

Planck 2013 results. XXVI. Background geometry and topology of the Universe

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ABSTRACT

The new cosmic microwave background (CMB) temperature maps from *Planck* provide the highest-quality full-sky view of the surface of last scattering available to date. This allows us to detect possible departures from the standard model of a globally homogeneous and isotropic cosmology on the largest scales. We search for correlations induced by a possible non-trivial topology with a fundamental domain intersecting, or nearly intersecting, the last scattering surface (at comoving distance χ_{rec}), both via a direct search for matched circular patterns at the intersections and by an optimal likelihood search for specific topologies. For the latter we consider flat spaces with cubic toroidal (T3), equal-sided chimney (T2) and slab (T1) topologies, three multi-connected spaces of constant positive curvature (dodecahedral, truncated cube and octahedral) and two compact negative-curvature spaces. These searches yield no detection of the compact topology with the scale below the diameter of the last scattering surface. For most compact topologies studied the likelihood maximized over the orientation of the space relative to the observed map shows some preference for multi-connected models just larger than the diameter of the last scattering surface. Since this effect is also present in simulated realizations of isotropic maps, we interpret it as the inevitable alignment of mild anisotropic correlations with chance features in a single sky realization; such a feature can also be present, in milder form, when the likelihood is marginalized over orientations. Thus marginalized, the limits on the radius \mathcal{R}_i of the largest sphere inscribed in topological domain (at log-likelihood-ratio $\Delta \ln \mathcal{L} > -5$ relative to a simply-connected flat *Planck* best-fit model) are: in a flat Universe, $\mathcal{R}_i > 0.92\chi_{\text{rec}}$ for the T3 cubic torus; $\mathcal{R}_i > 0.71\chi_{\text{rec}}$ for the T2 chimney; $\mathcal{R}_i > 0.50\chi_{\text{rec}}$ for the T1 slab; and in a positively curved Universe, $\mathcal{R}_i > 1.03\chi_{\text{rec}}$ for the dodecahedral space; $\mathcal{R}_i > 1.0\chi_{\text{rec}}$ for the truncated cube; and $\mathcal{R}_i > 0.89\chi_{\text{rec}}$ for the octahedral space. The limit for a wider class of topologies, i.e., those predicting matching pairs of back-to-back circles, among them tori and the three spherical cases listed above, coming from the matched-circles search, is $\mathcal{R}_i > 0.94\chi_{\text{rec}}$ at 99% confidence level. Similar limits apply to a wide, although not exhaustive, range of topologies.

We also perform a Bayesian search for an anisotropic global Bianchi VII_h geometry. In the non-physical setting where the Bianchi cosmology is decoupled from the standard cosmology, *Planck* data favour the inclusion of a Bianchi component with a Bayes factor of at least 1.5 units of log-evidence. Indeed, the Bianchi pattern is quite efficient at accounting for some of the large-scale anomalies found in *Planck* data. However, the cosmological parameters that generate this pattern are in strong disagreement with those found from CMB anisotropy data alone. In the physically motivated setting where the Bianchi parameters are coupled and fitted simultaneously with the standard cosmological parameters, we find no evidence for a Bianchi VII_h cosmology and constrain the vorticity of such models to $(\omega/H)_0 < 8.1 \times 10^{-10}$ (95% confidence level).

Key words. cosmology: observations – cosmic background radiation – cosmological parameters – gravitation – methods: data analysis – methods: statistical

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1. Introduction

This paper, one of a set of papers associated with the 2013 release of data from the *Planck*¹ mission (Planck Collaboration I 2014), describes the use of *Planck* data to limit departures from the global isotropy and homogeneity of spacetime. We will use *Planck*'s measurements of the cosmic microwave background (CMB) to assess the properties of anisotropic geometries (i.e., Bianchi models) and non-trivial topologies (e.g., the torus). The simplest models of spacetime are globally isotropic and simply connected. Although both are supported by both local observations and previous CMB observations, without a fundamental theory of the birth of the Universe, observational constraints on departures from global isotropy are necessary. General Relativity itself places no restrictions upon the topology of the Universe, as was recognised very early on (e.g., De Sitter 1917); most proposed theories of quantum gravity predict topology-change in the early Universe which could be visible at large scales today.

The Einstein field equations relate local properties of the curvature to the matter content in spacetime. By themselves they do not restrict the global properties of the space, allowing a universe with a given local geometry to have various global topologies. Friedmann-Robertson-Walker (FRW) models of the universe observed to have the same average local properties everywhere still have freedom to describe quite different spaces at large scales. Perhaps the most remarkable possibility is that a vanishing or negative local curvature ($\Omega_K \equiv 1 - \Omega_{\text{tot}} \geq 0$) does not necessarily mean that our Universe is infinite. Indeed we can still be living in a universe of finite volume due to the global topological multi-connectivity of space, even if described by the flat or hyperbolic FRW solutions. In particular, quantum fluctuations can produce compact spaces of constant curvature, both flat (e.g., Zeldovich & Starobinski 1984) and curved (e.g., Coule & Martin 2000; Linde 2004), within the inflationary scenario.

