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THE ATACAMA COSMOLOGY TELESCOPE: A MEASUREMENT OF THE 600 < ℓ < 8000 COSMIC MICROWAVE BACKGROUND POWER SPECTRUM AT 148 GHz


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ABSTRACT

We present a measurement of the angular power spectrum of the cosmic microwave background (CMB) radiation observed at 148 GHz. The measurement uses maps with 1.4 angular resolution made with data from the Atacama Cosmology Telescope (ACT). The observations cover 228° of the southern sky, in a 4:2 wide strip centered on declination 53° south. The CMB at arcminute angular scales is particularly sensitive to the Silk damping scale, to the Sunyaev–Zel’dovich (SZ) effect from galaxy clusters, and to emission by radio sources and dusty galaxies. After masking the 108 brightest point sources in our maps, we estimate the power spectrum between 600 < ℓ < 8000 using the adaptive multi-taper method to minimize spectral leakage and maximize use of the full data set. Our absolute calibration is based on observations of Uranus. To verify the calibration and test the fidelity of our map at large angular scales, we cross-correlate the ACT map to the WMAP map and recover the WMAP power spectrum from 250 < ℓ < 1150. The power beyond the Silk damping tail of the CMB (ℓ ~ 5000) is consistent with models of the emission from point sources. We quantify the contribution of SZ clusters to the power spectrum by fitting to a model normalized to σ8 = 0.8. We constrain the model’s amplitude A_SZ < 1.63 (95% CL). If interpreted as a measurement of σSZ, this implies σSZ < 0.86 (95% CL) given our SZ model. A fit of ACT and WMAP five-year data jointly to a six-parameter ΛCDM model plus point sources and the SZ effect is consistent with these results.

Key words: cosmic background radiation – cosmology: observations

Online-only material: color figures
1. INTRODUCTION

The cosmic microwave background (CMB) radiation captures a view of the universe at only ~400,000 years after the big bang. The angular power spectrum of temperature anisotropies in the CMB has been crucial in developing the current standard cosmological model in which the universe today contains some 5% baryonic matter, 23% dark matter, and 72% dark energy. We refer to this model throughout as the ΛCDM model at σ8 = 0.922 ± 0.047 (1σ error bars). Recently, the South Pole Telescope (SPT) group reported σ8 = 0.773 ± 0.025 based on the power spectrum at 150 and 220 GHz (Lueker et al. 2010).

In this paper, we present a new measurement of the CMB anisotropy power spectrum in the range 600 < ℓ < 8000 (corresponding to angular scales of approximately 1.4’ to 18’). The observations were made at 148 GHz in 2008 with the Atacama Cosmology Telescope (ACT). Figure 1 (bottom panel) shows the results, which are discussed in Section 7. We briefly describe the instrument, the observations, the calibration of the data, and the map-making procedure. The dynamic range of the power spectrum is large enough that we employ new techniques in spectral estimation. We confirm that difference maps of the data give power spectra consistent with no signal and that the power spectrum is insensitive to several details of our analysis. We use the shape of the power spectrum to bound the dusty galaxy contribution and the power from SZ clusters.

2. INSTRUMENT AND OBSERVATIONS

The ACT is a 6 m, off-axis Gregorian telescope optimized for arcminute-scale CMB anisotropy measurements (Fowler et al. 2007; Hincks et al. 2009). It was installed at an elevation of 5190 m on Cerro Toco41 in the Atacama Desert of northern Chile in 2007 March. Observing conditions in the Atacama are excellent owing to the elevation, the arid climate, and the stability of the atmosphere. After all cuts, the median precipitable water vapor (PWV) was 0.49 mm during the observations presented here.

The Millimeter Bolometer Array Camera (MBAC), the current focal-plane instrument for ACT, uses high-purity silicon lenses to reimage sections of the Gregorian focal plane onto three rectangular arrays of detectors. The arrays each contain 1000 transition edge sensor (TES) bolometers. Their spectral coverage is determined by metal-mesh filters (Adc et al. 2006) having measured band centers of 148 GHz, 218 GHz, and 277 GHz. The bolometers are cooled to 300 mK by a two-stage helium sorption fridge backed by commercial pulse-tube cryocoolers. The telescope performance and control systems (Hincks et al. 2008; Switzer et al. 2008), camera design (Thornton et al. 2008; Swetz et al. 2008), detector properties (Zhao et al. 2008; Niemack et al. 2008), and readout electronics (Battistelli et al. 2008) are described elsewhere.32

2.1. Observations

ACT operated with all three arrays from mid-August to late 2008 December for the data presented here. The observing time was divided between two regions away from the galactic plane. The deepest observations cover 900 deg2 of the southern sky in a strip 8° wide centered on δ = −53° with R.A. from 19h to 24° and 0° to 7°. The analysis presented in this paper uses only data from the central 228 deg2 of the southern strip and only observations made with the 148 GHz array. The slightly elliptical beam has full width at half-maximum (FWHM) of 1.40 by 1.34 at this frequency (Hincks et al. 2009).

The observations were made by scanning the sky at a constant elevation of 50°. Each scan is 4.5 wide on the sky (7:0 in azimuth angle). Scans repeat every 10.2 s. Each half-scan consists of 4.2 s of motion at a constant speed of 1.5 s~1 followed by 0.9 s of acceleration. The first half of each night is spent observing the field rising in the eastern sky, after which ACT turns to the western sky to observe the same field as it sets through the standard elevation of 50°. The scan strategy is designed to minimize changes in the telescope’s orientation with respect to the local environment while ensuring cross-linked observations in celestial coordinates. Sky rotation ensures that all detectors sample all points in the field each night, apart from small areas at the edges.

As the telescope scans in azimuth at constant elevation, each detector is sampled at 399 Hz. The data sampling, position reading, and all housekeeping data are synchronized by a shared

31 ACT is at 22° 9586 south latitude, 67° 7875 west longitude.
32 The site http://www.physics.princeton.edu/act/ archives papers by the ACT collaboration.
is an iterative and computationally intensive process. Determined by observations of Uranus. The map-making itself is an iterative process, and when necessary, each detector’s data may be removed for the feedback loop remained locked, keeping the amplifiers in an equilibrium state. We check the detector’s response speed; look at the detector’s response to atmospheric emission with that of the CDM model and differs from the Uranus calibration by 0.7σ. Recent WMAP observations of Uranus suggest the same rescaling factor (see footnote to Section 3.2). ACT bandpowers for ℓ > 4200 have been combined into bins of Δℓ = 600 for this figure; they are given in a note to Table 1.

