about 3 times larger than for the main galaxy sample.

We will therefore focus our analysis on the SDSS LRG sample. Although we also measure the SDSS main sample power spectrum, it adds very little in terms of statistical constraining power; increasing the effective volume by 15% cuts the error bar $\Delta P$ by only about $(1 + 0.15)^{1/2} - 1 \sim 7\%$. This tiny improvement is easily outweighed by the gain in simplicity from analyzing LRGs alone, where (as we will see) complications such as redshift-dependence of clustering properties are substantially smaller.

A complementary approach implemented by [41, 42] is to measure the angular clustering of SDSS LRGs with photometric redshifts, compensating for the loss of radial information with an order of magnitude more galaxies extending out to higher redshift. We will see that this gives comparable or slightly smaller error bars on very large scales $k \lesssim 0.02$, but slightly larger error bars on the smaller scales that dominate our cosmological constraints; this is because the number of modes down to a given scale $k$ grows as $k^3$ for our three-dimensional spectroscopic analysis, whereas they grow only as $k^2$ for a 2-dimensional angular analysis.

B. Relation between different methods

In the recent literature, two-point galaxy clustering has been quantified using a variety of estimators of both power spectra and correlation functions. The most recent power spectrum measurements for both the 2dFGRS [26, 29] and the SDSS [30, 38, 43] have all interpolated the galaxy density field onto a cubic grid and measured $P(k)$ using a Fast Fourier Transform (FFT).

Appendix A1 shows that as long as discretization errors from the FFT gridding are negligible, this procedure is mathematically equivalent to measuring the correlation function with a weighted version of the standard “DD-2DR+RR” method [52, 53], multiplying by “RR” and then Fourier transforming. Thus the only advantage of the FFT approach is numerical speedup, and comparing the results with recent correlation function analyses such as [36, 54–56] will provide useful consistency checks.

Another approach, pioneered by [45], has been to construct “lossless” estimators of the power spectrum with the smallest error bars that are possible based on information theory [23, 24, 27, 28, 34, 45, 46, 57, 58]. We will travel this complementary route in the present paper, following the matrix-based Pseudo Karhunen-Loève (PKL) eigenmode method described in [28], as it has the following advantages:

1. It produces power spectrum measurements with uncorrelated error bars.
2. It produces narrow and well-behaved window functions.
3. It is lossless in the information theory sense.
4. It treats redshift distortions without the small-angle approximation.
5. It readily incorporates the so-called integral constraint [16, 59], which can otherwise artificially suppress large-scale power.
6. It allows testing for systematics that produce excess power in angular or radial modes.

These properties make the results of the PKL-method very easy to interpret and use. The main disadvantage is that the PKL-method is numerically painful to implement and execute; our PKL analysis described below required about a terabyte of disk space for matrix storage and about a year of CPU time, which contributed to the long gestation period of this paper.

The rest of this paper is organized as follows. We describe our galaxy samples and our modeling of them in Section II and measure their power spectra in Section III. We explore what this does and does not reveal about cosmological parameters in Section IV. We summarize our conclusions and place them in context in Section V. Further details about analysis techniques are given in Appendix A.
Our analysis is based on 58,360 LRGs and 285,804 main galaxies (the “safe13” cut) from the 390,288 galaxies in the 4th SDSS data release (“DR4”) [74], processed via the SDSS data repository at New York University [75]. The details of how these samples were processed and modeled are given in Appendix A of [28] and in [36]. The bottom line is that each sample is completely specified by three entities:

1. The galaxy positions (RA, Dec and comoving redshift space distance $r$ for each galaxy),

2. The radial selection function $\bar{n}(r)$, which gives the expected number density of galaxies as a function of distance,

3. The angular selection function $\bar{n}(\hat{r})$, which gives the completeness as a function of direction in the sky, specified in a set of spherical polygons [76].

Our samples are constructed so that their three-dimensional selection function is separable, i.e., simply the product $\bar{n}(\mathbf{r}) = \bar{n}(\hat{r})\bar{n}(r)$ of an angular and a radial part; here $r \equiv |\mathbf{r}|$ and $\hat{r} \equiv \mathbf{r}/r$ are the comoving radial distance and the unit vector corresponding to the position $\mathbf{r}$. The effective sky area covered is $\Omega \equiv \int \bar{n}(\hat{r})d\Omega \approx 4259$ square degrees, and the typical completeness $\bar{n}(\hat{r})$ exceeds 90%. The radial selection function $\bar{n}(r)$ for the LRGs is the one constructed and described in detail in [36, 56], based on integrating an empirical model of the luminosity function and color distribution of the LRGs against the luminosity-color selection boundaries of the sample. Figure 1 shows that it agrees well with the observed galaxy distribution. The conversion from redshift $z$ to comoving distance was made for a flat $\Lambda$CDM cosmological model with $\Omega_m = 0.25$. If a different cosmological model is used for this conversion, then our measured dimensionless power spectrum $k^3P(k)$ is dilated very slightly (by $\lesssim 1\%$ for models consistent with our measurements) along the $k$-axis; we include this dilation effect in our cosmological parameter analysis as described in Appendix A.4.

For systematics testing and numerical purposes, we also analyze a variety of sub-volumes in the LRG sample. We split the sample into three radial slices, labeled NEAR (0.155 $< z <$ 0.300), MID (0.300 $< z <$ 0.380) and FAR (0.380 $< z <$ 0.474), containing roughly equal numbers of galaxies, as illustrated in Figure 2. Their galaxy-weighted mean redshifts are 0.235, 0.342 and 0.421, respectively. We also split the sample into the seven angular regions illustrated in Figure 3, each again containing roughly the same number of galaxies.

It is worth emphasizing that the LRGs constitute a remarkably clean and uniform galaxy sample, containing the same type of galaxy (luminous early-types) at all redshifts. Not only is it nearly complete ($\bar{n}(\hat{r}) \sim 1$ as mentioned above), but it is close to volume-limited for $z \lesssim 0.38$ [36, 49], i.e., for our NEAR and MID slices.
III. POWER SPECTRUM MEASUREMENTS

We measure the power spectrum of our various samples using the PKL method described in [28]. We follow the procedure of [28] exactly, with some additional numerical improvements described in Appendix A, so we merely summarize the process very briefly here. The first step is to adjust the galaxy redshifts slightly to compress so-called fingers-of-god (FOGs), virialized galaxy clusters that appear elongated along the line-of-sight in redshift space; we do this with several different thresholds and return to how this affects the results in Section IV F 2. The LRGs are not just brightest cluster galaxies; about 20% of them appear to reside in a dark matter halo with one or more other LRG’s. The second step is to expand the three-dimensional galaxy density field in $N$ three-
dimensional functions termed PKL-eigenmodes, whose variance and covariance retain essentially all the information about the \( k < 0.2h/\text{Mpc} \) power spectrum from the galaxy catalog. We use \( N = 42,000 \) modes for the LRG sample and 4000 modes for the main sample, reflecting their very different effective volumes. The third step is estimating the power spectrum from quadratic combinations of these PKL mode coefficients by a matrix-based process analogous to the standard procedure for measuring CMB power spectra from pixelized CMB maps. The second and third steps are mathematically straightforward but, as mentioned, numerically demanding for large \( N \).

A. Basic results

The measured real-space power spectra are shown in Figure 4 for the LRG and MAIN samples and are listed in Table 1. When interpreting them, two points should be borne in mind:

1. The data points (a.k.a. band power measurements) probe a weighted average of the true power spectrum \( P(k) \) defined by the window functions shown in Figure 5. Each point is plotted at the median \( k \)-value of its window with a horizontal bar ranging from the 20\(^{th} \) to the 80\(^{th} \) percentile.

2. The errors on the points, indicated by the vertical bars, are uncorrelated, even though the horizontal bars overlap. Other power spectrum estimation methods (see Appendix A1) effectively produce a smoothed version of what we are plotting, with error bars that are smaller but highly correlated.

<table>
<thead>
<tr>
<th>( k [h/\text{Mpc}] )</th>
<th>( P_k )</th>
</tr>
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<tbody>
<tr>
<td>0.012±0.005</td>
<td>124884 ± 18775</td>
</tr>
<tr>
<td>0.015±0.004</td>
<td>118184 ± 29400</td>
</tr>
<tr>
<td>0.018±0.004</td>
<td>134291 ± 21638</td>
</tr>
<tr>
<td>0.021±0.004</td>
<td>58644 ± 16647</td>
</tr>
<tr>
<td>0.024±0.004</td>
<td>105253 ± 12736</td>
</tr>
<tr>
<td>0.028±0.005</td>
<td>77699 ± 9666</td>
</tr>
<tr>
<td>0.032±0.006</td>
<td>57870 ± 7264</td>
</tr>
<tr>
<td>0.037±0.004</td>
<td>56516 ± 5466</td>
</tr>
<tr>
<td>0.04±0.008</td>
<td>39339 ± 2214</td>
</tr>
<tr>
<td>0.045±0.005</td>
<td>50125 ± 3991</td>
</tr>
<tr>
<td>0.049±0.007</td>
<td>45076 ± 2956</td>
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<tr>
<td>0.057±0.009</td>
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<td>0.065±0.010</td>
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<td>0.075±0.011</td>
<td>71566 ± 2536</td>
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<tr>
<td>0.087±0.012</td>
<td>24837 ± 991</td>
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<tr>
<td>0.106±0.013</td>
<td>21390 ± 778</td>
</tr>
<tr>
<td>0.115±0.014</td>
<td>17507 ± 629</td>
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<tr>
<td>0.133±0.015</td>
<td>15421 ± 516</td>
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<td>0.153±0.016</td>
<td>12399 ± 430</td>
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<tr>
<td>0.177±0.017</td>
<td>11237 ± 382</td>
</tr>
<tr>
<td>0.203±0.018</td>
<td>9345 ± 384</td>
</tr>
</tbody>
</table>

Figure 3: The angular distribution of our LRGs is shown in Hammer-Aitoff projection in celestial coordinates, with the seven colors/greys indicating the seven angular subsamples that we analyze.

Table 1 – The real-space galaxy power spectrum \( P_k(k) \) in units of \( (h^{-1}\text{Mpc})^3 \) measured from the LRG sample. The errors on \( P_k \) are 1\( \sigma \), uncorrelated between bands. The \( k \)-column gives the median of the window function and its 20\(^{th} \) and 80\(^{th} \) percentiles; the exact window functions from http://space.mit.edu/home/tegmark/sdss.html (see Figure 5) should be used for any quantitative analysis. Nonlinear modeling is definitely required if the six measurements on the smallest scales (below the line) are used for model fitting. These error bars do not include an overall calibration uncertainty of 3% (1\( \sigma \)) related to redshift space distortions (see Appendix A3).
FIG. 4: Measured power spectra for the full LRG and main galaxy samples. Errors are uncorrelated and full window functions are shown in Figure 5. The solid curves correspond to the linear theory ΛCDM fits to WMAP3 alone from Table 5 of [7], normalized to galaxy bias $b = 1.9$ (top) and $b = 1.1$ (bottom) relative to the $z = 0$ matter power. The dashed curves include the nonlinear correction of [29] for $A = 1.4$, with $Q_{nl} = 30$ for the LRGs and $Q_{nl} = 4.6$ for the main galaxies; see equation (4). The onset of nonlinear corrections is clearly visible for $k \gtrsim 0.09h$ Mpc (vertical line).

Our Fourier convention is such that the dimensionless power $\Delta^2$ of [77] is given by $\Delta^2(k) = 4\pi(k/2\pi)^3P(k)$.

Before using these measurements to constrain cosmological models, one faces important issues regarding their interpretation, related to evolution, nonlinearities and systematics.

B. Clustering evolution

The standard theoretical expectation is for matter clustering to grow over time and for bias (the relative clustering of galaxies and matter) to decrease over time [78–80] for a given class of galaxies. Bias is also
FIG. 5: The window functions corresponding to the LRG band powers in Figure 4 are plotted, normalized to have unit peak height. Each window function typically peaks at the scale \( k \) that the corresponding band power estimator was designed to probe.

luminosity-dependent, which would be expected to affect the FAR sample but not the MID and NEAR samples (which are effectively volume limited with a \( z \)-independent mix of galaxy luminosities [49]). Since the galaxy clustering amplitude is the product of these two factors, matter clustering and bias, it could therefore in principle either increase or decrease across the redshift range \( 0.155 < z < 0.474 \) of the LRG sample. We quantify this empirically by measuring the power spectra of the NEAR, MID and FAR LRG subsamples. The results are plotted in Figure 6 and show no evidence for evolution of the large-scale galaxy \( (k \lesssim 0.1 h/\text{Mpc}) \) power spectrum in either shape or amplitude. To better quantify this, we fit the WMAP-only best-fit \( \Lambda \)CDM model from Table 5 of [7] (solid line in Figure 6) to our power spectra, by scaling its predicted \( z = 0 \) matter power spectrum by \( b^2 \) for a constant bias factor \( b \), using only the 14 data points that are essentially in the linear regime, leftward of the dotted vertical line \( k = 0.09 h/\text{Mpc} \). For the NEAR, MID and FAR subsamples, this gives best fit bias factors \( b \approx 1.95, 1.91 \), and 2.02, respectively. The fits are all good, giving \( \chi^2 \approx 10.3, 11.2 \) and 15.9 for the three cases, in agreement with the expectation \( \chi^2 = 13 + \sqrt{2 \times 13} \approx 13 \pm 5 \) and consistent with the linear-theory prediction that the large-scale LRG power spectrum should not change its shape over time, merely (perhaps) its amplitude.

The overall amplitude of the LRG power spectrum is constant within the errors over this redshift range, in good agreement with the results of [41, 56] at the corresponding mean redshifts. Relative to the NEAR sample, the clustering amplitude is \( 2.4\% \pm 3\% \) lower in MID and \( 3.5\% \pm 3\% \) higher in FAR. In other words, in what appears to be a numerical coincidence, the growth over time in the matter power spectrum is approximately canceled by a drop in the bias factor to within our measurement uncertainty. For a flat \( \Omega_m = 0.25 \) \( \Lambda \)CDM model, the matter clustering grows by about 10% from the FAR to NEAR sample mean redshifts, so this suggests that the bias drops by a similar factor. For a galaxy population evolving passively, under the influence of gravity alone [78, 79], \( b \) would be expected to drop by about 5% over this redshift range; a slight additional drop could be caused by luminosity-dependent bias coupling to the slight change in the luminosity function for the FAR sample, which is not volume limited.

This cancellation of LRG clustering evolution is a fortuitous coincidence that simplifies our analysis: we can pool all our LRGs and measure a single power spectrum for this single sample. It is not a particularly surprising result: many authors have found that the galaxy clustering strength is essentially independent of redshift, even to redshifts \( z > 3 \) [81], and even the effect that is partly canceled (the expected 10% growth in matter clustering) is small, because of the limited redshift range probed.

C. Redshift space distortions

As described in detail in [28], an intermediate step in our PKL-method is measuring three separate power spectra, \( P_{gg}(k) \), \( P_{gv}(k) \) and \( P_{vv}(k) \), which encode clustering

FIG. 6: Same as Figure 4, but showing the NEAR (circles), MID (squares) and FAR (triangles) LRG subsamples. On linear scales, they are all well fit by the WMAP3 model with the same clustering amplitude, and there is no sign of clustering evolution.
Constraints on the redshift space distortion parameters

FIG. 8: Constraints on the redshift space distortion parameters $\beta$ and $r_{\text{gv}}$. The contours show the 1, 2 and 3$\sigma$ constraints from the observed LRG clustering anisotropy, with the circular dot indicating the best fit values. The diamond shows the completely independent $\beta$-estimate inferred from our analysis of the WMAP3 and LRG power spectra (it puts no constraints on $r_{\text{gv}}$, but has been plotted at $r_{\text{gv}} = 1$).