The primary CMB anisotropy alone is incapable of constraining curvature due to the well-known geometrical degeneracy which produces identical small-scale fluctuations when the recombination sound speed, initial fluctuations, and comoving distance to the last scattering surface are kept constant (e.g., Bond et al. 1997; Zaldarriaga & Seljak 1997; Stompor & Efstathiou 1999). The present results from *Planck* (Planck Collaboration XVI 2014) can therefore place restrictive constraints on the curvature of the Universe only when considering secondary anisotropies or non-CMB data: $\Omega_K = -K(R_0H_0)^{-2} = -0.0010^{+0.0018}_{-0.0019}$ at 95%, considering CMB primary anisotropy and lensing from *Planck* (in the natural units with $c = 1$ we use throughout). This is equivalent to constraints on the radius of curvature $R_0H_0 > 19$ for positive curvature ($K = +1$) and $R_0H_0 > 33$ for negative curvature ($K = -1$). CMB primary anisotropy alone gives limits on R_0H_0 roughly a factor of two less restrictive (and strongly dependent on priors).

Thus, the global nature of the Universe we live in is still an open question and studying the observational effects of a possible finite universe is one way to address it. With topology not affecting local mean properties that are found to be well described by FRW parameters, its main observational effect is in setting boundary conditions on perturbation modes that can be excited

and developed into the structure that we observe. Studying structure on the last scattering surface is the best-known way to probe the global organisation of our Universe and the CMB provides the most detailed and best understood dataset for this purpose.

We can also relax assumptions about the global structure of spacetime by allowing anisotropy about each point in the Universe. This yields more general solutions to Einstein's field equations, leading to the so-called Bianchi cosmologies. For small anisotropy, as demanded by current observations, linear perturbation about the standard FRW model may be applied. A universal shear and rotation induce a characteristic subdominant, deterministic signature in the CMB, which is embedded in the usual stochastic anisotropies. The deterministic CMB temperature fluctuations that result in the homogenous Bianchi models were first examined by Collins & Hawking (1973) and Barrow et al. (1985) (and subsequently Barrow 1986), however no dark energy component was included as it was not considered plausible at the time. More recently, Jaffe et al. (2006c), and independently Bridges et al. (2007), extended these solutions for the open and flat Bianchi VII_h models to include cosmologies with dark energy. It is these solutions to Bianchi VII_h models that we study in the current article. More accurate solutions were since derived by Pontzen & Challinor (2007), Pontzen (2009) and Pontzen & Challinor (2011), where recombination is treated in a more sophisticated manner and reionisation is supported. Furthermore, we note that in these works (Pontzen & Challinor 2007, 2011; Pontzen 2009) the induced CMB polarisation contributions that arise in Bianchi models have also been derived, although here focus is given to temperature contributions.

In this paper, we will explicitly consider models of global topology and anisotropy. In a chaotic inflation scenario, however, our post-inflationary patch might exhibit large-scale local topological features ("handles" and "holes") that can mimic a global multiply-connected topology in our observable volume. Similarly, it might also have residual shear or rotation which could mimic the properties of a global Bianchi spacetime.

Planck's ability to discriminate and remove large-scale astrophysical foregrounds (Planck Collaboration XII 2014) reduces the systematic error budget associated with measurements of the CMB sky significantly. *Planck* data therefore allow refined limits on the scale of the topology and the presence of anisotropy. Moreover, previous work in this field has been done by a wide variety of authors using a wide variety of data (e.g., COBE, WMAP 1-year, 3-year, 5-year, etc.) and in this work we perform a coherent analysis.

In Sect. 2, we discuss previous attempts to limit the topology and global isotropy of the Universe. In Sect. 3 we discuss the signals induced in topologically non-trivial and Bianchi universes. In Sect. 4 the *Planck* data we use in the analysis are presented, and in Sect. 5 the methods we have developed to detect those signals are discussed. We apply those methods in Sect. 6 and discuss the results in Sect. 7.

2. Previous results

The first searches for non-trivial topology on cosmic scales looked for repeated patterns or individual objects in the distribution of galaxies (Sokolov & Shvartsman 1974; Fang & Sato 1983; Fagundes & Wichoski 1987; Lehoucq et al. 1996; Roukema 1996; Weatherley et al. 2003; Fujii & Yoshii 2011). The last scattering surface from which the CMB is released represents the most distant source of photons in the Universe, and hence the largest scales with which we could probe the topology of the Universe. This first became possible with the

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DMR instrument on the COBE satellite (Bennett et al. 1996): various searches found no evidence for non-trivial topologies (e.g., Starobinskij 1993; Sokolov 1993; Stevens et al. 1993; De Oliveira-Costa & Smoot 1995; Levin et al. 1998; Bond et al. 1998, 2000b; Rocha et al. 2004; but see also Roukema 2000b,a), but sparked the creation of robust statistical tools, along with greater care in the enumeration of the possible topologies for a given geometry (see, for example, Lachieze-Rey & Luminet 1995 and Levin 2002 for reviews). With data from the WMAP satellite (Jarosik et al. 2011), these theoretical and observational tools were applied to a high-quality dataset for the first time. Luminet et al. (2003) and Caillierie et al. (2007) claimed the low value of the low multipoles (compared to standard Λ CDM cosmology) as evidence for missing large-scale power as predicted in a closed universe with a small fundamental domain (see also Aurich 1999; Aurich et al. 2004, 2005, 2006, 2008; Aurich & Lustig 2013; Lew & Roukema 2008; Roukema et al. 2008). However, searches in pixel space (Cornish et al. 2004; Key et al. 2007; Niarchou et al. 2004; Bielewicz & Riazuelo 2009; Dineen et al. 2005) and in harmonic space (Kunz et al. 2006) determined that this was an unlikely explanation for the low power. Bond et al. (1998, 2000a) and Riazuelo et al. (2004a,b) presented some of the mathematical formalism for the computation of the correlations induced by topology in a form suitable for use in cosmological calculations. Phillips & Kogut (2006) presented efficient algorithms for the computation of the correlation structure of the flat torus and applied it via a Bayesian formalism to the WMAP data; similar computations for a wider range of geometries were performed by Niarchou & Jaffe (2007).