(A color version of this figure is available in the online journal.)

50 MHz clock; absolute times are referenced to a GPS receiver with 0.25 ns accuracy. The data are stored in continuous 15 minute segments called time-ordered data sets (TODs). Each TOD requires 1.6 GB of storage per detector array, or 600 MB after applying lossless compression.

In addition to the main survey, we perform occasional calibration measurements. Most nights, a few minutes are used to measure a planet when it passes through our standard elevation. Any given target is used only once in every three nights to minimize non-uniformity in the coverage of the CMB regions. The planetary observations allow us to measure the system’s relative and absolute responsivities, pointing, and beam profiles; they also provide a way to check for any time variations in each. A series of tuning and biasing procedures are also followed each night before and after regular CMB observations to optimize and roughly calibrate the detector response.

3. DATA REDUCTION AND MAP-MAKING

The goal of the data reduction is to estimate the maximum likelihood map of the sky. We select properly tuned detectors and calibrate their pointing and relative gains. The absolute gains are determined by observations of Uranus. The map-making itself is an iterative and computationally intensive process.

3.1. Data Selection

The first step in the data reduction is to select TODs when the receiver is cold enough for stable operation and when the PWV is less than 3 mm. Of the ≈2880 hr between 2008 August 25 and December 24, MBAC was on line for 1352 hr, roughly corresponding to the night time hours. After cutting on PWV and instrument performance, 1031 hr (35% of calendar time) went into the pipeline. Of those, 850 hr were spent observing the southern strip.

In the second step of the data reduction, we check whether each detector’s data exhibit problems that warrant removal from the final analysis (Dünner 2009). In the case of rare and transient effects, data are removed for a single detector up to a few seconds at a time; other problems are diagnosed in 15 minute intervals, and when necessary, each detector’s data may be removed for the entire TOD. We check the detector’s response speed; look for transient effects such as cosmic ray hits; and ensure that the feedback loop remained locked, keeping the amplifiers in the linear regime (Battistelli et al. 2008). Finally, we compare the detector’s response to atmospheric emission with that of the array-wide average, in each case filtering out data at frequencies above 50 MHz. Detectors are cut if their data do not have at least a 0.98 correlation with the average or if the amplitude of the atmospheric signal is not within 15% of the median...
amplitude found among good detectors. This last cut is the largest and typically removes 100 detectors. Additionally, we bin the data by scan and compute the rms, skewness, and kurtosis in each bin. A cut on rms removes an additional 30 detectors on average, while the combined cut on skewness and kurtosis removes only one additional detector. When the instrument is well tuned, ACT typically has 680 science-grade detectors at 150 GHz. The remainder are discarded from the analysis or used to monitor instrumental effects. The effective sensitivity of the array, including all instrumental and atmospheric effects, is $\sim 30 \mu K s^{1/2}$ in CMB temperature units.

### 3.2. Calibration to Planets

After we have a reliable set of detectors to use, the next step is gain calibration. In practice, the cuts are computed twice, so that the final set of cuts is based on calibrated data. We use Uranus as our absolute reference standard.

The atmosphere is an excellent continuous flat-fielding calibrator. We find that the time variation in detector relative responsivities is not significant. Therefore, the detectors’ relative responses to the atmosphere (averaged over the entire season) are used to convert each output to a common scale. Observations of Saturn confirm these relative calibrations; it is the only source observed in 2008 bright enough to use for this motivation. The Saturn data give the absolute location of the array center, which contributes negligibly to any pointing error in the sky maps.

Saturn was used as a rough check on the absolute calibration result. The Saturn observations suggest that the ACT data are 10% brighter than inferred from the Uranus data. We attribute this inconsistency to the difficulty in modeling the brightness of the planet and its rings as they vary over time and across frequencies.

We track the stability of the system over time in a number of ways. The absolute celestial calibration is checked through measurements of a planet on most nights. The conversion between raw data and units of power absorbed on the detectors is calibrated in two ways. The conversion is estimated twice nightly by modulating the TES detector bias voltage with a small additional square wave and measuring the response. A calibration is also performed at the beginning of each night by sweeping all detectors through the full range of bias voltages. Both methods are described further in Niemack (2008) and Fisher (2009). Based on several tests (Switzer 2008), we find that the gain of each detector is constant over the season to better than 2%.

As part of the calibration, the detector temporal response is deconvolved from the detector time streams. We model the response with a single time constant. The detectors’ time constants are determined from the planet measurements and bias steps (Hincks 2009). The median time constant is $t_{\text{det}} = 1.9$ ms (thus $f_{\text{sub}} = 84$ Hz), which corresponds at the ACT scan rate to $\ell \approx 31,000$. Slow detectors with $f_{\text{sub}} \lesssim 15$ Hz are not used. We also deconvolve the anti-aliasing filter imposed in the data acquisition system (Battistelli et al. 2008). We then further filter and sample the time stream at half the raw rate to speed the map-making step. This last anti-aliasing filter acts only at angular scales smaller than the beam and does not affect the maps.

### 3.3. Pointing Reconstruction

The planet Saturn is bright enough that each bolometer detects it with high signal-to-noise in a single scan. We fit in the time domain to find the best two-dimensional location for each detector relative to the notional array center. The fit produces altitude and azimuth offsets for each, along with information about the relative gains, the beam sizes, and the detector time constants. The detector pointings are consistent with optical models of the telescope and reimaging optics (Fowler et al. 2007). The relative pointing used in this work is an average over 22 observations of Saturn. The rms uncertainty is $1\pm 2$ for the detector relative pointings, which contributes negligibly to any pointing error in the sky maps.

The location of the array center can also be found from each observation of Saturn. It has a scatter of $4\pm 3$ rms over the 22 observations, which we attribute to slight thermal deformations in the telescope mirrors and their support structures.

The Saturn data give the absolute location of the array center only for other observations taken at the same horizon coordinates. To determine the absolute location of the array center during science observations, we used approximately 20 known radio sources found in preliminary maps. We find this approach both simpler and better constrained than making a complete pointing model of the telescope. The ACT southern field was observed both rising and setting, at azimuths centered at $30^\circ$ on either side of south. These preliminary maps were made separately for rising and setting data and were used to determine separate pointing corrections for the two cases. After correction, we estimate the maps to have $5''$ pointing uncertainty.