FIG. 7: Same as Figure 4, but multiplied by $k$ and plotted with a linear vertical axis to more clearly illustrate departures from a simple power law.

In linear perturbation theory, these three power spectra are related by [82]

$$P_{\text{gv}}(k) = \beta r_{\text{gv}} P_{\text{gg}}(k),$$  \hspace{1cm} (2) \\
$$P_{\text{vv}}(k) = \beta^2 P_{\text{gg}}(k),$$  \hspace{1cm} (3)

where $\beta \equiv f/b$, $b$ is the bias factor, $r_{\text{gv}}$ is the dimensionless correlation coefficient between the galaxy and matter density fields [79, 83, 84], and $f \approx \Omega_m^{0.6}$ is the dimensionless linear growth rate for linear density fluctuations. (When computing $f$ below, we use the more accurate approximation of [85].)

The LRG power spectrum $P(k)$ tabulated and plotted above is a minimum-variance estimator of $P_{\text{gg}}(k)$ that linearly combines the $P_{\text{gg}}(k)$, $P_{\text{gv}}(k)$ and $P_{\text{vv}}(k)$ estimators as described in [28] and Appendix A 3, effectively marginalizing over the redshift space distortion parameters $\beta$ and $r_{\text{gv}}$. As shown in Appendix A 3, this linear combination is roughly proportional to the angle-averaged (monopole) redshift-space galaxy power spectrum, so for the purposes of the nonlinear modeling in the next section, the reader may think of our measured $P(k)$ as essentially a rescaled version of the redshift space power spectrum. However, unlike the redshift space power spectrum measured with the FKP and FFT methods (Appendix A 1), our measured $P(k)$ is unbiased on large scales. This is because linear redshift distortions are treated exactly, without resorting to the small-angle approximation, and account is taken of the fact that the anisotropic survey geometry can skew the relative abundance of galaxy pairs around a single point as a function of angle to the line of sight.

The information about anisotropic clustering that is discarded in our estimation of $P(k)$ allows us to measure $\beta$ and perform a powerful consistency test. Figure 8 shows the joint constraints on $\beta$ and $r_{\text{gv}}$ from fitting equations (2) and (3) to the 0.01$h$/Mpc $\leq k \leq 0.09h$/Mpc LRG data, using the best fit WMAP3 model from Figure 4 for $P_{\text{gg}}(k)$ and marginalizing over its amplitude. The data are seen to favor $r_{\text{gv}} \approx 1$ in good agreement with prior work [86, 87]. Assuming $r_{\text{gv}} = 1$ (that galaxy density linearly traces matter density on these large scales) gives the measurement $\beta = 0.309 \pm 0.035$ (1$\sigma$). This measurement is rather robust to changing the FOG compression threshold by a notch (Section IV F 2) or slightly altering the maximum $k$-band included, both of which affect the central value by of order 0.01. As a cross-check, we can compute $\beta = f(\Omega_m, \Omega_\Lambda)/b$ at the median survey redshift based on our multi-parameter analysis presented in Section IV, which for our vanilla class of models gives $\beta = 0.280 \pm 0.014$ (marked with a di-
and LRG power spectrum shapes. Power to CMB power, and determines $\Omega_m$ distortions, but rather extracts in the literature: several reasons for this that have been extensively studied contrast, Figures 4, 6 and 7 show clear departures from theory matter power spectrum predicted by WMAP3. In of the LRG power spectrum were well fit by the linear graph in Figure 8)\(^1\). That these two $\beta$-measurements agree within 1\(\sigma\) is highly non-trivial, since the second $\beta$-measurement makes no use whatsoever of redshift space distortions, but rather extracts $b$ from the ratio of LRG power to CMB power, and determines $\Omega_m$ from CMB and LRG power spectrum shapes.

D. Nonlinear modeling

Above we saw that our $k < 0.09h$/Mpc measurements of the LRG power spectrum were well fit by the linear theory matter power spectrum predicted by WMAP3. In contrast, Figures 4, 6 and 7 show clear departures from the linear theory prediction on smaller scales. There are several reasons for this that have been extensively studied in the literature:

1. Nonlinear evolution alters the broad shape of the matter power spectrum on small scales.
2. Nonlinear evolution washes out baryon wiggles on small scales.
3. The power spectrum of the dark matter halos in which the galaxies reside differs from that of the underlying matter power spectrum in both amplitude and shape, causing bias.
4. Multiple galaxies can share the same dark matter halo, enhancing small-scale bias.

We fit these complications using a model involving the three “nuisance parameters” ($b, Q_m, k_s$) as illustrated in Figure 9. Following [29, 88], we model our measured galaxy power spectrum as

$$P_s(k) = P_{\text{dewiggled}}(k)b^21 + Q_m k^2 \frac{1}{1 + 1.4k}.$$ (4)

where the first factor on the right hand side accounts for the non-linear suppression of baryon wiggles and the last factor accounts for a combination of the non-linear change of the global matter power spectrum shape and scale-dependent bias of the galaxies relative to the dark matter. For $P_{\text{dewiggled}}(k)$ we adopt the prescription [88]

$$P_{\text{dewiggled}}(k) = W(k)P(k) + [1 - W(k)]P_{\text{nowiggle}}(k),$$ (5)

where $W(k) \equiv e^{-(k/k_s)^2/2}$ and $P_{\text{nowiggle}}(k)$ denotes the “no wiggle” power spectrum defined in [89] and illustrated in Figure 9. In other words, $P_{\text{dewiggled}}(k)$ is simply a weighted average of the linear power spectrum and the wiggle-free version thereof. Since the $k$-dependent weight $W(k)$ transitions from 1 for $k \ll k_s$ to 0 for $k \gg k_s$, equation (5) retains wiggles on large scales and gradually fades them out beginning around $k = k_s$. Inspired by [88], we define the wiggle suppression scale $k_s \equiv 1/\sigma$, where $\sigma \equiv \sqrt{\frac{1}{1}/A_s(0.6841)^{1/2}}$ and $\sigma_\parallel$ and $\sigma_\perp$ are given by equations (12) and (13) in [88] based on fits to cosmological N-body simulations. The expression in parenthesis is an amplitude scaling factor that equals unity for the best fit WMAP3 normalization $A_s = 0.6841$ of [7]. Essentially, $\sigma$ is the characteristic peculiar-velocity-induced displacement of galaxies that causes the wiggle suppression; [88] define it for a fixed power spectrum normalization, and it scales linearly with fluctuation amplitude, i.e., $\propto A_s^{1/2}$.

For the cosmological parameter range allowed by WMAP3, we find that $k_s \sim 0.1h$/Mpc, with a rather rather weak dependence on cosmological parameters (mainly $\Omega_m$ and $A_s$).

The simulations and analytic modeling described by [29] suggest that the $Q_m$-prescription given by equation (4) accurately captures the scale-dependent bias of galaxy populations on the scales that we are interested in, though they examined samples less strongly biased than the LRGs considered here. To verify the applicability of this prescription for LRGs in combination with

---

1 Here $\beta = f(\Omega_m, \Omega_A)/b$ is computed with $\Omega_m$, $\Omega_A$ and $b$ evaluated at the median redshift $z = 0.35$, when $b = 2.25 \pm 0.08$, taking into account linear growth of matter clustering between then and now.
Figure 10: The points in the bottom panel show the ratio of the real-space power spectrum from 51 averaged $n$-body simulations (see text) to the linear power spectrum dewiggled with $k_\star = 0.1 h$/Mpc. Here LRGs were operationally defined as halos with mass exceeding $8 \times 10^{12} M_\odot$, corresponding to at least ten simulation particles. The solid curve shows the prediction from equation (4) with $b = 2.02$, $Q_{nl} = 27$, seen to be an excellent fit for $k \lesssim 0.4 h$/Mpc. The top panel shows the ratio of the simulation result to this fit. Although the simulation specifications and the LRG identification prescription can clearly be improved, they constitute the first and only that we tried, and were in no way adjusted to try to fit our $Q_{nl} = 30 \pm 4$ measurement from Table 2. This agreement suggests that our use of equations (4) and (5) to model nonlinearities is reasonable and that our measured $Q_{nl}$-value is plausible.

our dewiggling model, we reanalyze the 51 $n$-body simulations described in [90], each of which uses a $512 h^{-1}$ Mpc box with $256^3$ particles and WMAP1 parameters. Figure 10 compares these simulation results with our nonlinear modeling prediction defined by equations (4) and (5) for $b = 2.02$, $Q_{nl} = 27.0$, showing excellent agreement (at the 1% level) for $k \lesssim 0.4 h$/Mpc. Choosing a $k_\star$ very different from $0.1 h$/Mpc causes 5% wiggles to appear in the residuals because of a over- or under-suppression of the baryon oscillations. These simulations are likely to be underresolved and the LRG halo prescription used (one LRG for each halo above a threshold mass of $8 \times 10^{12} M_\odot$) is clearly overly simplistic, so the true value of $Q_{nl}$ that best describes LRGs could be somewhat different. Nonetheless, this test provides encouraging evidence that equation (4) is accurate in combination with equation (5) and that our $Q = 30 \pm 4$ measurement from Table 2 is plausible. Further corroboration is provided by the results in [41] using the Millennium Simulation [91]. Here LRG type galaxies were simulated and selected in an arguably more realistic way, yet giving results nicely consistent with Figure 10, with a best-fit value $Q_{nl} \approx 24$. (We will see in Section IVF that FOG-compression can readily account for these slight differences in $Q_{nl}$-value.) A caveat to both of these simulation tests is that they were performed in real space, and our procedure for measuring $P_g(k)$ reconstructs the real space power spectrum exactly only in the linear regime [28]. Thus, these results should be viewed as encouraging but preliminary, and more work is needed to establish the validity of the nonlinear modeling beyond $k \gtrsim 0.1 h$/Mpc; for up-to-date discussions and a variety of ideas for paths forward, see, e.g., [92–95].

In addition to this simulation-based theoretical evidence that our nonlinear modeling method is accurate, we have encouraging empirical evidence: Figure 9 shows an excellent fit to our measurements. Fitting the best-fit WMAP3 model from [32] to our first 20 data points (which extend out to $k = 0.2 h$/Mpc) by varying $(b, Q_{nl})$ gives $\chi^2 = 19.2$ for $20 - 2 = 18$ degrees of freedom, where the expected 1σ range is $\chi^2 = 18 \pm (2 \times 18)^{1/2} = 18 \pm 6$, so the fit is excellent. Moreover, Figures 7 and 9 show that that main outliers are on large and highly linear scales, not on the smaller scales where our nonlinear modeling has an effect.

The signature of baryons is clearly seen in the measured power spectrum. If we repeat this fit with baryons replaced by dark matter, $\chi^2$ increases by 8.8, corresponding to a baryon detection at $3.0 \sigma$ (99.7% significance). Much of this signature lies in the acoustic oscillations: if we instead repeat the fit with $k_\star = 0$, corresponding to fully removing the wiggles, $\chi^2$ increases by an amount corresponding to a detection of wiggles at $2.3 \sigma$ (98% significance). The data are not yet sensitive enough to distinguish between the wiggled and dewiggled spectra; dewiggling reduces $\chi^2$ by merely 0.04.

In summary, the fact that LRGs tend to live in high-mass dark matter halos is a double-edged sword: it helps by giving high bias $b \approx 2$ and luminous galaxies observable at great distance, but it also gives a stronger nonlinear correction (higher $Q_{nl}$) that becomes important on larger scales than for typical galaxies. Although Figure 10 suggests that our nonlinear modeling is highly accurate out to $k = 0.4 h$/Mpc, we retain only measurements with $k \lesssim 0.2 h$/Mpc for our cosmological parameter analysis to be conservative, and plan further work to test the validity of various nonlinear modeling approaches. In Section IV F 2, we will see that our data with $0.09 h$/Mpc $< k \lesssim 0.2 h$/Mpc, where nonlinear effects are clearly visible, allow us to constrain the nuisance parameter $Q_{nl}$ without significantly improving our constraints on cosmological parameters. In other words, the cosmological constraints that we will report below are quite insensitive to our nonlinear modeling and come mainly from the linear power spectrum at $k < 0.09 h$/Mpc. More sophisticated treatments of galaxy bias in which $Q_{nl}$ is effectively computed from theoretical models constrained by small scale clustering may eventually obviate the need...
subsection Analysis of subsets of galaxies

To test for effects that would be expected to vary across the sky (depending on, say, reddening, seasonally variable photometric calibration errors, or observing conditions such as seeing and sky brightness), we repeat our entire analysis for the seven different angular subsets of the sky shown in Figure 3 in search of inconsistencies. To search for potential zero-point offsets and other systematic effects associated with the southern Galactic stripes, they are defined as one of these seven angular subsets (see Figure 3). To test for effects that depend on redshift, we use the measurements for our three redshift slices, plotted in Figure 6.

To test the null hypothesis that all these subsamples are consistent with having the same power spectrum, we fit them all to our WMAP+LRG best-fit vanilla model described in Section IV, including our nonlinear correction (this $P(k)$ curve is quite similar to the best-fit WMAP3 model plotted above in, e.g., Figure 4).

We include the 20 band-powers with $k \lesssim 0.2$ in our fit, so if the null hypothesis is correct, we expect a mean $\chi^2$ of 20 with a standard deviation of $\sqrt{2} \times 20 \approx 6.3$. Our seven angular subsamples give a mean $\langle \chi^2 \rangle \approx 22.6$ and a scatter $\langle (\chi^2 - 20)^2 \rangle^{1/2} \approx 6.9$. Our three radial subsamples give $\langle \chi^2 \rangle \approx 18.6$ and $\langle (\chi^2 - 20)^2 \rangle^{1/2} \approx 2.4$. All of the ten $\chi^2$-values are statistically consistent with the null hypothesis at the 95% level. We also repeated the cosmological parameter analysis reported below with the southern stripes omitted, finding no significant change in the measured parameter values. In other words, all our angular and radial subsamples are consistent with having the same power spectrum, so these tests reveal no evidence for systematic errors causing radial or angular power spectrum variations.

2. Analysis of subsets of modes

Because of their angular or radial nature, all potential systematic errors discussed above create excess power mainly in the radial and angular modes. As mentioned above, one of the advantages of the PKL method is that it allows these modes to be excluded from the analysis, in analogy to the way potentially contaminated pixels in a CMB map can be excluded from a CMB power spectrum analysis. To quantify any such excess, we therefore repeat our full-sample analysis with radial and/or angular modes deleted. The results of this test are shown in Figure 11 and are very encouraging; the differences are tiny. Any systematic errors adding power to these special modes would cause the black circles to lie systematically above the other points, but no such trend is seen, so there is no indication of excess radial or angular power in the data.

The slight shifts seen in the power on the largest scales are expected, since a non-negligible fraction of the information has been discarded on those scales. Figure 11...
shows that removing the special modes results in a noticeable error bar increase on the largest scales and essentially no change on smaller scales. This can be readily understood geometrically. If we count the number of modes that probe mainly scales \( k < k_* \), then the number of purely radial, purely angular and arbitrary modes will grow as \( k_*^2 \), \( k_*^2 \) and \( k_*^2 \), respectively. This means that “special” modes (radial and angular) will make up a larger fraction of the total pool on large scales (at small \( k \)), and that the purely radial ones will be outnumbered by the purely angular ones. Conversely, the first 12 modes are all special ones: the monopole, the seven modes related to local-group motion, one radial mode and three angular modes. This means that almost all information on the very largest scales is lost when discarding special modes. Figure 11 illustrates this with the leftmost point labeled “generic” both having large error bars and being shifted to the right, where more information remains — yet it is consistent, lying about 1.3\( \sigma \) above an imaginary line between the two leftmost black points. We also repeated the cosmological parameter analysis reported below with the special modes omitted, finding no significant change in the measured parameter values.