These calculations used a variety of different vintages of the COBE and WMAP data, as well as a variety of different sky cuts (including the unmasked internal linear combination (ILC) map, not originally intended for cosmological studies). Nonetheless, none of the pixel-space calculations which took advantage of the full correlation structure induced by the topology found evidence for a multiply-connected topology with a fundamental domain within or intersecting the last scattering surface. Hence in this paper we will attempt to corroborate this earlier work and put the calculations on a consistent footing.

The open and flat Bianchi type VII_h models have been compared previously to both the COBE (Bunn et al. 1996; Kogut et al. 1997) and WMAP (Jaffe et al. 2005, 2006b) data, albeit ignoring dark energy, in order to place limits on the global rotation and shear of the Universe. A statistically significant correlation between one of the Bianchi VII_h models and the WMAP ILC map (Bennett et al. 2003) was first detected by Jaffe et al. (2005). However, it was noted that the parameters of this model are inconsistent with standard constraints. Nevertheless, when the WMAP ILC map was “corrected” for the best-fit Bianchi template, some of the so-called “anomalies” reported in WMAP data disappear (Jaffe et al. 2005, 2006b; Cayón et al. 2006; McEwen et al. 2006). A modified template fitting technique was performed by Land & Magueijo (2006) and, although a statistically significant template fit was not reported, the corresponding “corrected” WMAP data were again free of many large scale “anomalies”. Subsequently, Ghosh et al. (2007) used the bipolar power spectrum of WMAP data to constrain the amplitude of any Bianchi component in the CMB. Due to the renewed interest in Bianchi models, solutions to the CMB temperature fluctuations induced in Bianchi VII_h models when incorporating dark energy were since derived by Jaffe et al. (2006c) and Bridges et al. (2007). Nevertheless, the cosmological parameters of the Bianchi template embedded in WMAP data in this setting remain inconsistent with constraints from the CMB alone

(Jaffe et al. 2006a,c). Furthermore, Pontzen & Challinor (2007) compared the polarisation power spectra of the best-fit Bianchi VII_h model found by Jaffe et al. (2006a) with the WMAP 3-year data (Page et al. 2007) and also concluded that the model could be ruled out since it produced greater polarization than observed in the WMAP data. A Bayesian analysis of Bianchi VII_h models was performed by Bridges et al. (2007) using WMAP ILC data to explore the joint cosmological and Bianchi parameter space via Markov chain Monte Carlo sampling, where it was again determined that the parameters of the resulting Bianchi cosmology were inconsistent with standard constraints. In a following study by Bridges et al. (2008) it was suggested that the CMB “cold spot” (Vielva et al. 2004; Cruz et al. 2006; Vielva 2010) could be driving evidence for a Bianchi component. Recently, this Bayesian analysis has been revisited by McEwen et al. (2013) to handle partial-sky observations and to use nested sampling methods (Skilling 2004; Feroz & Hobson 2008; Feroz et al. 2009). McEwen et al. (2013) conclude that WMAP 9-year temperature data do not favour Bianchi VII_h cosmologies over Λ CDM.

3. CMB correlations in anisotropic and multiply-connected universes

3.1. Topology

All FRW models can describe multi-connected universes. In the case of flat space, there are a finite number of compactifications, the simplest of which are those of the torus. All of them have continuous parameters that describe the length of periodicity in some or all directions (e.g., Riazuelo et al. 2004b). In a space of constant non-zero curvature the situation is notably different – the presence of a length scale (the curvature radius R_0) precludes topological compactification at an arbitrary scale. The size of the space must now reflect its curvature, linking topological properties to $\Omega_{\text{tot}} = 1 - \Omega_K$. In the case of hyperbolic spacetimes, the list of possible compact spaces of constant negative curvature is still infinite, but discrete (Thurston 1982), while in the positive curvature spherical space there is only a finite set of well-proportioned possibilities (i.e., those with roughly comparable sizes in all directions; there are also the countably infinite lens and prism topologies) for a multi-connected space (e.g., Gausmann et al. 2001; Riazuelo et al. 2004a).

The effect of topology is equivalent to considering the full simply-connected three-dimensional spatial slice of the space-time (known as the *covering space*) as being filled with repetitions of a shape which is finite in some or all directions (the *fundamental domain*) – by analogy with the two-dimensional case, we say that the fundamental domain *tiles* the covering space. For the flat and hyperbolic geometries, there are infinite copies of the fundamental domain; for the spherical geometry, with a finite volume, there is a finite number of tiles. Physical fields repeat their configuration in every tile, and thus can be viewed as defined on the covering space but subject to periodic boundary conditions. Topological compactification always break isotropy, and for some topologies also the global homogeneity of physical fields. Positively curved and flat spaces studied in this paper are homogeneous, however hyperbolic multi-connected spaces are never homogeneous.