### 3.4. Map-making

The goal of the map-making step is to take the 3200 GB of cut, calibrated, and deconvolved raw data from both the rising and setting scans and produce from it a maximum likelihood estimate of the sky. This work produces a 200 MB map, 16,000 times smaller than the raw data set. We first multiply the data for each detector in each TOD by a window function, reducing the weight in the first and last 10 s of each file, then remove a single offset and slope for the entire 15 minute period. There is no additional filtering in the time domain, though the lowest frequencies are given no weight in the process of maximizing the likelihood of the final map. The detector data are combined into a data vector $d$. The mapping is done in a cylindrical equal-area projection with a standard latitude of $\delta = -53^\circ 5$ and pixels of $30'' \times 30''$, roughly one-third of the beam FWHM. The map is represented as a vector $m$ of length $N_{\text{pix}} \sim 10^7$. We model the data as $d = Pm + n$, where the matrix $P$ projects the map into the time stream and $n$ is the noise, which has covariance matrix $N$. The maximum likelihood solution, $\hat{m}$, is given by solving the mapping equation (e.g., Tegmark 1997):

$$P^T N^{-1} P \hat{m} = P^T N^{-1} d. \quad (1)$$

We solve for $\hat{m}$ iteratively using a preconditioned conjugate gradient method (Press et al. 2007; Hinshaw et al. 2007). Based on simulations, we find that the solution is an unbiased estimator of the sky for $\ell > 600$, the multipoles that we analyze in this paper. Additionally, the clear cross-correlation with WMAP (Section 3.5) indicates that the maps are likely unbiased as low as $\ell \approx 200$. This approach is different from that taken by...
other recent measurements of the fine angular scale anisotropy in which the time stream data are filtered and binned (e.g., Friedman et al. 2009; Reichardt et al. 2009a; Lueker et al. 2010).

Solving Equation (1) requires careful consideration of the noise structure of the ACT data. First, the data are weighted in Fourier space to account for the variation in noise with frequency, particularly to reduce the noise that the atmosphere adds at low frequencies. For frequencies less than 0.25 Hz, we set the statistical weights to zero. Higher frequencies are given successively more weight in inverse proportion to the noise variance in each band. Second, we find that there are several modes, particular combinations of the 1000 detectors, that correspond to signals unrelated to the celestial temperature; for example the common mode, atmospheric gradients, and other detector correlations. We handle them by finding the 10 modes with the largest eigenvalues in each TOD and solving for their amplitudes as a function of time along with the map \( \hat{m} \).

The power spectrum (Section 6) is robust to halving or doubling the number of modes.

Maps are made using only fractions of the valid data so that null tests can be performed. The subset maps are also used for finding the CMB power spectrum, as described further in Section 5.2. The final map used for source finding, however, is a complete run based on all the valid data. Figure 2 shows the map and the region used. The computational task is considerable. One iteration takes 100 s on 5000 cores of Canada’s SciNet GPC cluster each running at 2.53 GHz; computations required before the first iteration take an additional time equivalent to approximately 20 iterations. A converged map requires hundreds of iterations and approximately 10 CPU years.

3.5. Calibration to WMAP

The absolute calibration described in Section 3.2 is limited by the uncertainty in the 148 GHz brightness of Uranus, the primary calibrator. We complement the planetary calibration by cross-correlating with the W-band (94 GHz) measurements of WMAP (Hinshaw et al. 2009). In the angular scale range of 200 ≤ \( \ell \) ≤ 1200, both instruments measure the sky with sufficient signal-to-noise to permit the comparison.

The technique requires a single map from WMAP and two maps with independent noise from ACT. We assume that only one relative calibration ratio between ACT and WMAP must be estimated. The WMAP data set consists of the high-resolution W-band five-year map. The two independent ACT maps are each made with one half the current data set. We compare the maps only in the 228 deg\(^2\) strip used for the present power spectrum analysis (described in Section 5.1), where the noise is low and uniform.

We compute those cross-spectra in two-dimensional \( \ell \)-space that combine either the full ACT data set with the WMAP map or the two ACT maps with each other. We average over the polar angle in \( \ell \)-space to get a one-dimensional cross-spectrum for each. In the case of the ACT spectra, the noise is not isotropic, so the spectra are angle-averaged with appropriate weights. We then find the single calibration factor that minimizes \( \chi^2 \) for a model in which the minimum-variance weighted combination of the ACT–ACT and the ACT–WMAP cross-spectra equals the WMAP all-sky spectrum. The model accounts for the fact that the ACT and WMAP measurements have noise varying as different functions of \( \ell \).

33 The WMAP maps are at HEALPix resolution \( N_{\text{side}} = 1024 \), with 3:5 pixels.

The result is a calibration factor with <6% fractional uncertainty in temperature. The calibration factors derived from WMAP and from Uranus observations are consistent to \( \lesssim 6\% \).

4. FOREGRANDS

Radio and infrared galaxies are the dominant sources of foreground emission in the 150 GHz band at \( \ell \geq 1000 \). To study the underlying CMB spectrum, the sources must be identified and masked, and residual contamination in the power spectrum must be accounted for. Many approaches to the problem have been described in the CMB literature (e.g., Wright et al. 2009; Reichardt et al. 2009a, 2009b; Sharp et al. 2010; Dawson et al. 2006; Sievers et al. 2009). Our approach is to find the sources in the ACT maps, mask them, and assess the residual contribution with models.

We find the sources using a matched filter with noise weight- deriving from the statistical properties of the map (Tegmark & de Oliveira-Costa 1998). We identify as sources all pixels detected at \( \geq 5\sigma \) having at least three neighboring pixels detected at \( \geq 3\sigma \). A 5\( \sigma \) detection corresponds to roughly 20 mJy in the filtered map. The selection criteria are tested on simulated sky maps to determine the sample’s purity and completeness. The sample is approximately 85% complete at 20 mJy and \( \sim 100\% \) complete at 50 mJy. For fluxes greater than 20 mJy, the detections have a purity of \( \sim 95\% \) (that is, in a hundred detections approximately five are false). Simulations show that at 20 mJy deboosting (Condon 1974) is a \( \sim 9\% \) effect, and at 40 mJy a \( \lesssim 1\% \) effect.