F. Other tests

We have found no evidence for systematic errors affecting our power spectrum, suggesting that such effects, if present, are substantially smaller than our statistical errors. For additional bounds on potential systematic errors in the SDSS LRG sample, see [43].

A direct comparison of our \( P(k) \)-measurement and that of [43] is complicated because these are not measurements of the same function. [43] measures the angle-averaged redshift-space galaxy power spectrum, whereas our PKL-method attempts to recover the real space galaxy power spectrum, using finger-of-god (FOG) compression and linear theory to remove redshift-space distortion effects [28]. The galaxy selection is also different, with [43] mixing main sample galaxies in with the LRGs. Both of these differences are expected to affect the nonlinear corrections. In addition, the quantity \( P(k) \) plotted in [43] has correlated points with broader window functions than our uncorrelated points, and the angular coverage of the sample used in [43] is about 20% larger. To make a direct but approximate comparison with [43], we perform our own FKP analysis, both with and without FOG-compression, and as described in Appendix A 1, we obtain good agreement with [43] on linear scales for the case of no defogging.

We further investigate the robustness of our results to systematic errors in Section IV F below, this time focusing on their potential impact on cosmological parameters.

IV. COSMOLOGICAL PARAMETERS

Let us now explore the cosmological implications of our measurements by combining them with those from WMAP. As there has recently been extensive work on constraining cosmological parameters by combining multiple cosmological data sets involving CMB, galaxy clustering, Lyman α Forest, gravitational lensing, supernovae Ia and other probes (see, in particular, [7, 39]), we will focus more narrowly on what can be learned from WMAP and the LRGs alone. This is interesting for two reasons:

1. Less is more, in the sense that our results hinge on fewer assumptions about data quality and modeling. The WMAP and LRG power spectra suffice to break all major degeneracies within a broad class of models, yet they are also two remarkably clean measurements, probing gravitational clustering only on very large scales where complicated nonlinear physics is unlikely to cause problems.

2. Since the LRG power spectrum is likely to be included (together with WMAP and other data sets) in future parameter analyses by other groups, it is important to elucidate what information it contains about cosmological parameters. We will therefore place particular emphasis on clarifying the links between cosmological parameters and observable features of both the LRG and WMAP power spectra, notably the LRG matter-radiation equality scale, the LRG acoustic scale, the CMB acoustic scale, unpolarized CMB peak height ratios and large-scale CMB polarization.

We then compare our constraints with those from other cosmological probes in Section V C. We also compare our results with the analysis of [36] below, which had the narrower focus of measuring the LRG acoustic scale; the correlation function analysis in that paper complements our present analysis, since the acoustic oscillations in \( P(k) \) correspond to a readily measured single localized feature in real space [36, 96].

We work within the context of the arguably simplest inflationary scenario that fits our data. This is a hot Big Bang cosmology with primordial fluctuations that are adiabatic (i.e., we do not include isocurvature modes) and Gaussian, with negligible generation of fluctuations by cosmic strings, textures or domain walls. We assume the standard model of particle physics with three active neutrino species, very slightly heated during the era of electron/positron annihilation [97]. Within this framework, we parameterize our cosmological model in terms of 12 parameters that are nowadays rather standard, augmented with the two nuisance parameters \( b \) and \( Q_m \) from equation (4):

\[
p \equiv (\Omega_{\text{tot}}, \Omega_A, \omega_b, \omega_c, \omega_{\nu}, w, A_s, r, n_s, n_l, \alpha, \tau, b, Q_m).
\]

(6)

Table 2 defines these 14 parameters and another 45 that can be derived from them; in essence,
(Ω_{\text{tot}}, \Omega_{\Lambda}, \omega_b, \omega_c, \omega_r, w) define the cosmic matter budget, \((A_s, n_s, \alpha, r, n_I)\) specify the seed fluctuations and \((\tau, b, Q_{\text{nl}})\) are nuisance parameters. We will frequently use the term “vanilla” to refer to the minimal model space parametrized by \((\Omega_{\Lambda}, \omega_b, \omega_c, A_s, n_s, \tau, b, Q_{\text{nl}})\), setting \(\omega_r = \alpha = r = n_I = 0\), \(\Omega_{\text{tot}} = 1\) and \(w = -1\); this is the smallest subset of our parameters that provides a good fit to our data. Since current \(n_I\)-constraints are too weak to be interesting, we make the slow-roll assumption \(n_I = -r/8\) throughout this paper rather than treat \(n_I\) as a free parameter.

All our parameter constraints were computed using the now standard Monte Carlo Markov Chain (MCMC) approach [98–104] as implemented in [33] 2.

### A. Basic results

Our constraints on individual cosmological parameters are given in Tables 2 and 3 and illustrated in Figure 12, both for WMAP alone and when including our SDSS LRG information. Table 2 and Figure 12 take the Occam’s razor approach of marginalizing only over “vanilla” parameters \((\Omega_{\Lambda}, \omega_b, \omega_c, A_s, n_s, \tau, b, Q_{\text{nl}})\), whereas Table 3 shows how key results depend on assumptions about the non-vanilla parameters \((\Omega_{\text{tot}}, \omega_r, w, \alpha, r)\) introduced one at a time. In other words, Table 2 and Figure 12 use the vanilla assumptions by default; for example, models with \(\omega_r \neq 0\) are used only for the constraints on \(\omega_r\) and other neutrino parameters \((\Omega_{\nu}, \xi_{\nu}, f_{\nu})\).

The parameter measurements and error bars quoted in the tables correspond to the median and the central 68% of the probability distributions, indicated by three vertical lines in Figure 12. When a distribution peaks near zero, we instead quote an upper limit at the 95th percentile. Note that the tabulated median values are near but not identical to those of the maximum likelihood model. Our best fit vanilla model has \(\Omega_{\Lambda} = 0.763, \omega_b = 0.0223, \omega_c = 0.105, A_s = 0.685, n_s = 0.954, \tau = 0.0842, b = 1.90, Q_{\text{nl}} = 31.0\). As customary, the 2\(\sigma\) contours in the numerous two-parameter figures below are drawn where the likelihood has dropped to 0.0455 of its maximum value, which corresponds to \(\Delta \chi^2 \approx 6.18\) and 95.45% \(\approx 95\%\) enclosed probability for a two-dimensional Gaussian distribution.

We will spend most of the remainder of this paper digesting this information one step at a time, focusing on what WMAP and SDSS do and don’t tell us about the underlying physics, and on how robust the constraints are to assumptions about physics and data sets. The one-dimensional constraints in the tables and Figure 12 fail to reveal important information hidden in parameter correlations and degeneracies, so we will study the joint constraints on key 2-parameter pairs. We will begin with the vanilla 6-parameter space of models, then introduce additional parameters (starting in Section IV B) to quantify both how accurately we can measure them and to what extent they weaken the constraints on the other parameters.

First, however, some of the parameters in Table 2 deserve comment. The additional parameters below the double line in Table 2 are all determined by those above the double line by simple functional relationships, and fall into several groups.

Together with the usual suspects under the heading “other popular parameters”, we have included alternative fluctuation amplitude parameters: to facilitate comparison with other work, we quote the seed fluctuation amplitudes not only at the scale \(k = 0.05/\text{Mpc}\) employed by CMBfast [113], CAMB [114] and CosmoMC [102] (denoted \(A_s\) and \(r\)), but also at the scale \(k = 0.002/\text{Mpc}\) employed by the WMAP team in [7] (denoted \(A_{002}\) and \(r_{002}\)).

The “cosmic history parameters” specify when our universe became matter-dominated, recombined, reionized, started accelerating \((\ddot{a} > 0)\), and produced us.

Those labeled “fundamental parameters” are intrinsic properties of our universe that are independent of our observing epoch \(t_{\text{now}}\). (In contrast, most other parameters would have different numerical values if we were to measure them, say, 10 Gyr from now. For example, \(t_{\text{now}}\) would be about 24 Gyr, \(z_{\text{eq}}\) and \(\Omega_{\Lambda}\) would be larger, and \(\Omega_m\) and \(\omega_m\) would all be smaller. Such parameters are therefore not properties of our universe, but merely alternative time variables.)

The \(Q\)-parameter (not to be confused with \(Q_{\text{nl}}\)) is the primordial density fluctuation amplitude \(\sim 10^{-5}\). The curvature parameter \(\kappa\) is the curvature that the Universe would have had at the Planck time if there was no inflationary epoch, and its small numerical value \(\sim 10^{-61}\) constitutes the flatness problem that inflation solves. \((\xi_{\Lambda}, \xi_b, \xi_c, \xi_{\nu})\) are the fundamental parameters corresponding to the cosmologically popular quartet \((\Omega_m, \Omega_b, \Omega_c, \Omega_{\nu})\), giving the densities per CMB photon. The current densities are \(\rho_i = \rho_h \omega_i\), where \(i = m, b, c, \nu\) and \(\rho_h\) denotes the constant reference density.
<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
<th>Meaning</th>
<th>Definition</th>
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<tbody>
<tr>
<td><strong>Matter budget parameters:</strong></td>
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<td></td>
<td></td>
</tr>
<tr>
<td>Ω_m</td>
<td>0.301 ± 0.019</td>
<td>Hubble parameter</td>
<td>$h = \sqrt{(\omega_0 + \omega_r)/(\Omega_m - \Omega_k)}$</td>
</tr>
<tr>
<td>Ω_b</td>
<td>0.046 ± 0.018</td>
<td>Matter density/critical density</td>
<td>$\Omega_m = \Omega_{\text{tot}} - \Omega_k$</td>
</tr>
<tr>
<td>Ω_ν</td>
<td>0.017 ± 0.018</td>
<td>Baryon density</td>
<td>$\Omega_b = \rho_b / (1.88 \times 10^{-26} \text{kg/m}^3)$</td>
</tr>
<tr>
<td>ξ_c</td>
<td>0.593 ± 0.016</td>
<td>Cold dark matter density</td>
<td>$\rho_c = \rho_{\text{cdm}} / (1.88 \times 10^{-26} \text{kg/m}^3)$</td>
</tr>
<tr>
<td>r</td>
<td>&lt; 0.044</td>
<td>Tensor spectral index</td>
<td>$r = -\gamma$ assumed</td>
</tr>
<tr>
<td>σ_8</td>
<td>0.808 ± 0.03</td>
<td>Running of spectral index</td>
<td>$\sigma_8 = \delta_{\text{8, peak}} / \delta_{\text{8, peak}}$ (approximated as constant)</td>
</tr>
<tr>
<td><strong>Seed fluctuation parameters:</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>A_s</td>
<td>0.696 ± 0.045</td>
<td>Scaler fluctuation amplitude</td>
<td>Primordial scalar power at $k = 0.05,\text{Mpc}$</td>
</tr>
<tr>
<td>f</td>
<td>&lt; 0.30 (95%)</td>
<td>Tensor-to-scalar ratio</td>
<td>Primordial spectral index at $k = 0.05,\text{Mpc}$</td>
</tr>
<tr>
<td>b</td>
<td>0.069</td>
<td>Galaxy bias factor</td>
<td>$b =</td>
</tr>
<tr>
<td>Q_{nl}</td>
<td>30.37 ± 0.22</td>
<td>Nonlinear correction parameter</td>
<td>$P_N(k) = P_{\text{dewiggled}}(k) (1 + Q_{\text{nl}} k^2) / (1 + 1.7 k)$</td>
</tr>
<tr>
<td><strong>Other popular parameters (determined by those above):</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>τ</td>
<td>0.058 ± 0.016</td>
<td>Reionization optical depth</td>
<td>$b =</td>
</tr>
<tr>
<td>h</td>
<td>0.879 ± 0.016</td>
<td>Hubble parameter</td>
<td>$H_0 = 73.8 \pm 2.4,\text{km/s/Mpc}$</td>
</tr>
<tr>
<td><strong>Cosmic history parameters:</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>t_{eq}</td>
<td>4057 ± 102</td>
<td>Recombination redshift</td>
<td>$t_{eq} = 4274,\text{Gyr}$</td>
</tr>
<tr>
<td>t_{reion}</td>
<td>1090 ± 59.9</td>
<td>Recombination reheat</td>
<td>$t_{reion} = \gamma_{002} / (\Omega_{m}/\Omega_{\Lambda})$</td>
</tr>
<tr>
<td>t_{ion}</td>
<td>11.1 ± 2</td>
<td>Reionization redshift (abrupt)</td>
<td>$t_{reion} = 92(0.033,\text{yr} / (\Omega_{m}/\Omega_{\Lambda}) ) \times 3 / 3$ (assuming abrupt reionization: [109])</td>
</tr>
<tr>
<td>z_{acc}</td>
<td>0.855 ± 0.059</td>
<td>Acceleration redshift</td>
<td>$z_{acc} = (1 - 26 + 3,\text{Myr} / (\Omega_{m}/\Omega_{\Lambda}) - 1 - 3,\text{Gyr} = 1 - 1 / 3$</td>
</tr>
<tr>
<td>t_{eq}</td>
<td>0.0634 ± 0.045</td>
<td>Matter-redshift equality time</td>
<td>$z_{eq} = 9.778,\text{Gyr} \times h^{-1} [\Omega_{m}(1 + z)]^{1/3}$</td>
</tr>
<tr>
<td>t_{rec}</td>
<td>0.850 ± 0.040</td>
<td>Matter-redshift equality time</td>
<td>$t_{rec} = 9.778,\text{Gyr} \times h^{-1} [\Omega_{m}(1 + z)]^{1/3}$</td>
</tr>
<tr>
<td>t_{em}</td>
<td>0.437 ± 0.20</td>
<td>Recombination time</td>
<td>$t_{em} = 9.778,\text{Gyr} \times h^{-1} [\Omega_{m}(1 + z)]^{1/3}$</td>
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<tr>
<td>t_{acc}</td>
<td>0.747 ± 0.25</td>
<td>Acceleration time</td>
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<tr>
<td>t_{now}</td>
<td>13.76 ± 0.15</td>
<td>Age of Universe now</td>
<td>$t_{now} = 9.778,\text{Gyr} \times h^{-1} [\Omega_{m}(1 + z)]^{1/3}$</td>
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<tr>
<td><strong>Fundamental parameters (independent of observing epoch):</strong></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>Q</td>
<td>1.498 ± 0.03</td>
<td>Primordial fluctuation amplitude</td>
<td>$Q = \delta_{\text{8, peak}} / 59.2384,\text{kg/K} / \text{m}^3$</td>
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<tr>
<td>α</td>
<td>3.37 ± 0.13</td>
<td>Dimensionless spatial curvature [110]</td>
<td>$\kappa = 8 (\Omega_{m} - \Omega_{\Lambda})$</td>
</tr>
<tr>
<td>r_{halo}</td>
<td>1.485 ± 0.11</td>
<td>Dark energy density</td>
<td>$r_{halo} = 1.8 \times 10^{27} \text{g/cm}^3$</td>
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<tr>
<td>ξ_{halo}</td>
<td>6.1 ± 1.2</td>
<td>Halo formation density</td>
<td>$\delta_{\text{halo}} = 18.82 \Omega_{m}^{3/2}$</td>
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<tr>
<td>ξ_{v}</td>
<td>1.265 ± 0.11</td>
<td>Mass/energy density</td>
<td>$\xi = \rho_{\gamma} / \rho_{\Lambda}$</td>
</tr>
<tr>
<td>ξ</td>
<td>0.569 ± 0.018</td>
<td>mass: energy</td>
<td>$\xi = \rho_{\gamma} / \rho_{\Lambda}$</td>
</tr>
<tr>
<td>ξ_{ν}</td>
<td>2.269 ± 0.10</td>
<td>mass: energy</td>
<td>$\xi = \rho_{\gamma} / \rho_{\Lambda}$</td>
</tr>
<tr>
<td>θ</td>
<td>&lt; 0.26 (95%)</td>
<td>Neutrino mass: energy</td>
<td>$\theta = m_{\nu} / m_{\Lambda}$</td>
</tr>
<tr>
<td>η</td>
<td>0.60 ± 0.19</td>
<td>mass: energy</td>
<td>$\eta = m_{\nu} / m_{\Lambda}$</td>
</tr>
<tr>
<td>Λ_{eff}</td>
<td>0.077 ± 0.015</td>
<td>mass: energy</td>
<td>$\Lambda_{eff} = 0.077 \pm 0.015$</td>
</tr>
<tr>
<td>Σ_{gal}</td>
<td>0.561 ± 0.024</td>
<td>Mass/energy density</td>
<td>$\Sigma_{gal} = 0.561 \pm 0.024 \times 10^{-3}$</td>
</tr>
<tr>
<td>Σ_{gal}</td>
<td>0.561 ± 0.024</td>
<td>Mass/energy density</td>
<td>$\Sigma_{gal} = 0.561 \pm 0.024 \times 10^{-3}$</td>
</tr>
<tr>
<td><strong>CMB phenomenology parameters:</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>A_s</td>
<td>0.759 ± 0.013</td>
<td>Amplitude on CMB peak scale</td>
<td>$A_{s,\text{peak}} = 2.27$</td>
</tr>
<tr>
<td>n_s</td>
<td>0.959 ± 0.012</td>
<td>Amplitude at pivot point</td>
<td>$A_{s,\text{peak}} = 0.56 a^{-1} A_{s,\text{peak}}$</td>
</tr>
<tr>
<td>H_1</td>
<td>0.085 ± 0.011</td>
<td>1st CMB peak ratio</td>
<td>$H_1^{\text{1st}} = \Omega_{m} / (\Omega_{m} - \Omega_{v}) = \omega_{m} / a$</td>
</tr>
<tr>
<td>H_2</td>
<td>0.085 ± 0.011</td>
<td>2nd CMB peak ratio</td>
<td>$H_2^{\text{2nd}} = \Omega_{m} / (\Omega_{m} - \Omega_{v}) = \omega_{m} / (\Omega_{m} - \Omega_{v})$</td>
</tr>
<tr>
<td>H_3</td>
<td>0.085 ± 0.011</td>
<td>3rd CMB peak ratio</td>
<td>$H_3^{\text{3rd}} = \Omega_{m} / (\Omega_{m} - \Omega_{v}) = \omega_{m} / (\Omega_{m} - \Omega_{v})$</td>
</tr>
<tr>
<td>A_{s,\text{peak}}</td>
<td>0.14 ± 0.17</td>
<td>Comoving angular diameter distance to CMB</td>
<td>$A_{s,\text{peak}} = 0.14 \pm 0.17 ,\text{Gpc}$</td>
</tr>
<tr>
<td>A_{s,\text{peak}}</td>
<td>0.14 ± 0.17</td>
<td>Comoving sound horizon scale</td>
<td>$A_{s,\text{peak}} = 0.14 \pm 0.17 ,\text{Gpc}$</td>
</tr>
<tr>
<td>A_{s,\text{peak}}</td>
<td>0.14 ± 0.17</td>
<td>Comoving acoustic damping scale</td>
<td>$A_{s,\text{peak}} = 0.14 \pm 0.17 ,\text{Gpc}$</td>
</tr>
<tr>
<td>A_{s,\text{peak}}</td>
<td>0.14 ± 0.17</td>
<td>Comoving acoustic angular scale</td>
<td>$A_{s,\text{peak}} = 0.14 \pm 0.17 ,\text{Gpc}$</td>
</tr>
</tbody>
</table>
Table 3: How key cosmological parameter constraints depend on data used and on assumptions about other parameters. The columns compare different theoretical priors indicated by numbers in italics. \( \omega_c \) denotes dark energy that can cluster as in \([7]\). Rows labeled “+SDSS” combine WMAP3 and SDSS LRG data.