The primary observable effect of a multi-connected universe is the existence of directions in which light could circumnavigate the space in cosmological time more than once, i.e., the radial distance χ_{rec} to the surface of last scattering exceeds the size of the universe. In these cases, the surface of last scattering can

Table 1. Parameters of analysed curved spaces.

Size	Spherical			Hyperbolic	
	Dodecahedral	Truncated cube	Octahedral	m004(-5,1)	v3543(2,3)
\mathcal{V}/R_0^3	0.16	0.41	0.82	0.98	6.45
\mathcal{R}_i/R_0	0.31 ($\pi/10$)	0.39 ($\pi/8$)	0.45	0.54	0.89
\mathcal{R}_m/R_0	0.37	0.56	0.56	0.64	1.22
\mathcal{R}_u/R_0	0.40	0.58	0.79 ($\pi/4$)	0.75	1.33

intersect the (notional) edge of a fundamental domain. At this intersection, we can view the same spacetime event from multiple directions – conversely, it appears in different directions when observed from a single point.

Thus, temperature perturbations in one direction, $T(\hat{n})$, become correlated with those in another direction $T(\hat{m})$ by an amount that differs from the usual isotropic correlation function $C(\theta)$, where θ denotes the angle between \hat{n} and \hat{m} . Considering a pixelized map, this induces a correlation matrix $C_{pp'}$ which depends on quantities other than the angular distance between pixels p and p' . This break from statistical isotropy can therefore be used to constrain topological models. Hence, we need to calculate the pixel-space correlation matrix or its equivalent in harmonic space.

In this paper we consider the following topologies using the likelihood method: a) toroidal flat models with equal-length compactification size L in three directions, denoted $T[L, L, L]^2$; b) toroidal flat models with different compactification lengths, parametrized by L_x, L_y, L_z , denoted $T[L_x, L_y, L_z]$; c) three major types of single-action positively curved spherical manifolds with dodecahedral, truncated cubical and octahedral fundamental domains (I^* , O^* , T^* compactification groups correspondingly, see Gausmann et al. 2001); and d) two sample negative curvature hyperbolic spaces, m004(-5,1) being one of the smallest known compact hyperbolic spaces as well as the relatively large v3543(2,3)³. Scales of fundamental domains of compactified curved spaces are fixed in the units of curvature and are summarised in Table 1, where we quote the volume \mathcal{V} , radius of the largest sphere that can be inscribed in the domain \mathcal{R}_i (equal to the distance to the nearest face from the origin of the domain), the smallest sphere in which the domain can be inscribed \mathcal{R}_u (equal to the distance to the farthest vertex), and the intermediate scale \mathcal{R}_m that is taken to be the distance to the edges for spherical spaces and the “spine” distance for hyperbolic topologies. For the cubic torus with edge length L , these lengths are $\mathcal{R}_i = L/2$, $\mathcal{R}_m = \sqrt{2}L/2$ and $\mathcal{R}_u = \sqrt{3}L/2$. The ratio $\mathcal{R}_u/\mathcal{R}_i$ is a good indicator of the shape of the fundamental domain. Note that when χ_{rec} is less than \mathcal{R}_i , multiple images on large scales are not present, although the $C_{pp'}$ correlation matrix is still modified versus the singly-connected limit. The effects of topology usually become strong when χ_{rec} exceeds the intermediate \mathcal{R}_m ; conversely, for flat and nearly-flat geometries, there are limits to the allowed topologies (Mota et al. 2011).

A much wider class of topologies is explicitly constrained using the matched circles method. As discussed in Sect. 5.1, because of computational limitations we restrict our analysis to

pairs of circles centered around antipodal points, so called back-to-back circles. Thus, we can constrain all topologies predicting pairs of such circles. The strongest constraints are imposed on topologies predicting back-to-back circles in all directions i.e., all the single action manifolds, among them tori of any shape and the three spherical cases considered explicitly in the likelihood analysis. Weaker constraints are imposed on topologies with all back-to-back circles centred on a great circle of the celestial sphere such as half-turn, quarter-turn, third-turn and sixth-turn spaces, as well as Klein and chimney spaces. The statistic can also constrain the multi-connected spaces predicting one pair of antipodal matching circles such as Klein or chimney spaces with horizontal flip, vertical flip or half-turn and slab space translated without screw motion. Other topologies catalogued in Riazuelo et al. (2004b) are not constrained by this analysis: the Hantzsche-Wendt space; the chimney space with half-turn and flip; the generic slab space; the slab space with flip; spherical manifolds with double and linked action; and all the hyperbolic topologies including those two cases considered using the likelihood method.

3.1.1. Computing correlation matrices

The CMB temperature pixel-pixel correlation matrix is defined as the ensemble-average product of the temperature at two different pixels:

$$C_{pp'} = \langle T_p T_{p'} \rangle. \quad (1)$$

It can be calculated as a double radial integral of the ensemble average of the product of the source functions that describe the transport of photons through the universe from the last scattering surface to the observer:

$$C_{pp'} = \int_0^{\chi_{\text{rec}}} d\chi \int_0^{\chi_{\text{rec}}} d\chi' \langle S(\chi \hat{q}_p) S(\chi' \hat{q}_{p'}) \rangle, \quad (2)$$

where \hat{q}_p and $\hat{q}_{p'}$ are unit vectors that point at pixels p and p' on the sky, and χ and χ' are proper distances along radial rays pointing towards the last scattering surface.