In the 228 deg\(^2\) area, we detect 108 sources. Of these, 105 can be identified with sources in the PMN (Wright et al. 1994), SUMSS (Mauch et al. 2003), and/or AT20G (Murphy et al. 2010) radio source catalogs. One source, not in these catalogs, is in the Two Micron All Sky Survey (2MASS) catalog (and thus has IR emission). Two have no previously measured counterparts. In a 1’ search radius, 17 sources have both radio and IR identifications. With the exception of the single 2MASS source, all are treated as radio sources.

We fit the flux distribution of the radio sources to three models. The distribution agrees with the Toffolatti et al. (1998) radio model scaled by 0.49 ± 0.12 in source counts, where the uncertainty is statistical only. In the WMAP analysis of sources at 41 GHz, Hinshaw et al. (2007) found a good fit with the same model after scaling the counts by 0.64. Thus, it appears that on average the radio population at 150 GHz is composed predominantly of flat-spectrum sources. We prefer to compare the data to models by scaling the model’s flux rather than the counts, as the extrapolation of the radio flux into the 150 GHz range is typically the most uncertain part of any model. For the Toffolatti et al. (1998) model, the required flux scaling is 0.5. The sources are also well fit by the De Zotti et al. (2005) model scaled by 1.2 in flux. Lastly, we fit the radio sources model in Sehgal et al. (2010) and find that its flux should be scaled by 1 in flux. All scale factors have a statistical uncertainty of about 25%. Each of these models can be used to estimate the residual rms flux below the 20 mJy cut, as discussed in Section 7.1. We plan a more thorough exploration of the radio sources in an upcoming publication.

Before computing the power spectrum, we mask a 10’ diameter region around each of the 108 sources. We call this Mask-ACT. In total, 2.3 deg\(^2\) are masked, or 1.0% of the map. Doubling the size of the masked holes has a negligible effect on the power spectrum. We also create a cluster mask, Mask-C, by finding all 5\( \sigma \) clusters and check that using this mask combined...
with Mask-ACT has a negligible effect on the power spectrum. We also generate a mask, Mask-R, based on known radio sources measured in the SUMSS (0.8 GHz), PMN (5 GHz), and ATCA (20 GHz) catalogs. A model that has been fit to multiple data sets (Sehgal et al. 2010) is used to estimate the mean spectral index, and this index is then used to determine the flux cut in the radio catalogs. Since the model is based on ensemble properties of radio sources it may miss lower-flux low frequency sources having shallow indices. The resulting power spectrum is independent of which radio source mask we use, or if we use the union of the two, and does not depend on whether we include the cluster mask.

Diffuse dust emission becomes dominant near the galactic plane. We check for evidence of dust in the region of our map nearest to the plane, from right ascension 6 h to 7 h. Through a cross-correlation with the estimated dust map of (Finkbeiner et al. 1999, the FDS map), we can limit the dust contribution here to $\ell(\ell + 1)C_\ell/(2\pi) \lesssim 5 \mu K^2$ at $\ell = 1000$, while a direct power spectrum of the FDS map in the same region gives $\sim 1 \mu K^2$. As the diffuse dust component decreases with increasing $\ell$, it is not significant in our analysis.

5. POWER SPECTRUM METHOD

We estimate the CMB power spectrum using the adaptive multi-taper method with prewhitening, described in Das et al. (2009). We make independent maps from subsets of the data and use only cross-spectra between maps to estimate the final power spectrum. All operations are performed using the flat-sky approximation. We summarize the method in this section.

5.1. Fields Used for Power Spectrum Analysis

We find the power spectrum of our map by separate analysis of each of the 13 patches shown in Figure 2. Each patch is $4^2 \times 4^2$ in size, and together they cover a rectangular area of the map from $\alpha = 0^h48^m$ to $6^h52^m$ ($12^°$ to $103^°$) in right ascension and from $\delta = -55^°16^\prime$ to $-51^°05^\prime$ in declination. This area is the region of a larger survey having the lowest noise. We also split the raw data into four subsets of roughly equal size, with the data distributed so that any four successive nights go into different subsets. The four independent maps generated from these subsets cover the same area and have approximately the same depth. All maps are fully cross-linked. That is, they all contain data taken with the sky both rising and setting.

5.2. Spectrum of a Single Patch

We estimate the spectra of the 13 patches independently, before taking a weighted average to find the final spectrum. There are four independent sky maps, from which six cross-spectra and four auto-spectra are evaluated on each patch. We use a weighted mean of only the cross-spectra for the final spectral estimate. The weights depend on both the cross- and auto-spectra, as discussed below.

Before separating the four maps into 13 patches, each beam-convolved map, $T_b(\theta)$, is initially filtered in Fourier space with a high-pass function $F_c(\ell)$. This filter suppresses modes at large scales that are largely unconstrained. These modes arise from a combination of instrument properties, scan strategy, and atmospheric contamination. We choose a squared sine filter, given in Fourier space by the smooth function

$$F_c(\ell) = \begin{cases} 0 & : \ell < \ell_{\text{min}} \\ \sin^2 x(\ell) & : \ell_{\text{min}} \leq \ell < \ell_{\text{max}} \\ 1 & : \ell \leq \ell_{\text{max}} \end{cases}$$

where $x(\ell) \equiv (\pi/2)(\ell - \ell_{\text{min}})/\ell_{\text{max}} - \ell_{\text{min}}$. We choose $\ell_{\text{min}} = 100$ and $\ell_{\text{max}} = 500$. The 13 patches of the map are treated separately from this point. The map of each patch is prewhitened using a local, real-space operation to reduce the dynamic range of its Fourier components (Das et al. 2009). The prewhitening operation involves adding a fraction (2%) of the map to an approximation of its Laplacian. The Laplacian is computed in real space by taking the difference between the map convolved with disks of radius 1’ and 3’. Then the maps are multiplied by the point source mask. The prewhitening step greatly reduces the leakage of power from low to high multipoles caused by the action of the point source mask on the highly colored CMB power spectrum.