| Data       | \( \Omega_{m0} \) | 3(H/\kms)^2/8\pi G = 3(100 \kms^{-1} \Mpc^{-1})^2/8\pi G \approx 1.87882 \times 10^{-26} \text{kg/m}^3, \) so the conversion between the conventional and fundamental density parameters is \( \xi_i = \rho_i/n_c \approx 25.646 \text{eV} \times (T_{\text{cmb}}/2.726 \text{K}) \omega_i \) in units where \( c = 1 \). The parameter \( \xi_m \) is of the same order as the temperature at matter-radiation equality temperature, \( kT_{eq} \approx 0.22 \xi [115]^3 \).

The tiny value \( \sim 10^{-123} \) of the vacuum density \( \rho_\Lambda \) in Planck units where \( c = G = \hbar = 1 \) constitutes the well-known cosmological constant problem, and the tiny yet

\[
kT_{eq} = \frac{30\zeta(3)}{\pi^2} \left[ 1 + \frac{7}{8} N_\nu \left( \frac{4}{11} \right)^{4/3} \right]^{-1} \xi \approx 0.2195 \xi, \tag{7}
\]

where \( \zeta(3) \approx 1.202, \) and the effective number of neutrino species in the standard model is \( N_\nu \approx 3.022 \) \([97]\) when taking into account the effect of electron-positron annihilation on the relic neutrino energy density.

\[\Delta^2\text{WMAP} = 0.0 \quad \Delta^2\text{SDSS} = 0.0 \quad -0.5 \quad -2.4 \quad -0.5 \quad -0.9 \quad -0.3\]
similar value of the parameter combination $Q^3\xi^4$ explains the origin of attempts to explain this value anthropically [116–123]: $Q^3\xi^4$ is roughly the density of the universe at the time when the first nonlinear dark matter halos would form if $\rho_\Lambda = 0$ [115], so if $\rho_\Lambda \gg Q^3\xi^4$, dark energy freezes fluctuation growth before then and no nonlinear structures ever form.

The parameters $(A_\Lambda, \sigma_{gal}^2)$ are useful for anthropic
buffers, since they directly determine the density fluctuation history on galaxy scales through equation (5) in [111] (where \(\sigma^*_{\text{gal}}\) is denoted \(\sigma_M(0)\)). Roughly, fluctuations grow from the initial level \(\sigma^*_{\text{gal}}\) by a factor \(\Lambda\). Marginalizing over the neutrino fraction gives \(\Lambda^2 = 2279^{+240}_{-182}, \sigma^*_{\text{gal}} = 0.538^{+0.024}_{-0.022} \times 10^{-3}\).

The group labeled “CMB phenomenology parameters” contains parameters that correspond rather closely to the quantities most accurately measured by the CMB, such as heights and locations of power spectrum peaks. Many are seen to be measured at the percent level or better. These parameters are useful for both numerical and intuition-building purposes [105, 106, 112, 124–126]. Whereas CMB constraints suffer from severe degeneracies involving physical parameters further up in the table (involving, e.g., \(\Omega_\text{tot}\) and \(\Omega_\Lambda\)) as discussed below, these phenomenological parameters are all constrained with small and fairly uncorrelated measurement errors. By transforming the multidimensional WMAP3 log-likelihood function into the space spanned by \((H_0, \omega_m, f_p, \Omega_\Lambda, \omega_c, \Theta, A_\text{pivot}, H_3, \alpha, r, n_s, A_\tau, b, Q_\text{pol})\), it becomes better approximated by our quartic polynomial fit described in Footnote 2 and [106]: for example, the rms error is a negligible \(\Delta \ln \mathcal{L} \approx 0.03\) for the vanilla case. Roughly speaking, this transformation replaces the curvature parameter \(\Omega_\text{tot}\) by the characteristic peak scale \(\Theta\), the baryon fraction by the ratio \(H_0\) of the first two peak heights, the spectral index \(n_s\) by the ratio \(H_2\) of the third to first peak heights, and the overall peak amplitude \(A_{\text{peak}}\) by the amplitude \(A_{\text{pivot}}\) at the pivot scale where it is uncorrelated with \(n_s\). Aside from this numerical utility, these parameters also help demystify the “black box” aspect of CMB parameter constraints, elucidating their origin in terms of features in the data and in the physics [112].

### B. Vanilla parameters

Figure 12 compares the constraints on key parameters from the 1-year WMAP data (“WMAP1”), the 3-year WMAP data (“WMAP3”) and WMAP3 combined with our SDSS LRG measurements (“WMAP+LRG”). We include the WMAP1 case because it constitutes a well-tested baseline and illustrates both the dramatic progress in the field and what the key new WMAP3 information is, particularly from E-polarization.

#### 1. What WMAP3 adds

The first thing to note is the dramatic improvement from WMAP1 to WMAP3 emphasized in [7]. (Plotted WMAP1 constraints are from [33].) As shown in [127], this stems almost entirely from the new measurement of the low-\(\ell\) E power spectrum, which detects the reionization signature at about 3\(\sigma\) and determines the corresponding optical depth \(\tau = 0.09 \pm 0.03\). This measurement breaks the severe vanilla degeneracy in the WMAP1 data [32, 33] (see Figure 13) and causes the dramatic tightening of the constraints on \((\omega_\tau, \omega_c, \Omega_\Lambda, A_\tau, n_s)\) seen in the figures; essentially, with \(\tau\) well constrained, the ratio of large scale power to the acoustic peaks determines \(n_s\), and the relative heights of the acoustic peaks then determine \(\omega_\tau\) and \(\omega_c\) without residual uncertainty due to \(n_s\). Indeed, [127] has shown that discarding all the WMAP3 polarization data (both TE and EE) and replacing it with a Gaussian prior \(\tau = 0.09 \pm 0.03\) recovers parameter constraints essentially identical to those from the full WMAP3 data set. In Section IV F 1, we will return to the issue of what happens if this \(\tau\)-measurement is compromised by polarized foreground contamination.

The second important change from WMAP1 to WMAP3 is that the central values of some parameters have shifted noticeably [7]. Improved modeling of noise correlations and polarized foregrounds have lowered the low-\(\ell\) TE power and thus eliminated the WMAP1 evidence for \(\tau \sim 0.17\). Since the fluctuation amplitude scales as \(e^\tau\) times the CMB peak amplitude, this \(\tau\) drop of 0.08 would push \(\sigma_8\) down by about 8\%. In addition, better measurements around the 3rd peak and a change in analysis procedure (marginalizing over the SZ-contribution) have lowered \(\omega_m\) by about 13\%, causing fluctuation growth to start later (\(z_{eq}\) decreases) and end earlier (\(z_{ac}\) increases), reducing \(\sigma_8\) by another 8\%. These effects combine to lower \(\sigma_8\) by about 21\% when also taking into account the slight lowering of \(n_s\).

#### 2. What SDSS LRGs add

A key reason that non-CMB datasets such as the 2dFGRS and the SDSS improved WMAP1 constraints so dramatically was that they helped break the vanilla banana degeneracy seen in Figure 13, so that the fact that WMAP3 now mitigates this internally with its E-polarization measurement of \(\tau\) clearly reduces the value added by other datasets. However, Table 3 shows that our LRG measurements nonetheless give substantial improvements, cutting error bars on \(\Omega_\text{m}, \omega_c\) and \(h\) by a factor of two for vanilla models and by up to almost an order of magnitude when curvature, tensors, neutrinos or \(w\) are allowed.

The physics underlying these improvements is illustrated in Figure 14. The cosmological information in the CMB splits naturally into two parts, one “vertical” and one “horizontal”, corresponding to the vertical and horizontal positions of the power spectrum peaks.
FIG. 13: 95% constraints in the \((\Omega_m, h)\) plane. For 6-parameter “vanilla” models, the shaded red/grey region is ruled out by WMAP1 and the shaded orange/grey region by WMAP3; the main source of the dramatic improvement is the measurement of E-polarization breaking the degeneracy involving \(\tau\). Adding SDSS LRG information further constrains the parameters to the white region marked “Allowed”. The horizontal hatched band is required by the HST key project [137]. The dotted line shows the fit \(h = 0.72(\Omega_m/0.25)^{-0.32}\), explaining the origin of the percent-level constraint \(h(\Omega_m/0.25)^{0.32} = 0.719 \pm 0.008\) (1\(\sigma\)).

FIG. 14: Illustration of the physics underlying the previous figure. Using only WMAP CMB peak height ratios constrains \((\omega_m, \omega_b, n_s)\) independently of \(A_s\), \(\tau\), curvature and late-time dark energy properties. This excludes all but the white band \(\omega_m \equiv h^2\Omega_m = 0.127 \pm 0.017\) (2\(\sigma\)). If we assume \(\Omega_{tot} = 1\) and vanilla dark energy, we can supplement this with independent “standard ruler” information from either WMAP CMB (thin yellow/light grey ellipse) giving \(\Omega_m = 0.239 \pm 0.034\) (1\(\sigma\)), or SDSS galaxies (thicker blue/grey ellipse) giving \(\Omega_m = 0.239 \pm 0.027\) (1\(\sigma\)). These two rulers are not only beautifully consistent, but also complementary, with the joint constraints (small ellipse marked “allowed”) being tighter than those from using either separately, giving \(\Omega_m = 0.238 \pm 0.017\) (1\(\sigma\)). The plotted 2-dimensional constraints are all 2\(\sigma\). The three black curves correspond to constant “horizontal” observables: constant angular scales for the acoustic peaks in the CMB power (dotted, \(h \propto \Omega_m^{-0.3}\)), for the acoustic peaks in the galaxy power (solid, \(h \propto \Omega_m^{0.37}\)) and for the turnover in the galaxy power spectrum (dashed, \(h \propto \Omega_m^{0.93}\)). This illustrates why the galaxy acoustic scale is even more helpful than that of the CMB for measuring \(\Omega_m\); although it is currently less accurately measured, its degeneracy direction is more perpendicular to the CMB peak ratio measurement of \(h^2\Omega_m\).
By vertical information, we mean the relative heights of the acoustic peaks, which depend only on the physical matter densities \((\omega_m, \omega_b, \omega_c)\) and the scalar primordial power spectrum shape \((n_s, \alpha)\). They are independent of curvature and dark energy, since \(\Omega_{\Lambda}(z) \approx \Omega_{\Lambda}(0) \approx 0\) at \(z \gtrsim 10^3\). They are independent of \(h\), since the physics at those early times depended only on the expansion rate as a function of temperature back then, which is simply \(\xi^{1/2}T_{\text{eq}}^{3/2}\) times a known numerical constant, where \(\xi\) is given by \(\omega_m\) and the current CMB temperature (see Table 3 in [115]).

By horizontal (a.k.a. “standard ruler”) information, we mean the acoustic angular scale \(\ell_A \equiv \pi d_A(z_{\text{rec}})/r_s(z_{\text{rec}})\) defined in Table 2. The \(\ell\)-values of CMB power spectrum peaks and troughs are all equal to \(\ell_A\) times constants depending on \((\omega_m, \omega_b)\), so changing \(\ell_A\) by some factor by altering \((\Omega_k, \Omega_{\Lambda}, w)\) simply shifts the CMB peaks horizontally by that factor and alters the late integrated Sachs Wolfe effect at \(\ell \ll 10^2\). Although this single number \(\ell_A\) is now measured to great precision (\(\sim 0.3\%\)), it depends on multiple parameters, and it is popular to break this degeneracy with assumptions rather than measurements.