Two techniques have been developed to compute the CMB correlation function for multiply-connected universes. In one approach, one constructs the orthonormal set of basis functions that satisfy the boundary conditions imposed by compactification (eigenfunctions of the Laplacian operator furnish such a basis), and assembles the *spatial* correlation function of the source $\langle S(\chi \hat{q}_p) S(\chi' \hat{q}_{p'}) \rangle$ from such a basis (Cornish & Spergel 1999; Lehoucq et al. 2002). In the other approach, one applies the method of images to create the compactified version of $\langle S(\chi \hat{q}_p) S(\chi' \hat{q}_{p'}) \rangle$ from the one computed on the universal covering space by resumming the latter over the images of the 3D spatial positions $\chi \hat{q}_p$ (Bond et al. 1998, 2000a,b):

$$\langle S(\chi \hat{q}_p) S(\chi' \hat{q}_{p'}) \rangle^c = \sum_{\gamma \in \Gamma} \langle S(\chi \hat{q}_p) \gamma[S(\chi' \hat{q}_{p'})] \rangle^u, \quad (3)$$

² In a slight abuse of notation, the lengths L_i will be given in units of H_0^{-1} in $T[L_1, L_2, L_3]$, but in physical units elsewhere.

³ The nomenclature for hyperbolic spaces follows J. Weeks’ census, as incorporated in the freely available SnapPea software, <http://www.geometrygames.org/SnapPea>; see also Thurston & Levy (1997).

where the superscripts *c* and *u* refer to the quantity in the multiply-connected space and its universal cover, respectively. The tilde refers to the need for sum regularization in the models with an infinite set of images, e.g., hyperbolic and flat toroidal ones. Γ is the discrete subgroup of motions which defines the multiply-connected space and $\gamma[x]$ is the spatial point on the universal cover obtained by the action of the motion $\gamma \in \Gamma$ on the point x . Note that we can consider the location of one of the pixels as fixed and consider the action of γ on the other due to symmetry. This equation defines the action of γ on the source function itself, needed unless all the terms in the source function are scalar quantities (which is the case if one limits consideration to Sachs-Wolfe terms) when the action is trivial.

Both methods are general, but have practical considerations to take into account when one increases the pixel resolution. For computing $C_{pp'}$ up to the resolution corresponding to harmonic mode $\ell \approx 40$ both methods have been tested and were found to work equally well. In this paper we employ both approaches.

The main effect of the compactification is that $C_{pp'}$ is no longer a function of the angular separation between the pixels p and p' only, due to the lack of global isotropy. In harmonic space the two-point correlation function of the CMB is given by

$$C_{\ell\ell'}^{mm'} = \langle a_{\ell m} a_{\ell' m'}^* \rangle \neq C_{\ell} \delta_{\ell\ell'} \delta_{mm'}, \quad (4)$$

where $\delta_{\ell\ell'}$ is the Kronecker delta symbol and $a_{\ell m}$ are the spherical harmonic coefficients of the temperature on the sky when decomposed into the spherical harmonics $Y_{\ell m}(\hat{q})$ by

$$T(\hat{q}) = \sum_{\ell m} a_{\ell m} Y_{\ell m}(\hat{q}). \quad (5)$$

Note that the two-point correlation function $C_{\ell\ell'}^{mm'}$ is no longer diagonal, nor is it m -independent, as in an isotropic universe.

A flat universe provides an example when the eigenfunctions of the Laplacian are readily available in a set of plane waves. The topological compactification in the flat space discretizes the spectrum of the wavevector magnitudes k^2 and selects the subset of allowed directions. For example, for a toroidal universe the length of the fundamental cell needs to be an integer multiple of the wavelength of the modes. We therefore recover a discrete sum over modes $\mathbf{k}_n = (2\pi/L)\mathbf{n}$ for $\mathbf{n} = (n_x, n_y, n_z)$ a triplet of integers, instead of an integral over \mathbf{k} ,

$$C_{\ell\ell'}^{mm'} \propto \int d^3k \Delta_{\ell}(k, \Delta\eta) \Delta_{\ell'}(k, \Delta\eta) P(k) \rightarrow \sum_n \Delta_{\ell}(k_n, \Delta\eta) \Delta_{\ell'}(k_n, \Delta\eta) P(k_n) Y_{\ell m}(\hat{\mathbf{n}}) Y_{\ell' m'}^*(\hat{\mathbf{n}}), \quad (6)$$

where $\Delta_{\ell}(k, \Delta\eta)$ is the radiation transfer function (e.g., Bond & Efstathiou 1987; Seljak & Zaldarriaga 1996). We refer to the cubic torus with three equal sides as the T3 topology; it is also possible for the fundamental domain to be compact in only two spatial dimensions (e.g., the so-called T2 “chimney” space) or one (the T1 “slab”, similar to the “lens” spaces available in manifolds with constant positive curvature) in which case the sum is replaced by an integral in those directions. These models serve as approximations to modifications to the local topology of the global manifold (albeit on cosmological scales): for example, the chimney space can mimic a “handle” connecting different regions of an approximately flat manifold.