For each patch, we compute the six two-dimensional cross-spectra and four two-dimensional auto-spectra. The axes...
correspond to right ascension and declination. Each spectrum is computed using the adaptively weighted multi-taper method, using \( N_{\text{tap}} = 5^2 \) tapers having resolution parameter \( N_{\text{res}} = 3 \) (see Das et al. 2009). Windowing the maps with 25 orthogonal taper functions allows us to extract most of the statistical power available in the maps, at the expense of broadening the resolution in angular frequency by a factor of approximately \( N_{\text{res}} \).

At this stage, each two-dimensional cross-spectrum, \( \tilde{c}_{\ell}^{\alpha\beta} \), between submaps \( \alpha \) and \( \beta \) on patch \( i \) incorporates the effects of the filter, prewhitening, tapering, the point source mask, and the beam; they are analogous to a “pseudo power spectrum” of an apodized map (e.g., Hivon et al. 2002). Each two-dimensional spectrum is then averaged in annuli with a narrow range of \( |\ell| \) to give the binned pseudo-spectrum \( \tilde{B}_b \), with

\[
\tilde{B}_b^{\alpha\beta} = P_{\text{st}} \tilde{c}_{\ell}^{\alpha\beta}.
\]

Throughout this paper, we define \( B_i \equiv (\ell + 1)C_\ell / 2\pi \). The binning function \( P_{\text{st}} \) is set to one for pixels lying in an annular bin (indexed by \( b \)) of width \( \Delta \ell = 300 \) centered on \( |\ell| = \ell_b \), and zero elsewhere. The size of the patches and the resolution of the tapers dictates the width of the bins. For our square patches of side \( s = 4/2 \), the fundamental frequency resolution in Fourier space is \( \Delta \ell = 2\pi / s \approx 90 \). The application of the tapers with a resolution parameter \( N_{\text{res}} = 3 \) further degrades the resolution to \( N_{\text{res}} \Delta \ell \), so bins chosen to be smaller than \( \Delta \ell \approx 270 \) would be unreasonably correlated. The binning function is also set to zero where \( |\ell| < 270 \). This region of Fourier space is particularly sensitive to scan-synchronous effects, either fixed to the ground or in phase with the azimuth scan.

The binned pseudo-spectrum \( \tilde{B}_b^{\alpha\beta} \) is then deconvolved with the mode-mode coupling matrix, which takes into account the combined effects of tapering and masking and can be computed exactly. Lastly, we divide by the \( \ell \)-space representations of the prewhitening filter, of the high-pass filter \( F_c(\ell) \), and of the beam to obtain an unbiased estimate of the true underlying spectrum \( B_b^{\alpha\beta} \).

This procedure has been tested with simulations. The number and resolution of the tapers are chosen as the optimal balance between maximizing information and minimizing bias caused by leakage of power. The simulations confirm that increasing the number of tapers beyond \( 5^2 \) has a negligible effect on the spectrum errors.

5.3. Combining Patches

The final power spectrum estimator is given by a weighted mean over \( N = 13 \) patches,

\[
\tilde{B}_b = \frac{\sum_{i=1}^{N} w_i \tilde{B}_b^i}{\sum_{i=1}^{N} w_i^2},
\]

where \( \tilde{B}_b^i \equiv \sum_{\alpha,\beta} B_b^{\alpha\beta}(\ell^i) / 6 \) is the mean of the six deconvolved cross-power spectra in patch \( i \) (assuming equal weights), and \( \alpha \) and \( \beta \) index the four independent maps of that patch. The weights are chosen as the inverse of the variance of this estimator in each patch, i.e., \( w_i = 1 / \sigma^2(\tilde{B}_b^i) \), where

\[
\sigma^2(\tilde{B}_b^i) \equiv \langle (\tilde{B}_b^i)^2 \rangle - \langle \tilde{B}_b^i \rangle^2.
\]

\( \tilde{X} \) as an unbiased estimator of the quantity \( X \) in the sense \( \langle \tilde{X} \rangle = X \).

and the average is taken over the several cross-spectra computed for patch \( i \). The first term contains the four-point function of the temperature field and is approximated as

\[
\langle (\tilde{B}_b^\alpha)^2 \rangle \approx \frac{2}{M(M - 1)} \sum_{\alpha,\beta,\gamma,\delta} (B_b^{\alpha\beta}B_b^{\gamma\delta} + B_b^{\alpha\gamma}B_b^{\beta\delta} + B_b^{\alpha\delta}B_b^{\beta\gamma}).
\]

where \( M = 4 \) is the number of submaps per patch. We have neglected any connected (non-Gaussian) part of the four-point function due to components such as point sources. This is a reasonable approximation when choosing the 13 weights, because the expression (Equation (6)) is dominated by the auto-spectrum terms, which in turn are noise dominated.

5.4. Power Spectrum Covariance

We estimate the bandpower covariance matrix \( \Sigma \) using the scatter in the power spectrum among the \( N = 13 \) patches,

\[
\Sigma_{\ell \ell'} \equiv \langle \Delta B_\ell \Delta B_{\ell'} \rangle = \sum_{i=1}^{N} w_i^2 \langle (\tilde{B}_b^i - \bar{B}_b)(\tilde{B}_b^i - \bar{B}_b) \rangle.
\]

\( \langle \tilde{X} \rangle \) is given by the non-Gaussian part of the four-point function.

\[
\sigma^2(\tilde{C}_b) = \frac{2 \tilde{C}_b}{n_b} + \frac{4 \tilde{C}_b \tilde{N}_b / M + \tilde{N}_b^2 / n_w}{n_b} + \frac{\sigma_p^2}{f_{\text{sky}}},
\]

where \( \tilde{N}_b = (\tilde{C}_b - \bar{C}_b) \), the difference between the weighted mean auto- and cross-spectra, estimates the average power spectrum of the noise; \( M = 4 \) is the number of maps with independent noise properties per patch; \( n_w = M(M - 1) / 2 = 6 \) is the number of cross-spectra per patch; \( n_b \) counts the number of Fourier modes measured in bin \( b \) (that is, the number of pixels falling in the appropriate annulus of Fourier space); \( f_{\text{sky}} \) is the patch area divided by the full-sky solid angle, \( 4\pi \) steradians; and \( \bar{C}_b \) is the weighted mean cross-spectrum. In the last term, \( \sigma_p^2 \) is given by the non-Gaussian part of the four-point function. Thus,

\[
\sigma_p^2 = \frac{1}{4\pi} \left[ \langle \delta T^4(\hat{n}) \rangle - 3 \langle \delta T^2(\hat{n}) \rangle^2 \right].
\]
term does not dominate, and with bins chosen to be sufficiently large.