The sound horizon at recombination \(r_s(z_{\text{rec}})\) in the denominator depends only weakly on \((\omega_m, \omega_b)\), which are well constrained from the vertical information, and Table 2 shows that it is now known to about 1%. In contrast, the comoving angular diameter distance to recombination \(d_A(z_{\text{rec}})\) depends sensitively on both the spatial curvature \(\Omega_k\) and the cosmic expansion history \(H(z)\), which in turn depends on the history of the dark energy density:

\[
\frac{H(z)}{H_0} = \left[ X(z)\Omega_{\Lambda} + (1+z)^2\Omega_k + (1+z)^3\Omega_m + (1+z)^4\Omega_r \right]^{1/2}.
\]

(8)

Here \(X(z)\) is defined as the dark energy density relative to its present value [128], with vanilla dark energy (a cosmological constant) corresponding to \(X(z) = 1\). The most common (although physically unmotivated) parametrization of this function in the literature has been a simple power law \(X(z) = (1+z)^{3(1+w)}\), although it has also been constrained with a variety of other parametric and non-parametric approaches (see [129] and references therein). The parameter \(\Omega_r\) refers to the radiation contribution from photons and massless neutrinos, which is given by \(h^2\Omega_r \approx 0.0000416(T_{\text{cmb}}/2.726\text{K})^4\) and makes a negligible contribution at low redshift.

Using the vertical WMAP information alone gives a tight constraint on \(\omega_m \equiv h^2\Omega_m\), corresponding to the white band in Figure 14, independent of assumptions about curvature or dark energy.\(^4\) To this robust measurement, we can now add two independent pieces of information if we are willing to make the vanilla assumptions that curvature vanishes and dark energy is a cosmological constant: If we add the WMAP horizontal information, the allowed region shrinks to the thin ellipse hugging the \(h \propto \Omega_{m}^{-0.3}\) line of constant \(\ell_A\) (dotted). If we instead add the LRG information (which constrains \(h\Omega_m^{0.93}\) via the \(P(k)\) turnover scale and \(h\Omega_m^{-0.37}\) via the acoustic oscillation scale\(^5\)), the allowed region shrinks to the thick

\(^4\)To obtain this \(\omega_m\)-constraint, we marginalized over \(\ell_A\) by marginalizing over either \(\Omega_k\) or \(w\); Table 3 shows that these two approaches give essentially identical answers.

\(^5\)The origin of these scalings can be understood as follows. The matter-radiation equality horizon scale \(r_{\text{eq}} \propto \omega_m^{-1}\). The sound horizon scales as \(r_s(z_{\text{rec}}) \propto \omega_m^{-0.25}\) with a weak dependence on \(\omega_b\) that is negligible in this context [108]. For the LRG mean redshift \(z = 0.35\), the power law fit \(d_A(z, \Omega_m) = 3.3474(\Omega_m/0.25)^{-0.4} \times H_0^{-1} \propto h^{-1}\Omega_m^{-0.665}\) is quite good within our range of interest, accurate to within about 0.1% for \(0.2 < \Omega_m < 0.3\). For \(z = 1100\), the power law fit \(d_A(z, \Omega_m) \approx 3.4374(\Omega_m/0.25)^{-0.4} \times H_0^{-1} \propto h^{-1}\Omega_m^{-0.4}\) retains 0.1% accuracy for \(0.19 < \Omega_m < 0.35\). The \(P(k)\) turnover angle \(\propto r_{\text{eq}}/d_A(0.35) \propto (h^2\Omega_m)^{-1/2}\) is therefore constant for \(h \propto \Omega_m^{-0.93}\), the \(P(k)\) acoustic angle \(\propto r_s/d_A(0.35) \propto (h^2\Omega_m)^{-1/2}\) }
FIG. 16: 95% constraints in the \((\Omega_{\text{tot}}, h)\) plane for 7-parameter curved models. The shaded red/dark grey region was ruled out by WMAP1 alone, and WMAP3 tightened these constraints (orange/grey region), illustrating that CMB fluctuations alone do not simultaneously show space to be flat and measure the Hubble parameter. The yellow/light grey region is ruled out when adding SDSS LRG information. Here the yellow banana has been artificially cut off for \(h > 0.4\) by a hardwired prior in the CosmoMC software.

FIG. 17: 95% constraints in the \((\Omega_{\text{tot}}, t_{\text{now}})\) plane for 7-parameter curved models. The shaded red/dark grey region is ruled out by WMAP1 alone, and WMAP3 tightened these constraints (orange/grey region), illustrating that CMB fluctuations do not simultaneously show space to be flat and measure the age of the Universe. The yellow/light grey region is ruled out when adding SDSS LRG information. The age limit \(t_{\text{now}} > 12\) Gyr shown is the 95% lower limit from white dwarf ages by [132]; for a review of recent age determinations, see [7].

Let us now spice up the vanilla model space by including spatial curvature \(\Omega_k\) and a constant dark energy equation of state \(w\) as free parameters, both to constrain them and to quantify how other constraints get weakened when dropping these vanilla assumptions.

1. LRGs as a standard ruler at \(z = 0.35\)

Before constraining specific spacetime geometry parameters, let us review the relevant physics to intuitively understand what CMB and LRGs do and do not teach us about geometry. As discussed in the previous section, current CMB data accurately measure only a single number that is sensitive to the spacetime geometry information in \(\Omega_k\) and \(H(z)\). This number is the peak angular scale \(\ell_A\), and it in turn depends on the four independent parameters \((\Omega_m, \Omega_k, w, h)\). \((\Omega_\Lambda\) is of course not independent, fixed by the identity \(\Omega_\Lambda = 1 - \Omega_k - \Omega_m\).) Since the sound horizon size \(r_s\) is now accurately known independently of spacetime geometry from CMB peak ratios, the CMB \(\ell_A\)-measurement provides a precise determination of the comoving angular diameter distance to the last scattering surface, \(d_A(z_{\text{rec}})\), thus allowing one function

\[
(h^2\Omega_m)^{-0.25} / h^{-1} \Omega_m^{-0.065} \text{ is constant for } h \propto \Omega_m^{0.37}, \text{ and the } C_\ell \text{ acoustic angle } \propto r_s / d_A(z_{\text{rec}}) \propto (h^2\Omega_m)^{-0.25} / h^{-1} \Omega_m^{-0.4} \text{ is constant for } h \propto \Omega_m^{-0.3}.
\]
of $(\Omega_m, \Omega_k, w, h)$ to be accurately measured. As emphasized in [36, 128, 130, 131], measuring the acoustic angular scale at low redshift in galaxy clustering similarly constrains a second independent combination of $(\Omega_m, \Omega_k, w, h)$, and measuring $d_A(z)$ at multiple redshifts with future redshift surveys and current and future SN Ia data can break all degeneracies and allow robust recovery of both $\Omega_k$ and the dark energy history $X(z)$. For the galaxy approach the point is that leaving the early universe physics ($\omega_b, \omega_m, n_s$, etc.) fixed, changing the spacetime geometry merely scales the horizontal axis of the angular power spectrum of galaxies at a given redshift $z$ as $d_A(z)$. More generally, as described in detail in [36], the main effect of changing the spacetime geometry is to shift our measured three-dimensional power spectrum horizontally by rescaling the $k$-axis. The $k$-scale for angular modes dilates as the comoving angular diameter distance $d_A(z)$ to the mean survey redshift $z \approx 0.35$, whereas that for radial modes dilates as $d(dA)/dz = c/H(z)$ for the flat case. For small variations around our best fit model, the change in $H(0.35)$ is about half that of the angular diameter distance. To model this, [36] treats the net dilation as the cube root of the product of the radial dilation times the square of the transverse dilation, defining the distance parameter

$$d_V(z) \equiv \left[ d_A(z)^2 \frac{cz}{H(z)} \right]^{1/3}. \tag{9}$$

Using only the vertical WMAP peak height information as a prior on $(\omega_b, \omega_d, n_s)$, our LRG power spectrum gives the measurement $d_V(0.35) = 1.300 \pm 0.088$ Gpc, which agrees well with the value measured in [36] using the LRG correlation function. It is this geometric LRG information that explains most of the degeneracy breaking seen in the Figures 15, 16, 17 and 18 below. As more LRG data become available and strengthen the baryon bump detection from a few $\sigma$ to $> 5\sigma$, this measurement should become even more robust, not requiring any $\omega_m$-prior from WMAP peak heights.

2. Spatial curvature

Although it has been argued that closed inflation models require particularly ugly fine-tuning [133], a number of recent papers have considered nearly-flat models either to explain the low CMB quadrupole [134], in string theory landscape-inspired short inflation models, or for anthropic reasons [110, 135, 136], so it is clearly interesting and worthwhile to continue sharpening observational tests of the flatness assumption. In the same spirit, measuring the Hubble parameter $h$ independently of theoretical assumptions about curvature and measurements of galaxy distances at low redshift provides a powerful consistency check on our whole framework.

Figures 15, 16 and 17 illustrate the well-known CMB degeneracies between the curvature $\Omega_k = 1 - \Omega_{\text{tot}}$ and dark energy $\Omega_\Lambda$, the Hubble parameter $h$, and the age of the universe $t_{\text{now}}$: without further information or priors, one cannot simultaneously demonstrate spatial flatness and accurately measure $\Omega_\Lambda$, $h$ or $t_{\text{now}}$, since the CMB accurately constrains only the single combination $\ell_A$. Indeed, the WMAP3 degeneracy banana extends towards even larger $\Omega_{\text{tot}}$ than these figures indicate; the plotted banana has been artificially truncated by a hardwired lower limit on $h$ in the CosmoMC software used to compute this particular MCMC.

Including our LRG information is seen to reduce the curvature uncertainty by about a factor of five, providing a striking vindication of the standard inflationary prediction $\Omega_{\text{tot}} = 1$. The physical reason for this LRG improvement is obvious from the thick ellipse in Figure 14: WMAP vertical peak height information combined with LRG standard ruler information on $d_V(0.35)$ measures $\Omega_m$, rather independently of curvature.

Yet even with WMAP+LRG information, the figures show that a strong degeneracy persists between curvature and $h$, and curvature and $t_{\text{now}}$, leaving the measurement uncertainty on $h$ comparable with that from the HST key project [137]. If we add the additional assumption that space is exactly flat, then uncertainties shrink by factors around 4 and 10 for $h$ and $t_{\text{now}}$, respectively, still in beautiful agreement with other measurements.

In conclusion, within the class of almost flat models, the WMAP-only constraints on $h$, $t_{\text{now}}$, $\Omega_\Lambda$ and $\Omega_{\text{tot}}$ remain weak, and including our LRG measurements provides a huge improvement in precision.

3. Dark energy

Although we now know its present density fairly accurately, we still know precious little else about dark energy, and much interest is focused on understanding its nature. Assuming flat space1, Table 3 and Figure 18 show our constraints on constant $w$ for two cases: assuming that dark energy is homogeneous (does not cluster) and that it allows spatial perturbations (does cluster) as modeled in [7]. We see that adding $w$ as a free parameter does not significantly improve $\chi^2$ for the best fit, and all data are consistent with the vanilla case $w = -1$, with $1\sigma$ uncertainties in $w$ in the 10% - 30% range, depending on dark energy clustering assumptions.

As described above, the physical basis of these constraints is similar to those for curvature, since (aside from low-$\ell$ corrections from the late ISW effect and dark energy clustering), the only readily observable effect of the dark energy density history $X(z)$ is to alter $d_A(z_{\text{rec}})$ and $d_A(0.35)$, and hence the CMB and LRG acoustic angular scales. (The dark energy history also affects fluctuation growth and hence the power spectrum amplitude, but we do not measure this because our analysis marginalizes over the galaxy bias parameter $b$.)

It has been argued (see, e.g., [138]) that it is inappropriate to assume $\Omega_k = 0$ when constraining $w$, since there
is currently no experimental evidence for spatial flatness unless \( w = -1 \) is assumed. We agree with this critique, and merely note that no interesting joint constraints can currently be placed on as many as four spacetime geometry parameters \( (\Omega_m, \Omega_k, w, h) \) from WMAP and our LRG measurements alone, since they accurately constrain only the two combinations \( d_A(z_{\text{rec}}) \) and \( d_V(0.35) \). Other data such as SN Ia need to be included for this; [7] do this and obtain \( w = -1.06^{+0.13}_{-0.08} \).

One can also argue, in the spirit of Occam’s razor, that the fact that vanilla works so well can be taken as evidence against both \( \Omega_k \neq 0 \) and \( w \neq -1 \), since it would require a fluke coincidence for them to both have significantly non-vanilla values that conspire to lie on the same \( d_V(0.35) \) and \( d_A(z_{\text{rec}}) \) degeneracy tracks as the vanilla model.

### D. Inflation

Inflation [139–143] remains the leading paradigm for what happened in the early universe because it can solve the flatness, horizon and monopole problems (see, e.g., [144]) and has, modulo minor caveats, successfully predicted that \( \Omega_{\text{tot}} \approx 1 \), \( n_s \approx 1 \), \( |\alpha| \ll 1 \) and \( r \lesssim 1 \) as well as the facts that the seed fluctuations are mainly Gaussian and adiabatic. For the ekpyrotic universe alternative [145], controversy remains about whether it can survive a “bounce” and whether it predicts \( n_s \approx 1 \) [146] or \( n_s \approx 3 \) [147].

In the quest to measure the five parameters \( (Q, n_s, \alpha, r, n_t) \) characterizing inflationary seed fluctuations, the first breakthrough was the 1992 COBE discovery that \( Q \sim 10^{-5} \) and that the other four quantities were consistent with zero [149]. The second breakthrough is currently in progress, with WMAP3 suggesting \( 1 - n_s > 0 \) at almost the 3\( \sigma \) level \((1 - n_s = 0.049^{+0.019}_{-0.015}) \) [7]. This central value is in good agreement with classic (single slow-rolling scalar field) inflation models, which generically predict non-scale invariance in the ballpark \( 1 - n_s \sim 2/N \sim 0.04 \), assuming that the number of e-foldings between the time horizon the observed fluctuations exit the horizon and the end of inflation is \( 50 < N < 60 \) as per [150]. This central value of \( n_s \) agrees well with numerous
measurements in the recent literature (e.g., [151]); it is merely the error bars that have changed.

As illustrated in Figure 19 and discussed in [7], \( n_s = 1 \) becomes allowed if the tensor fluctuation parameter \( r \) is included (as it clearly should be when constraining inflation models), but the “vanilla lite” Harrison-Zeldovich model \((n_s = 1, r = 0)\) remains ruled out. In contrast, the arguably simplest of all inflation models, a single slow-rolling scalar field with potential \( V(\phi) \propto \phi^2 \), remains viable: it predicts \((n_s, r) = (1 - 2/N, 8/N) \approx (0.96, 0.15)\). The string-inspired “N-flation” model makes a similar prediction [152, 153].

Our constraints on the inflation parameters \((Q, n_s, \alpha, r, n_l)\) in Table 3 and Figure 19 are seen to confirm those reported in [7] — the main addition of our LRG analysis is simply to provide a clean way of tightening the WMAP-only constraints on both \( \Omega_{\text{tot}} \) and \( r \) (by factors of 5 and 2, respectively). Lyman-\( \alpha \) Forest (Ly\( \alpha \)F) constraints provide valuable complementary information on smaller scales, constraining the running of the spectral index to vanish at the percent level [39, 40].