In Fig. 1 we show rows of the pixel-space correlation matrix for a number of multiply-connected topologies as maps, showing the magnitude of the correlation within a particular pixel. For the simply-connected case, the map simply shows the same

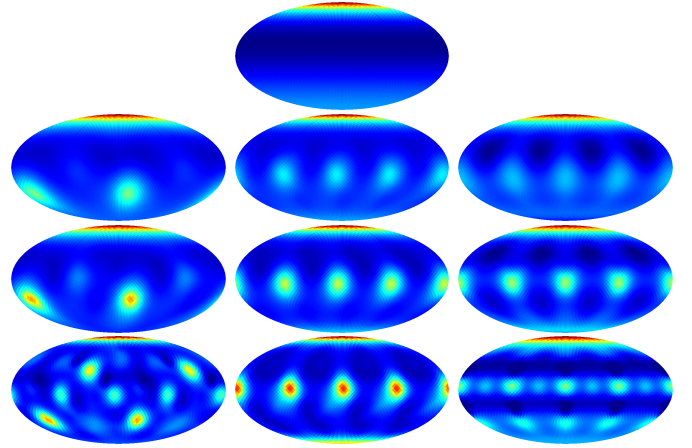


Fig. 1. *Top row:* correlation structure (i.e., a single row of the correlation matrix) of a simply-connected universe with isotropic correlations. For subsequent rows, the left and middle column show positively curved multiply-connected spaces (*left:* dodecahedral, *middle:* octahedral) and the right column shows equal sided tori. *The upper row* of three maps corresponds to the case when the size of the fundamental domain is of the size of the diameter to the last scattering surface and hence the first evidence for large angle excess correlation appears. Subsequent rows correspond to decreasing fundamental domain size with respect to the last scattering diameter, with parameters roughly chosen to maintain the same ratio between the models.

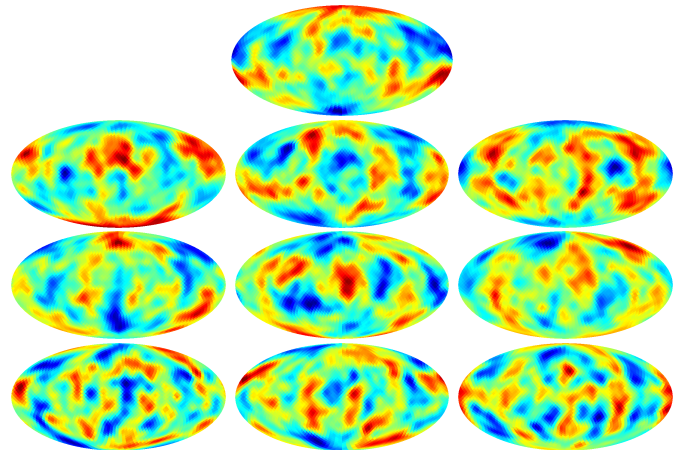


Fig. 2. Random realisations of temperature maps for the models in Fig. 1. The maps are smoothed with a Gaussian filter with full-width-half-maximum $FWHM = 640'$.

information as the correlation function $C(\theta)$; for the topologically non-trivial cases, we see the correlations depend on distance and direction and differ from pixel to pixel (i.e., from row to row of the matrix). In Fig. 2 we show example maps of CMB anisotropies in universes with these topologies, created by direct realisations of Gaussian fields with the correlation matrices of Fig. 1.

3.2. Bianchi

Bianchi cosmologies include the class of homogeneous but anisotropic cosmologies, where the assumption of isotropy about each point in the Universe is relaxed. For small anisotropy, as demanded by current observations, linear perturbation theory about the standard FRW model may be applied, leading to a subdominant, deterministic contribution to the CMB fluctuations. In this setting CMB fluctuations may be viewed as the sum

of a deterministic Bianchi contribution and the usual stochastic contribution that arises in the Λ CDM model. The deterministic CMB temperature fluctuations that result in the Bianchi models were derived by Barrow et al. (1985), although no dark energy component was included. More recently, Jaffe et al. (2006c), and independently Bridges et al. (2007), extended these solutions for the open and flat Bianchi VII_h models to include cosmologies with dark energy. We defer the details of the CMB temperature fluctuations induced in Bianchi models to these works and give only a brief description here.

Bianchi VII_h models describe a universe with overall rotation, parameterized by an angular velocity, ω , and a three-dimensional rate of shear, parameterized by the tensor σ_{ij} ; we take these to be relative to the z axis. The model has a free parameter, first identified by Collins & Hawking (1973), describing the comoving length-scale over which the principal axes of shear and rotation change orientation. The ratio of this length scale to the present Hubble radius is typically denoted x , which defines the h parameter of type VII_h models through (Barrow et al. 1985)

$$x = \sqrt{\frac{h}{1 - \Omega_{\text{tot}}}}, \quad (7)$$

where the total energy density $\Omega_{\text{tot}} = \Omega_{\text{m}} + \Omega_{\Lambda}$. The parameter x acts to change the “tightness” of the spiral-type CMB temperature contributions that arise due to the geodesic focusing of Bianchi VII_h cosmologies. The shear modes σ_{ij} of combinations of orthogonal coordinate axes are also required to describe a Bianchi cosmology. The present dimensionless vorticity $(\omega/H)_0$ may be related to the dimensionless shear modes $(\sigma_{ij}/H)_0$ by (Barrow et al. 1985)