Figure 3 shows the three components of the errors. The Gaussian part due to sample variance dominates below $\ell = 2000$; it scales as $C_b/\sqrt{n_i}$. The Poisson term due to the non-Gaussian clusters and unresolved point sources contributes about 15% of the error budget for $2500 \lesssim \ell \lesssim 6000$ but is sub-dominant at all scales. The noise term dominates at $\ell \gtrsim 2000$; atmospheric noise is the main contribution at angular scales of $\ell \lesssim 1000$. As a cross-check of the error estimates, we also plot the errors derived from the scatter of the power spectra from the 13 patches.

The uncertainty in the analytic errors is shown as the shaded band in Figure 3. It was estimated by Monte Carlo simulations. One thousand patches were simulated with white noise and their power spectra taken by the same methods used on the ACT maps. The results demonstrate that the errors estimated from the scatter among our 13 patches are consistent with the expected uncertainty. The same simulations were used to verify that the covariance between different power spectrum bins (Equation (8)) is less than 1%.

6. POWER SPECTRUM RESULTS

The binned estimate of the power spectrum $\hat{B}_b$ is shown in Figures 1 and 4, and bandpowers are given in Table 1. With our method the bandpowers are estimated to have less than 1% correlation with neighboring bins, but the window functions have a small overlap, which we account for in the analysis. At $\ell \lesssim 2500$, the estimated power is consistent with previous observations by ground and balloon-based experiments. The features of the acoustic peaks are not distinguished with the coarse binning, but with the fluctuation bandpower measured to 5%, this spectrum offers a powerful probe of cosmological fluctuations at small scales. A clear excess of power is seen at $\ell \gtrsim 2500$ which can be attributed to point sources and, to a lesser extent, to the SZ effect.

The large-scale modes are recovered only after iterating the maps. To test for convergence of noiseless maps, we compute the spectra $\hat{C}_b^i$ of difference maps, between the processed simulation map and input simulated sky, at successive iteration numbers $i$. To account for differences that are non-uniform between data subsets, we use the auto-spectra. We show that the amplitude of fluctuations in the difference map ($\sqrt{C_b^i}$) is small, less than 1% of the amplitude of fluctuations in the input map ($\sqrt{C_b}$) at all scales by iteration 500, as shown in Figure 5. To test for convergence in the data, where we do not know the input map, we estimate the maximum change in power between the processed map at iteration $i$ and the final iteration, estimated as $2\sqrt{C_b^i/C_b}$ using auto-spectra. Here, $\hat{C}_b^i$ is the spectrum of the difference map between iteration $i$ and the final iteration, number 1000. We define the convergence ratio $r_i$ as this change in power given as a fraction of the uncertainty in the power, $\sigma(\hat{C}_b)$, and find it to be sufficiently small (less than 0.5) by iteration 500 at all scales. The cross-correlation with WMAP suggests that the maps are well converged down to $\ell \sim 200$. We do not divide the angular power spectrum by a transfer function at any value of $\ell$.

We test the isotropy of the power spectrum by estimating the power as a function of phase $\theta = \tan^{-1}(\ell_x/\ell_z)$. We compute the inverse-noise-weighted two-dimensional pseudo-spectrum co-added over map subsets and patches. The mean cross-power pseudo-spectrum is shown in Figure 6, indicating the region masked at $|\ell_x| < 270$. The spectrum is symmetric for $\ell$ to $-\ell$. 

![Image](image_url)
Table 1

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<th>$\ell(\ell+1)C_\ell/2\pi$ ($\mu K^2$)</th>
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<th>$\Delta C_\ell$ Model</th>
<th>$\Delta C_\ell$ Total</th>
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Notes. Temperatures are in CMB units.

a There is negligible covariance between bins. The maximum likelihood fit agrees best with $\Lambda$CDM if the $C_\ell$ data given here are multiplied by 0.91.

b For comparison with the SPT results of Lueker et al. (2010), we also compute the spectrum with wider bins. For five bins of width $\Delta \ell = 400$ from 3000 $< \ell < 5000$, we find $39 \pm 6, 41 \pm 7, 49 \pm 9, 36 \pm 12,$ and $35 \pm 13$. For four bins of width $\Delta \ell = 900$ from 5000 $< \ell < 8600$, we find $56 \pm 19, 45 \pm 25, 140 \pm 70,$ and $80 \pm 250 \mu K^2$. The ACT data are also shown in Figure 1 with bins of width $\Delta \ell = 600$ starting at $\ell = 4200$ to 4800; the bandpowers are given by $\Delta \ell = 31 \pm 11, 39 \pm 18, 36 \pm 19, 61 \pm 26, 29 \pm 58,$ and $28 \pm 116 \mu K^2$.

This and all “model” columns are in the same units as the data: thermodynamic $\mu K^2$.

Assumes the best-fit value of $A_{SZ} = 0.63$.

Assumes the best-fit value of $A_p = 11.9$ and uncorrelated Poisson sources.

7. CONSTRAINTS ON SZ AND IR EMISSION

We perform a simple analysis of the power spectrum to quantify the combined contribution from dusty galaxies and radio sources, and the level of SZ emission. We assume an $\Lambda$CDM cosmology with lensing of the CMB and parameters from the five-year WMAP analysis combined with baryon acoustic oscillations and supernovae measurements (Komatsu et al. 2009). We defer a full investigation of cosmological parameter constraints until we improve the absolute calibration and better account for astrophysical foregrounds.