Since the WMAP3 announcement, there has been substantial discussion of how strong the evidence against Harrison-Zeldovich \((n_s = 1, r = 0)\) really is [39, 127, 154–160]. For example, the WMAP team marginalized over the SZ-amplitude on small scales, which lowered the \( n_s \)-estimate by about 0.01, but did not model the CMB lensing effect, which would raise the \( n_s \)-estimate by a comparable amount [127]. It has also been argued that improved modeling of point source contamination increases the \( n_s \)-estimate [158]. Inclusion of smaller-scale CMB data and Ly\( \alpha \)F information clearly affects the significance as well. The bottom line is therefore that even modest improvements in measurement accuracy over the next few years can significantly improve our confidence in distinguishing between competing early-universe models — even without detecting \( r > 0 \).

E. Neutrinos

It has long been known [161] that galaxy surveys are sensitive probes of neutrino mass, since they can detect the suppression of small-scale power caused by neutrinos streaming out of dark matter overdensities. For detailed discussion of post-WMAP3 astrophysical neutrino constraints, see [7, 39, 162–167].

Our neutrino mass constraints are shown in Figure 20 and in the \( \Omega_{\nu} \)-panel of Figure 12, where we allow our standard 6 vanilla parameters and \( f_\nu \) to be free\(^6\). Assuming three active neutrinos with standard freezeout abundance, we obtain a 95% upper limit \( \Omega_{\nu} < 0.9 \) eV, so combining this with the atmospheric and solar neutrino oscillation results [168, 169], which indicate small mass differences between the neutrino types, implies that none of the three masses can exceed \( \Omega_{\nu}/3 \approx 0.3 \) eV. In other words, the heaviest neutrino (presumably in a hierarchical model mostly a linear combination of \( \nu_e \) and \( \nu_\tau \)) would have a mass in the range 0.04 – 0.3 eV.

If one is willing to make stronger assumptions about the ability to model smaller-scale physics, notably involving the Ly\( \alpha \)F, one can obtain the substantially sharper upper bound \( \Omega_{\nu} < 0.17 \) eV [39]. However, it should be noted that [39] also find that these same assumptions rule out the standard model with three active neutrino species at 2.5\( \sigma \), preferring more than three species.

F. Robustness to data details

Above, we explored in detail how our cosmological parameter constraints depend on assumptions about physics in the form of parameter priors \((\Omega_k = 0, w = -1, \text{etc.})\). Let us now discuss how sensitive they are to details related to data modeling.

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\(^6\) It has been claimed that the true limits on neutrino masses from the WMAP1 (but not WMAP3) CMB maps are tighter than represented in these figures [37, 163, 164].
1. CMB modeling issues

With any data set, it is prudent to be extra cautious regarding the most recent additions and the parts with the lowest signal-to-noise ratio. In the WMAP case, this suggests focusing on the \( T \) power spectrum around the third peak and the large-scale \( E \)-polarization data, which as discussed in Section IV B1 were responsible for tightening and lowering the constraints on \( \omega_m \) and \( \tau \), respectively.

The large-scale \( E \)-polarization data appear to be the most important area for further investigation, because they are single-handedly responsible for most of the dramatic WMAP3 error bar reductions, yet constitute only a \( 3\sigma \) detection after foregrounds an order of magnitude larger have been subtracted from the observed polarized CMB maps [2]. As discussed in [127] and Section IV B1, all the WMAP3 polarization information is effectively compressed into the probability distribution for \( \tau \), since using the prior \( \tau = 0.09 \pm 0.03 \) instead of the polarized data leaves the parameter constraints essentially unchanged. This error bar \( \Delta \tau = 0.03 \) found in [7] and Table 2 reflects only noise and sample variance and does not include foreground uncertainties. If future foreground modeling increases this error bar substantially, it will reopen the vanilla banana degeneracy described in [33]: Increasing \( \tau \) and \( A_s \) in such a way that \( A_{\text{peak}} \equiv A_s e^{-2\tau} \) stays constant, the peak heights remain unchanged and the only effect is to increase power on the largest scales. The large-scale power relative to the first peak can then be brought back down to the observed value by increasing \( n_s \), after which the second peak can be brought back down by increasing \( \omega_b \). Since quasar observations of the Gunn-Peterson effect allow \( \tau \) to drop by no more than about \( 1\sigma \) \( (0.03) \) [170, 171], the main change possible from revised foreground modeling is therefore that \( (\tau, \Omega_b, \omega_d, \omega_b, A_s, n_s, h) \) all increase together [33]. For a more detailed treatment of these issues, see [172].

A separate issue is that, as discussed in Section IV D, reasonable changes in the CMB data modeling can easily increase \( n_s \) by of order 0.01 [39, 127, 154–156, 158], weakening the significance with which the Harrison-Zeldovich model \( (n_s = 1, \tau = 0) \) can be ruled out.

With the above-mentioned exceptions, parameter measurements now appear rather robust to WMAP modeling details. We computed parameter constraints using the WMAP team chains available on the LAMBDA archive. We created our own chains using the CosmoMC package [102] for the vanilla case (of length 310,817) as a cross-check and for the case with curvature (of length 226,456) since this was unavailable on LAMBDA. The parameter constraints were in excellent agreement between these two vanilla chains. For a fair comparison between WMAP team and CosmoMC-based chains, the best-fit \( \chi^2 \) values listed in Table 3 have been offset-calibrated so that they all give the same value for our best fit vanilla model.

2. LRG modeling issues

Since we marginalize over the overall amplitude of LRG clustering via the bias parameter \( b \), the LRG power spectrum adds cosmological information only through its shape. Let us now explore how sensitive this shape is to details of the data treatment. A popular way to parametrize the power spectrum shape in the literature has been in terms of the two parameters \( (\Omega_m, f_b) \) shown in Figure 21, where \( f_b \equiv \Omega_b/\Omega_m \) is the baryon fraction. Since we wish to use \( (\Omega_m, f_b) \) merely to characterize this shape here, not for constraining cosmology, we will ignore all CMB data and restrict ourselves to vanilla models with \( n_s = 1, h = 0.72 \) and \( A_s = 1 \), varying only the four parameters \( (\Omega_m, f_b, b, Q_m) \). Figure 21 suggest that for vanilla models, the two parameters \( (\Omega_m, f_b) \) do in fact capture the bulk of this shape information, since the WMAP+LRG joint constraints from our full 6-parameter analysis are seen to be essentially the intersection of the WMAP and “ALL LRG” allowed regions, indicating that these two shape parameters carry the bulk of the cosmologically useful LRG information.

FIG. 21: The key information that our LRG measurements add to WMAP comes from the power spectrum shape. Parametrizing this shape by \( \Omega_m \) and the baryon fraction \( \Omega_b/\Omega_m \) for vanilla models with \( n_s = 1, h = 0.72 \), the 95% constraints above are seen to be nicely consistent between the various radial subsamples. Moreover, the WMAP+LRG joint constraints from our full 6-parameter analysis are seen to be essentially the intersection of the WMAP and “ALL LRG” allowed regions, indicating that these two shape parameters carry the bulk of the cosmologically useful LRG information.
power spectrum $k P(k)$ [h Mpc$^{-1}$] for the approximate scaling reducing the number of very close pairs, suppressing the galaxy clusters along the line of sight, thereby strongly identified and compressed, which suppresses small-scale power while leaving the large scale power essentially unchanged.

FIG. 22: Effect of finger-of-god (FOG) compression. Raising the FOG compression threshold $\delta_c$ means that fewer FOGs are identified and compressed, which suppresses small-scale power while leaving the large scale power essentially unchanged.

Our above-mentioned 4-parameter fits give highly stable best-fit values $\Omega_m = (0.244, 0.242, 0.243)$ and $f_b = (0.168, 0.169, 0.168)$ together with the strongly varying best-fit values $Q_{nl} = (27.0, 30.9, 34.2)$. If we fix the baryon density at the best fit WMAP3 value and vary only the three parameters $(\Omega_m, b, Q_{nl})$, the corresponding results are $\Omega_m = (0.246, 0.243, 0.244)$ and $Q_{nl} = (27.1, 31.0, 34.3)$. Note that the cosmological parameter values do not show a rising or falling trend with $\delta_c$. For comparison, the $1\sigma$ uncertainty on $\Omega_m$ from Table 2 is $\Delta \Omega_m \approx 0.02$, an order of magnitude larger than these variations. In other words, the $Q_{nl}$-parameter closely emulates the effect of changing $\delta_c$, so that marginalizing over $Q_{nl}$ is tantamount to marginalizing over $\delta_c$, making our treatment rather robust to the modeling of nonlinear redshift distortions.

b. Sensitivity to k-cutoff. This is all very reassuring, showing that our cosmological constraints are almost completely unaffected by major changes in the $k \gtrsim 0.1 h/\text{Mpc}$ power spectrum. (The reason that we nonetheless perform the $Q_{nl}$-marginalization is if course that we wish to immunize our results against any small nonlinear corrections that extend to $k \lesssim 0.1 h/\text{Mpc}$.) To further explore this insensitivity to nonlinearities, we repeat the above analysis for the default $\delta_c = 200$ case, including measurements for $0.01 h/\text{Mpc} \leq k \leq k_{\text{max}}$, and vary the upper limit $k_{\text{max}}$. We apply a prior $0 \leq Q_{nl} \leq 50$ to prevent unphysical $Q_{nl}$-values for small $k_{\text{max}}$-values (where $Q_{nl}$ becomes essentially unconstrained). If no nonlinear modeling is performed, then as emphasized in [43], the recovered value of $\Omega_m$ should increase with $k_{\text{max}}$ as nonlinear effects become important. In contrast, Figure 23 shows that with our nonlinear modeling, the recovered $\Omega_m$-value is strikingly insensitive to $k_{\text{max}}$. For $k_{\text{max}} \ll 0.07 h/\text{Mpc}$, the constraints are weak and fluctuate noticeably as each new band power is included in the analysis. The yellow band shows the result when marginalizing over the baryon density $\Omega_b$, the thinner cyan/grey band shows the result when fixing $\Omega_b$ at the best-fit WMAP3 value.
cluded, but for $k_{\text{max}}$ beyond the first baryon bump at $k \sim 0.07h/$Mpc, both the central value and the measurement uncertainty remain essentially constant all the way out to $k_{\text{max}} = 0.2h/$Mpc.

The above results tells us that, to a decent approximation, our $k \geq 0.1$/Mpc data are not contributing information about cosmological parameters, merely information about $Q_{\text{nl}}$. Indeed, the error bar $\Delta Q_{\text{nl}}$ is larger when using $k < 0.2h/$Mpc data and marginalizing over $Q_{\text{nl}}$ then when using merely $k < 0.09h/$Mpc data and fixing $Q_{\text{nl}}$. In other words, our cosmological constraints come almost entirely from the LRG power spectrum shape at $k \lesssim 0.1h/$Mpc.

c. Comparison with other galaxy $P(k)$-measurements

Let us conclude this section by briefly comparing with $\Omega_m$-values obtained from other recent galaxy clustering analyses.

Our WMAP3+LRG measurement $\Omega_m = 0.24 \pm 0.02$ has the same central as that from WMAP3 alone [7], merely with a smaller error bar, and the most recent 2dFGRS team analysis also prefers $\Omega_m \approx 0.24$ [37]. This central value is 1.5$\sigma$ below the result $\Omega_m = 0.30 \pm 0.04$ reported from WMAP1 + SDSS main sample galaxies in [33]; part of the shift comes from the lower third peak in WMAP3 as discussed in Section IV B. Post-WMAP3 results are also consistent with ours. Analysis of an independent SDSS LRG sample with photometric redshifts gave best-fit $\Omega_m$-values between 0.26 and 0.29 depending on binning [41], while an independent analysis including acoustic oscillations in SDSS LRGs and main sample galaxies preferred $\Omega_m \approx 0.256$ [18].

The galaxy power spectra measured from the above-mentioned data sets are likely to be reanalysed as nonlinear modeling methods improve. This makes it interesting to compare their statistical constraining power. [41] do so by comparing the error bar $\Delta \Omega_m$ from fitting two parameter ($\Omega_m, b$)-models to all $k \leq 0.2h/$Mpc data, with all other parameters, including $Q_{\text{nl}}$ or other nonlinear modeling parameters, fixed at canonical best fit values. This gives $\Delta \Omega_m \approx 0.020$ for 2dFGRS and $\Delta \Omega_m \approx 0.012$ for the SDSS LRG sample with photometric redshifts [41]. Applying the same procedure to our LRGs yields $\Delta \Omega_m = 0.007$. This demonstrates both the statistical power of our sample, and that our cosmological analysis has been quite conservative in the sense of marginalizing away much of the power spectrum information (marginalizing over $Q_{\text{nl}}$ doubles the error bar to $\Delta \Omega_m = 0.014$).

3. Other issues

A fortunate side effect of improved cosmological precision is that priors now matter less. Monte Carlo Markov Chain generators usually assume a uniform Bayesian prior in the space of its “work parameters”. For example, if two different papers parametrize the fluctuation amplitude with $A_s$ and $\ln A_s$, respectively, they implicitly assign $A_s$-priors that are constant and $\propto 1/A_s$, respectively (the new prior picks up a factor from the Jacobian of the parameter transformation). Such prior differences could lead to substantial ($\sim 1\sigma$) discrepancies on parameter constraints a few years ago, when some parameters were still only known to a factor of order unity. In contrast, Table 2 shows that most parameters are now measured with relative errors in the range $1\% - 10\%$. As long as these relative measurement errors are $\ll 1$, such priors become unimportant: Since the popular reparametrizations in the literature and in Table 2 involve smooth functions that do not blow up except perhaps where parameters vanish or take unphysical values, the relative variation of their Jacobian across the allowed parameter range will be of the same order as the relative variation of the parameters ($\ll 1$), i.e., approximately constant. Chosing a uniform prior across the allowed region in one parameter space is thus essentially equivalent to choosing a uniform prior across the allowed region of anybody else’s favorite parameter space.

V. CONCLUSIONS

We have measured the large-scale real-space power spectrum $P(k)$ using luminous red galaxies in the Sloan Digital Sky Survey (SDSS) with narrow well-behaved window functions and uncorrelated minimum-variance errors. The results are publicly available in an easy-to-use form at http://space.mit.edu/home/tegmark/sdss.html.

This is an ideal sample for measuring the large-scale power spectrum, since its effective volume exceeds that of the SDSS main galaxy sample by a factor of six and that of the 2dFGRS by an order of magnitude. Our results are robust to omitting purely angular and purely radial density fluctuations and are consistent between different parts of the sky. They provide a striking model-independent confirmation of the predicted large-scale $\Lambda$CDM power spectrum. The baryon signature is clearly detected (at $3\sigma$), and the acoustic oscillation scale provides a robust measurement of the distance to $z = 0.35$ independent of curvature and dark energy assumptions.

Although our measured power spectrum provides independent cross-checks on $\Omega_m$ and the baryon fraction, in good agreement with WMAP, its main utility for cosmological parameter estimation lies in complementing CMB measurements by breaking their degeneracies; for example, Table 3 shows that it cuts error bars on $\Omega_m$, $\omega_m$, and $h$ by about a factor of two for vanilla models (ones with a cosmological constant and negligible curvature, tensor modes, neutrinos and running spectral index) and by up to almost an order of magnitude when curvature, tensors, neutrinos or $w$ are allowed. We find that all these constraints are essentially independent of scales $k > 0.1h/$Mpc and associated nonlinear complications.

Since the profusion of tables and figures in Section IV can be daunting to digest, let us briefly summarize them and discuss both where we currently stand regarding cos-
mological parameters and some outstanding issues.