$$\left(\frac{\omega}{H}\right)_0 = \frac{(1+h)^{1/2}(1+9h)^{1/2}}{6x^2\Omega_{\text{tot}}} \sqrt{\left(\frac{\sigma_{12}}{H}\right)_0^2 + \left(\frac{\sigma_{13}}{H}\right)_0^2}, \quad (8)$$

where H is the Hubble parameter. The spherical harmonic coefficients of the Bianchi VII_h induced temperature component are proportional to $[(\sigma_{12} \pm i\sigma_{13})/H]_0$ and are non-zero for azimuthal modes $m = \mp 1$ only (Barrow et al. 1985; McEwen et al. 2006; Pontzen & Challinor 2007). Hence, varying the phase of $\sigma_{12} + i\sigma_{13}$ corresponds to an azimuthal rotation, i.e., a change of coordinates, while the rotationally invariant part depends on $\sigma_{12}^2 + \sigma_{13}^2$, and we are thus free to choose equality of shear modes $\sigma = \sigma_{12} = \sigma_{13}$ (Pontzen & Challinor 2007), which we do for consistency with previous studies (e.g. Jaffe et al. 2005). The amplitude of the deterministic CMB temperature fluctuations induced in Bianchi VII_h cosmologies may be characterised by either $(\sigma/H)_0$ or $(\omega/H)_0$ since these parameters influence the amplitude of the induced temperature contribution only and not its morphology. The handedness of the coordinate system is also free in Bianchi VII_h models, hence both left- and right-handed models arise. Since the Bianchi-induced temperature fluctuations are anisotropic on the sky the orientation of the resulting map may vary also, introducing three additional degrees-of-freedom. The orientation of the map is described by the Euler angles⁴ (α, β, γ) , where for $(\alpha, \beta, \gamma) = (0^\circ, 0^\circ, 0^\circ)$ the swirl pattern typical of Bianchi templates is centred on the South pole.

Examples of simulated Bianchi VII_h CMB temperature maps are illustrated in Fig. 3 for a range of parameters. In the analysis performed herein the BIANCHI2⁵ (McEwen et al. 2013)

⁴ The active zyz Euler convention is adopted, corresponding to the rotation of a physical body in a *fixed* coordinate system about the z , y and z axes by γ , β and α respectively.

⁵ <http://www.jasonmcewen.org/>

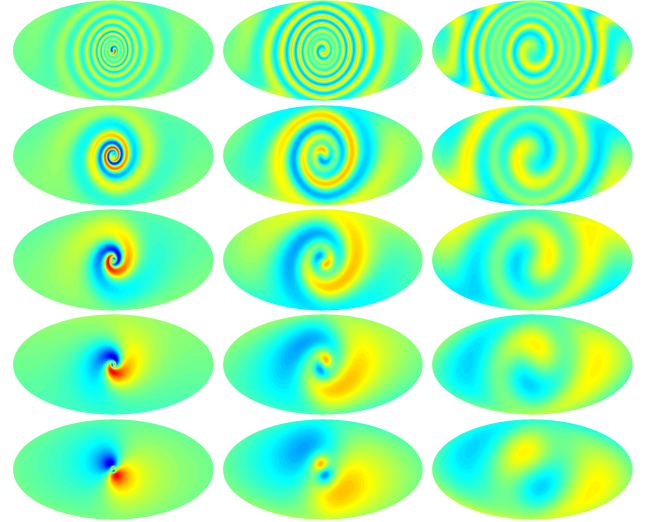


Fig. 3. Simulated deterministic CMB temperature contributions in Bianchi VII_h cosmologies for varying x and Ω_{tot} (left-to-right $\Omega_{\text{tot}} \in \{0.10, 0.50, 0.95\}$; top-to-bottom $x \in \{0.1, 0.3, 0.7, 1.5, 6.0\}$). In these maps the swirl pattern typical of Bianchi-induced temperature fluctuations is rotated from the South pole to the Galactic centre for illustrational purposes.

code is used to simulate the temperature fluctuations induced in Bianchi VII_h models. Bianchi VII_h models induce only large scale temperature fluctuations in the CMB and consequently Bianchi maps have a particularly low band-limit, both globally and azimuthally (i.e., in both ℓ and m in spherical harmonic space; indeed, as mentioned only those harmonic coefficients with $m = \pm 1$ are non-zero).

4. Data description

We use *Planck* maps that have been processed by the various component-separation pipelines described in Planck Collaboration XII (2014). The methods produce largely consistent maps of the sky, with detailed differences in pixel intensity, noise properties, and masks. Here, we consider maps produced by the Commander-Ruler, NILC, SMICA and SEVEM methods. Each provides its own mask and we also consider the conservative common mask.

We note that because our methods rely on rather intensive pixel- or harmonic-space calculations, in particular considering a full set of three-dimensional orientations and, for the likelihood methods, manipulation of an anisotropic correlation matrix, computational efficiency requires the use of data degraded from the native HEALPIX (Górski et al. 2005) $N_{\text{side}} = 2048$ resolution of the *Planck* maps. Because the signatures of either a multiply-connected topology or a Bianchi model are most prominent on large angular scales, this does not result in a significant loss of ability to detect and discriminate amongst the models (see Sect. 5.3). However, it is worth pointing out that the likelihood and matched circles methods are sensitive to different angular scales as applied to *Planck* data here. The likelihood method explicitly retains only low- ℓ (large-scale) information in its correlation matrix, whereas the matched-circles method considers anisotropies at angular scales down to tens of arcminutes (still large in comparison to the native resolution of the *Planck* maps). Of course, the matched-circles method exploits the correlation of the small-scale patterns along matched circles potentially separated by large angles; this effect is not generated by

intrinsically large angular scale anisotropies but by the boundary conditions of the fundamental domain imposed by the multiply connected topology. As described in Sect. 5.1, the matched circles statistic used here damps the anisotropies at the largest angular scales relative to those at smaller scales so sensitivity of the method does not rely on the former.