Our model for the power at 148 GHz is

$$B_\ell^{\text{SZ}} = B_\ell^{\text{SZ}} + A_{\text{SZ}} B_\ell^{\text{SZ}} + A_p \left( \frac{\ell}{3000} \right)^2 + B_\ell^{\text{corr}},$$

(13)

where $B_\ell^{\text{SZ}}$ is the lensed primary CMB power spectrum; $B_\ell^{\text{SZ}}$ is a template spectrum corresponding to a prediction for the SZ emission in a model with $\sigma_8 = 0.8$ at 148 GHz; $A_p$ quantifies the Poisson point source power, required to be positive; and $B_\ell^{\text{corr}}$ corresponds to correlated point source power from clustered galaxies. The SZ template we use includes the correlated thermal and kinetic SZ (kSZ) effect derived from numerical simulations and is described in detail in Sehgal et al. (2010). Its amplitude is assumed to scale with $\sigma_8$ as the seventh power, such that $A_{\text{SZ}} = (\sigma_8/0.8)^7$ for fixed baryon density. This accounts approximately for the frequency-dependent combination of the thermal SZ component scaling as the 7.5 to 8th power of $\sigma_8$ and the sub-dominant kSZ scaling as the fifth power. The expected point source power $C_p$ is given by

$$C_p = g(\nu) \int_0^{S_{\text{lim}}} S^2 \frac{dN}{dS} dS (\mu K^2 \text{sr}),$$

(14)

an integral over all sources with flux $S$ up to some maximum flux $S_{\text{lim}}$, where $g(\nu) \equiv (c^2/2k\nu^2) \times ([e^\nu - 1]/(x^2e^x))$ converts flux density to thermodynamic temperature, and $x \equiv h\nu/kT_{\text{CMB}}$. Then $A_p$ is the binned $\ell(\ell+1)C_\ell/2\pi$ Poisson power at pivot $\ell_0 = 3000$. In thermodynamic $\mu K^2$, $C_p = 0.698 \times 10^{-6}A_p$. The conversion to $J_{\mu K}^2 \text{sr}^{-1}$ at 148 GHz is $C_p[J_{\mu K}^2 \text{sr}^{-1}] = 1.55C_p[10^{-5} \mu K^2 \text{sr}]$.

The infrared sources are expected to be clustered. At small angles galaxies cluster with typical correlation function $C(\theta) \propto \theta^{-0.8}$ (e.g., Peebles 1980), which would give $C_\ell \propto \ell^{-1.2}$ on nonlinear scales. Motivated by this, we first adopt a simple template for the correlated power,

$$B_\ell^{\text{corr}} = A_c \left( \frac{\ell}{3000} \right)^2,$$

(15)

and fit for the amplitude $A_c$. Note that on larger scales, $\ell < 300$, $B_\ell^{\text{corr}}$ is expected to flatten and gradually turn over (e.g., Scott & White 1999), but at these scales at 148 GHz the CMB dominates. Models suggest that the power from source clustering is less than the Poisson component at scales smaller than $\ell = 2000$ (Scott & White 1999; Negrello et al. 2007; Righi et al. 2008). This is consistent with observations at 600 GHz by BLAST (Devlin et al. 2009; Viero et al. 2009). We therefore impose a prior that the correlated power be less than the Poisson power at $\ell = 3000$, i.e., $0 < A_c < A_p$. This model is likely too simplistic, and the correlated power may have an alternative shape (e.g., Sehgal et al. 2010 or the halo model considered in Viero et al. 2009). If the power from dusty galaxies is instead better described by
Figure 5. Convergence of the maps as a function of iteration. The map-making algorithm converges well by iteration 500. For signal-only simulations (top), the amplitude of fluctuations in the difference between the processed output map at iteration $i$ and the input map (denoted $(C_i)^{1/2}$) is less than 1% of the amplitude of fluctuations in the input map $(C_b)^{1/2}$ at all scales by $i = 500$. Iterations $i = 5, 20, 100$, and 500 are shown. For simulations with noise and for the data (middle and bottom, respectively), convergence is tested by estimating the maximum change in power between the processed map at iteration $i$ and iteration 1000, as a fraction of the uncertainty in the power in the final map. For iteration 500 this fraction $r_c$ (described in Section 6) is sufficiently small, less than 0.5 at all scales.

(A color version of this figure is available in the online journal.)

Given this model, the likelihood of the data is given by

$$-2 \ln \mathcal{L} = (B_{\text{th}}^b - B_b)^T \Sigma^{-1} (B_{\text{th}}^b - B_b) + \ln \det \Sigma, \quad (16)$$

with covariance matrix $\Sigma$ defined in Equation (8). The theoretical spectrum $B_{\text{th}}^b$ is computed from the model using $B_{\text{th}}^b = w_{b\ell} B_{\ell}^b$, where $w_{b\ell}$ is an approximate form of the bandpower window function in band $b$. We marginalize over the calibration uncertainty analytically (Ganga et al. 1996; Bridle et al. 2002).

The uncertainty on the shape of the window function is small, of order 1.5% for the 148 GHz band, compared to 12% overall calibration uncertainty in power. The window uncertainty is therefore neglected throughout this analysis. We verify that this approximation is valid by including the window function uncertainty in the likelihood calculation. The beam Legendre transform is first expanded in orthogonal basis functions, and the uncertainties on the basis function coefficients are used to derive the window function covariance matrix. The covariance is dominated by a small number of modes, so a singular value decomposition is taken. The 10 largest modes are included in the likelihood, following the method described in Appendix A of Hinshaw et al. (2007). We find that including the window function uncertainty has only a negligible effect on parameter estimates ($<0.05\sigma$). The ACT/MBAC beam measurements and the orthogonal function expansion are described in detail in Hincks et al. (2009).

Figure 6. Estimated two-dimensional power spectrum $C_\ell$ multiplied by a factor of $\ell$ to emphasize the angular, rather than the radial, variation. The power is consistent with being isotropic, when divided into wedges of $\Delta \theta = 20^\circ$. The vertical lines indicate the narrow region $|\ell_x| < 270$, where excess power from scan-synchronous signals contaminates the power spectrum. This region is not used for the power spectrum analysis. The regions near $(|\ell_x|, |\ell_y|) \approx (4000, 4000)$ are more noisy but not biased. A power spectrum computed without these regions is consistent with the one we present.

(A color version of this figure is available in the online journal.)

We confirm that the expected values of parameters $A_{\text{SZ}}$, $A_p$, and $A_c$ are recovered from maps of the ACT simulated sky. We tested simulations with and without noise in the map, and simulations with realistic timestream noise run through the mapper.
In the range $600 < \ell < 1800$, where foreground emission and secondary effects are sub-dominant, the data are consistent with the lensed $\Lambda$CDM model alone, with $\chi^2 = 7.1$ for 4 degrees of freedom (dof). If we rescale the maps to check consistency with the model (multiplying temperatures by $0.96$, a 0.7 change in the calibration), then the $\chi^2$ value becomes 3.0 with 3 dof. We do not rescale the maps in the analysis that follows though the scale factor is allowed to vary when we marginalized over uncertainties.