A. The success of vanilla

The first obvious conclusion is that “vanilla rules OK”. We have seen several surprising claims about cosmological parameters come and go recently, such as a running spectral index, very early reionization and cosmologically detected neutrino mass — yet the last two rows of Table 3 show that there is no strong evidence in the data for any non-vanilla behavior: none of the non-vanilla parameters reduces $\chi^2$ significantly relative to the vanilla case. The WMAP team made the same comparison for the CMB-only case and came to the same conclusion [7]. Adding a generic new parameter would be expected to reduce $\chi^2$ by about unity by fitting random scatter. Although WMAP alone very slightly favor spatial curvature, this preference disappears when SDSS is included. The only non-vanilla behavior that is marginally favored is running spectral index $\alpha < 0$, although only at 1.6$\sigma$. This persistent success of the vanilla model may evoke disturbing parallels with the enduring success of the standard model of particle physics, which has frustrated widespread hopes for surprises. However, the recent evidence for $n_s < 1$ represents a departure from the $n_s = 1$ “vanilla lite” model that had been an excellent fit ever since COBE [149], and as we discuss below, there are good reasons to expect further qualitative progress soon.

B. Which assumptions matter?

When quoting parameter constraints, it is important to know how sensitive they are to assumptions about both data sets and priors. The most important data assumptions discussed in Section IV F are probably those about polarized CMB foreground modeling for constraining $\tau$ and those about nonlinear galaxy clustering modeling for constraining the power spectrum shape. The effect of priors on other parameters is seen by comparing the seven columns of Table 3, and the effect of including SDSS is seen by comparing odd and even rows.

WMAP alone has robustly nailed certain parameters so well that that neither adding SDSS information nor changing priors have any significant effect. Clearly in this camp are the baryon density $\omega_b$ (constrained by WMAP even-odd peak ratios) and the reionization optical depth $\tau$ (constrained by WMAP low-$\ell$ E-polarization); indeed, Table 1 in [39] shows that adding LyOF and other CMB and LSS data does not help here either. The spectral index $n_s$ is also in this nailed-by-WMAP category as long as we assume that $\alpha$ is negligible; generic slow-roll inflation models predict $|\alpha| \lesssim 10^{-3}$, well below the limits of detectability with current data sets.

For many other parameters, e.g., $\Omega_m$, $h$ and $t_{\text{now}}$, the WMAP-only constraints are extremely sensitive to priors, with the inclusion of SDSS information tightening them by factors 2 - 10. The prior assumptions of the vanilla model ($\Omega_k = r = f_{\nu} = \alpha = 0$, $w = -1$) matter a lot with WMAP alone, and when one of them is dropped, the best fit values of $\Omega_k$ and $h$ are typically very different, with much larger errors. These assumptions no longer matter much when SDSS is included, greatly simplifying the caveat list that the cautious cosmologist needs to keep in mind. This is quite different from the recent past, when the joint constraints from older WMAP and SDSS data were sensitive to prior assumptions such as spatial flatness [33]; a major reason for this change is clearly the SDSS measurement of the baryon acoustic scale. Indeed, one of the most interesting results of our analysis is the strengthened evidence for a flat universe, with the constraint on $\Omega_{\text{tot}}$ tightening from 1.054$^{+0.064}_{-0.046}$ (WMAP3 only) to 1.003$^{+0.010}_{-0.009}$ (WMAP3+SDSS).

In other words, large-scale cosmic clustering data now robustly constrain all the vanilla parameters, even when any one of $(f_{\nu}, \Omega_k, r, f_{\nu}, w)$ are included as in Table 3. If $w$ is varied jointly with $\Omega_k$ (as it arguably should be [138]), one expects dramatically weakened constraints on the two (since two standard rulers cannot determine the three parameters $(w, \Omega_k, \Omega_m)$), but rather unaffected degradation for the rest.

C. Other data

Our cosmological parameter analysis has been very conservative, using the bare minimum number of data sets (two) needed to break all major degeneracies, and using measurements which mainly probe the large-scale linear regime. It is therefore interesting to compare our results with the complementary approach of [39] of pushing the envelope by using essentially all available data (including LyOF, supernovae Ia and smaller-scale CMB experiments), which gives tighter constraints at the cost of more caveats. Comparing with the error bars in Table 1 of [39] shows that the additional data give merely modest improvements for $(\omega_b, \omega_g, n_s, r, h)$, a halving of the error bars on $\Omega_{\text{tot}}$ (still consistent with flatness), and great gains for $\alpha$ and $M_\nu$. These last two parameters are strongly constrained by the small-scale LyOF information, with [39] reporting $\alpha = -0.015 \pm 0.012$ and $M_\nu/3 < 0.06$ eV (95%), a factor of six below our constraint and bumping right up against the atmospheric lower bound $\sim 0.04$ eV. On the other hand, the same analysis also rules out the standard model with three active neutrino species at $2.5\sigma$ [39]; one can always worry about pushing the envelope too far by underestimating modeling uncertainties and systematics. [39] also highlight interesting tension at the $2\sigma$-level between the LyOF and WMAP3 data regarding the fluctuation amplitude $\sigma_8$, and weak gravitational lensing may emerge as the decisive arbiter here, by directly pinning down the matter fluctuation amplitude independently of bias [174, 175].
D. Future challenges

The impressive improvement of cosmological measurements is likely to continue in coming years. For example, the SDSS should allow substantially better cosmological constraints from LRGs for several reasons. When the SDSS-II legacy survey is complete, the sky area covered should be about 50% larger than the DR4 sample we have analyzed here, providing not only smaller error bars, but also narrower window functions as the gaps in Figure 3 are filled in. Global photometric calibration will be improved [176]. Various approaches may allow direct measurements of the bias parameter $b$, e.g., galaxy lensing [178], higher-order correlations [177], halo luminosity modeling [179] and reionization physics [180]. A bias measurement substantially more accurate than our 11% constraint from redshift space distortions would be bias measurement substantially more accurate than our cosmology modeling [179] and reionization physics [180]. A bias measurement substantially more accurate than our 11% constraint from redshift space distortions would be bias measurement substantially more accurate than our cosmology modeling [179] and reionization physics [180]. A bias measurement substantially more accurate than our cosmology modeling [179] and reionization physics [180].

As a result of such data progress in many areas, parameter constraints will clearly keep improving. How good is good enough? The baryon density $\omega_b$ is a parameter over which it is tempting to declare victory and move on: The constraints on it from cosmic clustering are in good agreement, and are now substantially tighter than those from the most accurate competing technique against which it can be cross-checked (namely Big Bang nucleosynthesis), and further error bar reduction appears unlikely to lead to qualitatively new insights. In contrast, there are a number of parameters where cosmic clustering constraints are only now beginning to bump up against theory and other measurements, so that further sensitivity gains give great discovery potential. We have $(n_s, r, \alpha, \Omega_b)$ to test inflation, $M_\nu$ to cosmologically detect neutrino mass, $w$ and more generally $X(z)$ to constrain dark energy, and $\sigma_8$ to resolve tension between different cosmological probes.

Cosmology has now evolved from Alan Sandage’s “search for two numbers” $(h, \Omega_m)$ to Alan Alexander Milne’s “Now we are six” $(h, \Omega_m, \Omega_c, \sigma_8, n_s, \tau)$. Each time a non-trivial value was measured for a new parameter, nature gave up a valuable clue. For example, $\Omega_c > 0$ revealed the existence of dark matter, $\Omega_A > 0$ revealed the existence of dark energy and the recent evidence for $n_s < 1$ may sharpen into a powerful constraint on inflation. Milestones clearly within reach during the next few years include a measurement of $n_s < 1$ at high significance and $M_\nu > 0$ from cosmology to help uncover the neutrino mass hierarchy. If we are lucky and $r \sim 0.1$ (as suggested by classic inflation and models such as [152]), an $r > 0$ detection will push the frontier of our ignorance back to $10^{-15}$ s and the GUT scale. Then there is always the possibility of a wild surprise such as $\Omega_{\text{tot}} \neq 1$, large $|\alpha|$, $X(z) \neq 1$, demonstrable non-Gaussianity, isocurvature contributions, or something totally unexpected. Our results have helped demonstrate that challenges related to survey geometry, bias and potential systematic errors can be overcome, giving galaxy clustering a valuable role to play in this ongoing quest for greater precision measurements of the properties of our universe.

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FIG. 24: 95% constraints in the $(\Omega_m, Q_{nl})$ plane for vanilla models. The shaded regions are ruled out by WMAP1 (red/dark grey), WMAP3 (orange/grey) and when adding SDSS LRG information.
APPENDIX A: POWER SPECTRUM ESTIMATION DETAILS

1. Relation between methods for measuring the power spectrum and correlation function

In this section, we clarify the relationship between different popular techniques for quantifying galaxy clustering with pair-based statistics, including correlation function estimation with the “DD-2DR+RR” method [52, 53] and power spectrum estimation with the FKP [19], FFT [26, 29, 30, 38, 43] and PKL [23, 24, 27, 28, 34, 58] methods.

Suppose we have $N_d$ data points giving the comoving redshift space position vectors $r_i$ of galaxies numbered $i = 1, N_d$, and $N_r$ random points $s_i$ from a mock catalog which has the same selection function $\bar{n}(r)$ as the real data. The number densities of data points and random points are then sums of Dirac $\delta$-functions:

$$n_d(r) = \sum_{i=1}^{N_d} \delta(r - r_i),$$  \hspace{1cm} (A1)

$$n_r(r) = \sum_{i=1}^{N_r} \delta(r - s_i).$$  \hspace{1cm} (A2)

By definition of the selection function $\bar{n}(r)$, the quantity

$$\hat{\delta}(r) = \frac{n_d(r) - \alpha n_r(r)}{\bar{n}(r)},$$

where $\alpha = N_d/N_r$, is then an unbiased estimator of the underlying density fluctuation field $\delta(r)$ in the sense that $\langle \hat{\delta} \rangle = \delta$, where the averaging is over Poisson fluctuations as customary. Except for the PKL method, all techniques we will discuss take the same general form, weighting galaxy pairs in a form that depends only on the position of each galaxy and on the distance between the two, so we will now describe them all with a unified notation. (For an even more general pair-weighting formalism that also incorporates the PKL method, see [181].) As long as one uses $N_r \gg N_d$ random points, they will contribute negligible Poisson noise; their role is in effect to evaluate certain cumbersome integrals by Monte Carlo integration.

Let us define the quantity

$$\hat{\xi}(f) \equiv \int \int w(r) \hat{\delta}(r) w(r') \hat{\delta}(r') f(|r - r'|) d^3rd^3r'.$$  \hspace{1cm} (A4)

Here $w(r)$ and $f(d)$ are the above-mentioned weight functions that depend on position and distance, respectively. As we will see, the “DD-2DR+RR”, FKP and FFT methods simply correspond to different choices of $w$ and $f$. Substituting equations (A1)-(A3) into equation (A4), we find that

$$\hat{\xi}(f) = \hat{\xi}_{dd}[f] - 2\hat{\xi}_{dr}[f] + \hat{\xi}_{rr}[f],$$  \hspace{1cm} (A5)

where we have defined

$$\hat{\xi}_{dd}[f] = \sum_{i=1}^{N_d} \sum_{j=1}^{N_d} w(r_i) w(r-j) f(|r_i - r-j|),$$  \hspace{1cm} (A6)

$$\hat{\xi}_{dr}[f] = \alpha \sum_{i=1}^{N_d} \sum_{j=1}^{N_r} w(r_i) w(s-j) f(|r_i - s-j|),$$  \hspace{1cm} (A7)

$$\hat{\xi}_{rr}[f] = \alpha^2 \sum_{i=1}^{N_r} \sum_{j=1}^{N_r} w(s_i) w(s-j) f(|s_i - s-j|).$$  \hspace{1cm} (A8)

As a first example, let us consider the FKP method [19]. This corresponds to [181]

$$f(d) = j_0(kd),$$  \hspace{1cm} (A9)

$$w(r) \propto \frac{\bar{n}(r)}{1 + \bar{n}(r)P(k)},$$  \hspace{1cm} (A10)

and turns $\hat{\xi}$ into the FKP estimator of the window-convolved power spectrum $P(k)$. Here $j_0(x) \equiv \sin(x)/x$, $w$ is normalized so that $\int w(r)^2 d^3r = 1$ and $P$ is an a priori guess as to what the galaxy power spectrum is. For details, see [181] around equations (25) and (56). The main point is that Fourier transforming $\hat{\delta}$ and averaging $|\hat{\delta}(k)|^2$ over a spherical shell in $k$-space gives the factor $\int e^{-ik \cdot |r - r'|} d\Omega_k/4\pi = j_0(k(|r - r'|)) = f$. We apply this method to our LRG data and compare the results with those of [43] in Figure 25, finding good agreement.

The FFT method [26, 29, 30, 38, 43] is identical to the FKP method except for two simplifications: $P$ in equation (A10) is taken to be a $k$-independent constant and the density field is binned onto a three-dimensional grid to replace the time-consuming double sums above with a fast Fourier transform.

The “DD-2DR+RR” method [52, 53] estimates the correlation function $\xi(r)$ by the Landy-Szalay estimator

$$\hat{\xi}_{LS} = \frac{\hat{\xi}_{dd} - 2\hat{\xi}_{dr} + \hat{\xi}_{rr}}{\hat{\xi}_{rr}}.$$  \hspace{1cm} (A11)
** FIG. 25:** Comparison of power spectrum estimation techniques. Our FKP measurement without defogging is seen to agree quite well with the measurement of [43] considering that the latter includes also main sample galaxies with different \( \beta \) and small-scale clustering properties. These curves cannot be directly compared with the PKL measurements or theoretical models, because they are not corrected for the effects of redshift distortions, window function, and small-scale clustering properties. These curves cannot be directly compared with the PKL measurements or theoretical models, because they are not corrected for the effects of redshift distortions, window functions and the integral constraint; the qualitative agreement that is nonetheless seen is as good as one could expect given these caveats.

which is often written informally as \((DD - 2DR + RR)/RR\). Here two common weighting choices in the literature are \(w(r) = \bar{n}(r)\) \([52]\) and \(w(r) = \bar{n}(r)/[1 + \bar{n}(r)J]\) \([53]\), where \(J \equiv \int_0^r \xi(r')d^3r'\) tends to be of the same order of magnitude as \(P(k)\). To measure the binned correlation function using equations (A6)-(A8), one thus sets \(f(d) = 1\) when \(d\) is inside the bin and \(f(d) = 0\) otherwise.

These close relationships between the FKP, FFT and “DD-2DR+RR” methods lead to interesting conclusions regarding all three methods.

First, it can be interesting for some applications to replace \(J\) by \(P\) when measuring the correlation function, using \(w(r) = \bar{n}(r)/[1 + \bar{n}(r)P]\), as was done for the analysis of the QDOT survey in [19] and for the LRG analysis in [36]. For instance, one could use a constant \(P\) evaluated at the baryon wiggle scale if the goal is to measure the baryon bump in the correlation function.

Second, there is an interesting equivalence between the methods. For reasons that will become clear below, let us refer to the numerator of equation (A11), \(\hat{\xi}_{dd} - 2\hat{\xi}_{dr} + \hat{\xi}_{rr}\), as the “convolved” correlation function estimator and full expression \(\xi_{LS}\) as the “deconvolved” estimator. The information content in the convolved and deconvolved estimators is clearly the same, since dividing by \(\hat{\xi}_{rr}\) in equation (A11) is a reversible operation. Moreover, it is straightforward to show that the FKP estimator of \(P(k)\) is simply the 3D Fourier transform of the convolved correlation function estimator as long as the same weighting function \(w(r)\) is used for both. [182] also comment on this. (Note that this is a quite different statement from the well-known fact that \(P(k)\) is the 3D Fourier transform of the correlation function \(\xi(r)\).) This implies that the measured FKP power spectrum and the measured correlation function contain exactly the same information. In particular, it means that cosmological constraints from one are no better than cosmological constraints from the other, since they should be identical as long as window functions, covariance matrices, etc., are handled correctly. (An analogous correspondence for purely angular data is discussed in [183].) In contrast, the information content in the PKL measurement of the power spectrum is not identical; it uses a more general pair weighting than equation (A4) and by construction contains more cosmological information; a more detailed discussion of this point is given in Appendix A.3 in [184].

Third, this Fourier equivalence between the convolved correlation function estimator and the FKP power spectrum estimator sheds light on the fact that the deconvolved correlation function estimator \(\hat{\xi}_{LS}\) is unbiased \((\langle \hat{\xi}_{LS}(d) = \xi(d)\), the true correlation function), whereas the expectation value of the FKP estimator is merely the true power spectrum convolved with a so-called window function. This difference stems from the division by \(\hat{\xi}_{rr}\) in equation (A11): Multiplication by \(\hat{\xi}_{rr}\) in real space corresponds to convolution with the Fourier transform of \(\hat{\xi}_{rr}\) (the window function) in Fourier space. The reason that one cannot deconvolve this windowing in Fourier space is that one cannot Fourier transform \(\hat{\xi}_{LS}\), as it is completely unknown for large \(d\)-values that exceed all pair separations in the survey.

Fourth, this equivalence implies that gridding errors in the 3D FFT method (which become important at large \(k\) [30]) can be completely eliminated by simply computing the correlation function with \(w(r) = \bar{n}(r)/[1 + \bar{n}(r)P]\) by summation over pairs and then transforming the convolved correlation function with the kernel \(j_0(kr)\).

Figure 25 compares the LRG power spectra measured with the different techniques discussed above. A direct comparison between our PKL \(P(k)\)-measurement and that of [43] is complicated both by window function effects and by the fact that the latter was performed in redshift space without FOG compression, with SDSS MAIN galaxies mixed in with the LRG sample. To facilitate comparison, we performed our own FKP analysis using the direct summation method as described above, with constant \(P = 30000(h^{-1}\text{Mpc})^3\) and \(\alpha \approx 0.06\). This is seen to agree with the measurement of [43] to within a few percent for \(0.04h/\text{Mpc} < k < 0.2h/\text{Mpc}\) for the case of no defogging, with the remaining differences presumably due mainly to the inclusion of main-sample galaxies,
particularly on small scales where nonlinear behavior becomes important. Figure 25 also shows that our defogged FKP measurements agree qualitatively between the PKL and FKP techniques, and that the FKP power spectrum continues to track our nonlinear WMAP model beautifully all the way out to \( k = 1h/\text{Mpc} \) even though \( Q_{\text{nl}} \) was only fit to the \( k < 0.2h/\text{Mpc} \) PKL data.

An important caveat must be borne in mind when interpreting Figure 25: The PKL points are constructed in such a way as to allow direct visual comparison with a model power spectrum [28], but the FKP and [43] curves are not, and should not be expected to fall right on top the PKL points or the best-fit cosmological model because they are not corrected for the effects of redshift distortions, window functions and the integral constraint. Redshift distortions should boost the FKP LRG curves slightly above the true real space power spectrum (see Section A 3), and should boost the curve from Figure 25 slightly more because the main sample galaxies have a slightly more because the main sample galaxies have a higher \( \beta \) than the LRGs. The FKP window functions are broader than their PKL counterparts, and the steeper the power spectrum is, the more power leaks in from larger scales, causing the plotted measurements to lie above the true power spectrum. Finally, the integral constraint suppresses the plotted FKP power on the largest scales.

In conclusion, the agreement seen in Figure 25 is as good as one could expect given these many caveats.

2. Numerical acceleration of the PKL method

In this section, we describe a numerical improvement over the PKL power spectrum estimation method described in [28] that enables us to increase the number of modes from 4000 to 42,000.

The cosmological information content in a galaxy redshift survey, quantified by the Fisher information matrix [46, 47, 57], scales approximately as the effective volume \( V_{\text{eff}} \) defined in equation (1), with error bars on cosmological parameters optimally measured from the survey scaling as \( V_{\text{eff}}^{-1/2} \). However, actually extracting all this information in a numerically feasible way is far from trivial, contributing to the extensive literature on power spectrum estimation methods.

Our PKL method expands the galaxy density field in \( N \) functions (“PKL modes”) that probe successively smaller scales, and the number of modes needed to retain all information down to some length scale \( \lambda = 2\pi/k \) is clearly of order \( V_{\text{eff}}/\lambda^3 \). In [28], \( N = 4000 \) modes were used, and it was empirically determined that this retained essentially all information for \( k \leq 0.1h/\text{Mpc} \) with a gradual tapering off towards smaller scales. This was a convenient coincidence, since using \( N \gg 4000 \) becomes numerically painful: because of the many \( N \times N \) matrix operations involved in the analysis, the disk usage is about 80 GB times \( (N/4000)^2 \) and the CPU time required on a current workstation is about 20 days times \( (N/4000)^3 \).

The effective volume of our LRG sample is about ten times larger than that of the above-mentioned main galaxy analysis because the sky area covered has increased and because the sample is significantly deeper. To extract all the \( k \lesssim 0.1h/\text{Mpc} \) information, we would therefore like to use about ten times more modes, but without the analysis taking \( 10^3 \) times longer (~ 50 years).

We therefore combine the method of [33] with a divide-and-conquer approach, performing a separate 2000-mode analysis on each of the 21 sub-volumes described in Section II (3 radial \( \times \) 7 angular subsets) and combining the results with minimum-variance weighting (which, following the notation of [28], corresponds to simply summing both the \( \mathbf{F} \)-matrices and the \( \mathbf{q} \)-vectors). Although this combined analysis with its \( 21 \times 2000 = 42,000 \) modes becomes lossless in the information theory sense on scales substantially smaller than each of the 21 sub-volumes, it destroys most of the information on scales comparable to these volumes, because the mean density in each volume is projected out (effectively marginalized over) [28]. It also becomes suboptimal on these largest scales because it neglects correlations between different sub-volumes when optimizing the pair weighting. We therefore complement the combined analysis with a 4000-mode global analysis of the entire volume, which is optimal on the largest scales.

Both of these analyses produce uncorrelated band power estimators, and we use the first 8 (with \( k < 0.04h/\text{Mpc} \)) from the global analysis and the remaining ones from the combined analysis. This splice point was chosen because the Fisher matrices show that the global analysis contains the most information (gives the smallest power spectrum error bars) for smaller \( k \), and the combined analysis contains the most information for larger \( k \). For the radial subsamples, the corresponding splice points are after bands 11 (NEAR), 10 (MID) and 8 (FAR). We confirm that, as the above scaling arguments suggest, the two analyses give essentially identical results in the intermediate \( k \)-range where they both retain virtually all the information. For example, the two analyses agree for band number 9 to about 0.7% in power, a difference which is completely negligible compared to the statistical error bars.

3. Redshift space distortion details

As described in detail in [28], our PKL method produces three estimators \( (\hat{P}_{\text{gg}}(k), \hat{P}_{\text{gv}}(k), \hat{P}_{\text{vv}}(k)) \) of the galaxy-galaxy, galaxy-velocity and velocity-velocity power spectra \( (P_{\text{gg}}(k), P_{\text{gv}}(k), P_{\text{vv}}(k)) \). These estimators are uncorrelated, both with each other and between different \( k \)-bands, but not unbiased: the expectation value of \( \hat{P}_{\text{gg}}(k) \), say, includes contributions from all three power spectra. As explained in [28], we therefore construct our final power spectrum estimator \( \hat{P}_g \) as a linear combination of \( \hat{P}_{\text{gg}}(k), \hat{P}_{\text{gv}}(k) \) and \( \hat{P}_{\text{vv}}(k) \) that makes it an an unbiased estimator of the real-space galaxy power spec-
trum $P_{gg}(k)$. This linear combination corresponds to the process of marginalizing over the relative amplitudes of $P_{gg}(k)$ and $P_{vv}(k)$, which according to equations (2) and (3) are $\beta r_{gv}$ and $\beta^2$, respectively, so it can also be thought of as a marginalization over $\beta$ and $r_{gv}$.

Two ways of forming this linear combination were explored in [28], referred to as the modeling method and the disentanglement method, respectively. The former corresponds to marginalizing over $\beta$ and $r_{gv}$ globally, treating them as scale-independent constants, whereas the latter corresponds to treating them as arbitrary functions of $k$ and marginalizing over them separately for each $k$-band. We used the former approach for the “official” $P(k)$-measurement in [28] that was used for cosmological parameter estimation, and we make the same choice in the present paper, using only $k < 0.09h$/Mpc data to find the best-fitting values $(\beta, r_{gv}) \approx (0.3, 1)$. The latter approach is more conservative, at the price of producing much larger error bars.

To facilitate the interpretation of our thus-measured power spectrum $\hat{P}_k(k)$, it is helpful to re-express it in terms of multipoles of the redshift space power spectrum. In the small-angle (distant observer) approximation where all galaxy pairs subtend a small angle relative to the line of sight, $(P_{gg}, P_{gv}, P_{vv})$ reduce to simple linear combinations of the monopole, quadrupole and hexadecapole power spectra in redshift space [185, 186]:

\[
\begin{bmatrix}
P_{gg}(k) \\
P_{gv}(k) \\
P_{vv}(k)
\end{bmatrix}
= \begin{bmatrix}
1 - \frac{1}{2} & \frac{3}{8} \\
\frac{3}{8} & -\frac{15}{64} \\
0 & \frac{35}{64}
\end{bmatrix}
\begin{bmatrix}
P_0(k) \\
P_2(k) \\
P_4(k)
\end{bmatrix}. \tag{A12}
\]

Inverting equation (A12) gives

\[
\begin{bmatrix}
P_0(k) \\
P_2(k) \\
P_4(k)
\end{bmatrix}
= \begin{bmatrix}
1 - \frac{2}{3} & \frac{1}{4} \\
\frac{1}{3} & \frac{4}{7} \\
0 & \frac{7}{4}
\end{bmatrix}
\begin{bmatrix}
P_{gg}(k) \\
P_{gv}(k) \\
P_{vv}(k)
\end{bmatrix}. \tag{A13}
\]

Equation (A12) tells us that, in the small angle approximation, the disentanglement method would correspond to measuring $\langle \hat{P}_k(k) \rangle = P_0(k) - \frac{1}{2} P_2(k) + \frac{3}{8} P_4(k) = P_{gg}(k)$. The corresponding weights for the modeling method are found by minimizing the variance among the class of all unbiased estimators, and thus depend on the detailed survey geometry, the shot noise level, etc. Empirically, we find $\langle \hat{P}_0(k) \rangle \approx 0.8 P_0(k) - 0.07 P_2(k) + 0.006 P_4(k)$, with the weights roughly independent of $k$. This can be intuitively understood from the fact that the estimators of $P_2$ and $P_4$ are much noisier than that for $P_0$, and thus get assigned low statistical weight. If $P_2$ and $P_4$ were so noisy that they were discarded altogether, only the estimator of $P_0$ would be used.

The relation $P_0(k) = (1 + \frac{3}{8} r_{gv} \beta + \frac{1}{2} \beta^2) P_{gg}(k)$ following from equation (A13) would then give the simple estimator $\hat{P}_k(k) = \hat{P}_0(k)/ (1 + \frac{3}{8} r_{gv} \beta + \frac{1}{2} \beta^2) \approx 0.8 \hat{P}_0(k)$ for $\beta = 0.3$, $r_{gv} = 1$, i.e., weights close to those we find empirically. Our measured uncertainty in this normalization factor $(1 + \frac{3}{8} r_{gv} \beta + \frac{1}{2} \beta^2)$ is about 3% (see Figure 8), in good agreement with the exact numerical calculation described in [28], and this translates into an overall 3% calibration uncertainty of our measured power spectrum which is perfectly correlated between all $k$-bands.

The fact that the quantity measured by our power spectrum estimator $\hat{P}_k(k)$ is so similar to the rescaled redshift space monopole spectrum is convenient, since it implies that nonlinear simulations of the redshift space power spectrum (as discussed in Section III D) should apply rather well to our results. However, it is important to keep in mind that our measurement $\hat{P}_k(k)$ is a more accurate estimator of $P_{gg}(k)$ than the rescaled redshift space power spectrum would be, for several reasons. First, it never resorts to the small-angle approximation. Second, full account is taken of the fact that anisotropic survey geometry can skew the relative abundance of galaxy pairs around a single point that are aligned along or perpendicularly to the line-of-sight. These two caveats matter because $P^2_2(k)$ and $P^5_4(k)$ are undefined except in the small angle limit, which could cause the correction factor $(1 + \frac{1}{2} r_{gv} \beta + \frac{1}{4} \beta^2)$ to be inaccurate on large scales. Finally, our estimator $\hat{P}_k(k)$ by construction has smaller error bars than a standard FKP estimator of the redshift space power spectrum, and one expects this advantage to be most important on the largest scales, comparable to and exceeding the thickness of the slices seen in Figure 3.

4. How spacetime geometry affects the power spectrum measurement

We performed our power spectrum analysis in comoving three-dimensional space, with the conversion of redshifts into comoving distances performed for a fiducial flat ΛCDM model with $\Omega_m = 0.25$. As described in Section IV C 1, the conversion between redshift and comoving distance (measured in $h^{-1}$Mpc) depends on the cosmological parameters ($\Omega_m, \Omega_{\text{tot}}, w$), so if a different fiducial model had been used for the conversion, then the inferred three-dimensional galaxy distribution in comoving coordinates would be radially dilated. As discussed in [36] and Section IV C 1, this would approximately dilate the dimensionless power spectrum $k^3 P(k)$ by scaling the $k$-axis by a factor

\[
a \equiv \frac{dV(z)}{dV_{\text{fid}}(z)}, \tag{A14}
\]

where $dV(z)$ is given by equation (9) and $z = 0.35$ is the median survey redshift. For the parameter range allowed by WMAP3 and our LRG data,

\[
a \approx \left( \frac{\Omega_m}{0.25} \right)^{-0.087} \left( -w\Omega_{\text{tot}} \right)^{0.19}. \tag{A15}
\]

This means that the typical correction is very small: the rms scatter in the scaling factor $a$ is 0.7% for vanilla
models, 1% for curved models and 3% for $w$-models. For example, increasing the fiducial $\Omega_m$-value by 25%, from 0.24 to 0.30, alters the scaling factor by 2% and, since the power spectrum turnover scale $\propto \Omega_m$, ignoring this correction could potentially bias the measured $\Omega_m$-value from 0.240 to 0.245.

To be conservative, we nonetheless correct for this scaling effect in our likelihood software. Reanalyzing the galaxy data with the fiducial model replaced by the one to be tested would shift the measured $P(k)$ curve up to the left on a log-log plot if $a > 1$, with $k \rightarrow k/a$ and $P \rightarrow Pa^3$. We therefore apply the opposite scaling ($k \rightarrow ka$ and $P \rightarrow P/a^3$) to the theoretically predicted power spectrum $P(k)$ before computing its $\chi^2$ against our measurement power spectrum from Table 1. We repeated our entire power spectrum analysis for $\Omega_m = 0.30$ and confirmed that this scaling is accurate. Our likelihood software, which is available at http://space.mit.edu/home/tegmark/sdss/, evaluates a exactly instead of using equation (A15).

In summary, the correction discussed in this section is quite small, especially since marginalizing over bias erases the effect of the $a^3$ amplitude shift, but we include it anyway to ensure that there is no bias on cosmological parameter estimates.

[26] W. J. Percival et al., MNRRS, 327, 1297 (2001)