Using the full range $600 < \ell < 8100$, we find marginalized constraints on $A_p = 11.2 \pm 3.3 \mu K^2$ (thus, $C_p = (0.78 \pm 0.23) \times 10^{-5} \mu K^2$) and $A_{SZ} < 1.63$ (95% CL). The minimum $\chi^2 = 27.0$ for 23 dof. Assuming the scaling of $A_{SZ}$ as $(\sigma_8^{SZ})^3$, this implies an upper limit of $\sigma_8^{SZ} < 0.86$ (95% CL). The one- and two-dimensional distributions are shown in Figure 8, with limits given in Table 2. Marginalizing over the possible SZ power, ACT detects a residual point source component at $3\sigma$ ($\delta \chi^2 = 10$). Marginalizing over a residual term with $A_p < A_p$ gives $A_p = 9.7 \pm 2.8 \mu K^2$, and $\sigma_8^{SZ} < 0.84$ with $\chi^2 = 26.7$ for 22 dof. Because $A_p$ is forced to be positive, its inclusion has the effect of lowering the limits on $A_p$ and $A_{SZ}$. Nevertheless, we do not find evidence for a correlated component with current sensitivity levels. The estimated parameters vary by less than $0.6\sigma$ when the minimum angular scale is varied in the range $5000 < \ell_{\text{max}} < 8000$, or the SZ template is replaced by the spectrum of Komatsu & Seljak (2002), which is approximately 15% lower than our template in the relevant range of $\ell$ from 1000 to 5000. We note that there are multiple models for predicting $A_{SZ}$ (e.g., Bond et al. 2005; Kravtsov et al. 2005) and that the relation between $A_{SZ}$ and $\sigma_8$ is an active area of research.

We also combine the ACT spectrum with the WMAP five-year data (Dunkley et al. 2009) to constrain the six-parameter $\Lambda$CDM
The ACT constraints on the SZ power indicate an amplitude of fluctuations \( \sigma^S_{8} < 0.86 \) (95% CL). This result is consistent with estimates that combine the primordial CMB anisotropy with distance measures, \( \sigma_8 = 0.81 \pm 0.03 \) (Komatsu et al. 2009), and improves on SZ-inferred limits at 150 GHz from Bolocam (\( \sigma^S_{8} < 1.57 \); Sayers et al. 2009), Boomerang (\( \sigma^S_{8} < 1.14 \) at 95% CL; Veneziani et al. 2009), and APEX (\( \sigma^S_{8} < 1.18 \); Reichardt et al. 2009a). We do not see evidence for an excess of SZ power, in contrast to lower frequency observations by CBI at 30 GHz which prefer a value 2.5\( \sigma \) higher than the concordance value (\( \sigma^S_{8} = 0.922 \pm 0.047 \); Sievers et al. 2009).

The ACT result is also consistent with the recently reported \( \sigma^S_{8} = 0.773 \pm 0.025 \) (\( \sigma_{SZ} = 0.42 \pm 0.21 \)) from the SPT (Lueker et al. 2010). The SPT team prefers a form for the correlated point sources that is covariant with the SZ template in the \( \ell = 3000 \) range (Hall et al. 2010). If such a form is correct, then the \( \sigma_{SZ} \) we report should be interpreted as an upper limit on correlated point sources plus the SZ effect. Analysis of the ACT’s 218 and 277 GHz data will shed light on possible forms of the correlated component.

The ACT results on \( \sigma_8 \) are also consistent with several recent studies based on the analysis of ROSAT X-ray flux-selected clusters. We note that these X-ray cluster studies themselves are consistent with measures of \( \sigma_8 \) from richness or weak-lensing-selected cluster samples (see references in the three articles cited here). Henry et al. (2009) find \( \sigma_8(\Omega_m/0.3)^{0.30} = 0.86 \pm 0.04 \) (for \( \Omega_m < 0.32 \) using cluster gas temperatures measured with the ASCA satellite. Vikhlinin et al. (2009) obtain \( \sigma_8(\Omega_m/0.25)^{0.47} = 0.813 \pm 0.013 \) (stat) \pm 0.024 (sys) using temperatures derived from Chandra observations. Similarly, Mantz et al. (2010) find that in spatially flat models with a constant dark energy equation of state, ROSAT X-ray flux-selected clusters yield \( \Omega_m = 0.23 \pm 0.04 \), \( \sigma_8 = 0.82 \pm 0.05 \).

After this article was completed, WMAP seven-year measurements of the brightness of Mars and Uranus were released (Weiland et al. 2010). They suggest that \( T_U \approx 107 \) K, which is
\(\sim 4\%\) dimmer than the value used here. If the new Uranus temperature is adopted, then all absolute temperatures and source brightnesses in this work would be reduced by 4\% and all \(C_\ell\) values by 8\%.

8. CONCLUSIONS

We have presented the first map and the power spectrum of the CMB sky made using data from the ACT at 148 GHz. With this map we can compare to the CMB sky made using data from the ACT at 148 GHz. With an unbiased estimator, we extract the power spectrum, \(C_\ell\), over a range of power exceeding \(10^4\).

We have interpreted the spectrum with a simple model composed of the primary CMB, a possible SZ contribution, and uncorrelated point sources. This analysis provides a new upper bound on the SZ signal from clusters (\(\sigma_{SZ}^2 < 0.86\) at 95\% CL, though this is subject to uncertainty in the SZ models). A coordinated program of X-ray, optical, infrared, and millimeter-wavelength observations of the largest SZ clusters is underway.

These high angular resolution measurements probe the microwave power spectrum out to arcminute scales. Above \(\ell \sim 2500\), the spectrum is sensitive to nonlinear processes such as the formation of galaxy clusters and dusty galaxies. On the low-\(\ell\) end, the spectrum measures the Silk damping tail of the CMB which can be computed using linear perturbation theory as applied to the primordial plasma. It is clear that to understand the \(\ell > 1000\) end of the primary CMB, and thus to improve significantly on measurements of the scalar spectral index and its running, source modeling will be required. Future analyses of the ACT data will include the two higher-frequency channels and additional sky coverage. In another approach, one can measure the high-\(\ell\) \(E\)-modes, because the polarized CMB to foreground ratio is expected to be higher than that for the temperature. We are pursuing both programs, as a polarization-sensitive camera is currently under development.

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