The topology analyses both rely on degraded maps and masks. The matched-circles method smooths with a $30'$ Gaussian filter and degrades the maps to $N_{\text{side}} = 512$, and uses a mask derived from the SEVEM component separation method. Because the performance of the matched-circles statistic depends on anisotropies on smaller angular scales, it can be significantly degraded by the point source cut. As there are more point sources detected in the *Planck* maps than in the WMAP maps, the problem of point source masking is more severe in the present case. We mask only those point sources from the full-resolution $f_{\text{sky}} = 0.73$ SEVEM mask with amplitude, after smoothing and extrapolation to the 143 or 217 GHz channels, greater than the faintest source originally detected at those frequencies. The mask derived in this way retains $f_{\text{sky}} = 0.76$ of the sky.

The likelihood method smooths the maps and masks with an 11° Gaussian filter and then degrades them to $N_{\text{side}} = 16$ and conservatively masks out any pixel with more than 10% of its original subpixels masked. At full resolution, the common mask retains a fraction $f_{\text{sky}} = 0.73$ of the sky, and $f_{\text{sky}} = 0.78$ when degraded to $N_{\text{side}} = 16$ (the high-resolution point-source masks are largely filled in the degraded masks). The Bianchi analysis is performed in harmonic space, and so does not require explicit degradation in pixel space. Rather, a *noisy mask* is added in pixel space to effectively marginalise the pixel values in the masked region (as described in more detail below; see also McEwen et al. 2013), before the data are transformed at full resolution into harmonic space and considered only up to a specified maximum harmonic ℓ , where correlations due to the mask are taken into account.

Different combinations of these maps and masks are used to discriminate between the topological and anisotropic models described in Sect. 3.

5. Methods

5.1. Topology: circles in the sky

The first set of methods, exemplified by the circles-in-the-sky of Cornish et al. (1998), involves a frequentist analysis using a statistic which is expected to differ between the models examined. For the circles, this uses the fact that the intersection of the topological fundamental domain with the surface of last scattering is a circle, which one potentially views from two different directions in a multiply-connected universe. Of course, the matches are not exact due to noise, foregrounds, the integrated Sachs-Wolfe (ISW) and Doppler effects along the different lines of sight.

By creating a statistic based on the matching of different such circles, we can compare Monte Carlo simulations of both a simply-connected, isotropic null model with specific anisotropic or topological models. We may then calibrate detections and non-detections using Monte Carlo simulations. In principle, these simulations should take into account the complications of noise, foreground contributions, systematics, the ISW and Doppler effects. However, they do not include gravitational lensing of the CMB as the lensing deflection angle is small compared to the minimal angular scale taken into account

in our analysis. Note that the null test is generic (i.e., not tied to a specific topology) but any detection must be calibrated with specific simulations for a chosen topology or anisotropic model. A very similar technique can be used for polarisation by taking into account the fact that the polarisation pattern itself is now not directly repeated, but rather that the underlying quadrupole radiation field around each point on the sky is now seen from different directions (Bielewicz et al. 2012). These methods have been applied successfully to COBE DMR and WMAP data, and have recently been shown to be feasible for application to *Planck* data (Bielewicz et al. 2012).

The idea of using the matched circles to study topology is due to Cornish et al. (1998). In that work, a statistical tool was developed to detect correlated circles in all sky maps of the CMB anisotropy – the circle comparison statistic. In our studies we will use version of this statistic optimised for the small-scale anisotropies as defined by Cornish et al. (2004):

$$S_{i,j}^+(\alpha, \phi_*) = \frac{2 \sum_m |m| \Delta T_{i,m} \Delta T_{j,m}^* e^{-im\phi_*}}{\sum_n |n| (|\Delta T_{i,n}|^2 + |\Delta T_{j,n}|^2)}, \quad (9)$$

where $\Delta T_{i,m}$ and $\Delta T_{j,m}$ denote the Fourier coefficients of the temperature fluctuations around two circles of angular radius α centered at different points on the sky, i and j , respectively, with relative phase ϕ_* . The m th harmonic of the temperature anisotropies around the circle is weighted by the factor $|m|$, taking into account the number of degrees of freedom per mode. Such weighting enhances the contribution of small-scale structure relative to large-scale fluctuations and is especially important since the large-scale fluctuations are dominated by the ISW effect. This can obscure the image of the last scattering surface and reduce the ability to recognise possible matched patterns on it.

The above S^+ statistic corresponds to pair of circles with the points ordered in a clockwise direction (phased). For alternative ordering, when along one of the circles the points are ordered in an anti-clockwise direction (anti-phased), the Fourier coefficients $\Delta T_{i,m}$ are complex conjugated, defining the S^- statistic. This allows the detection of both orientable and non-orientable topologies. For orientable topologies the matched circles have anti-phased correlations while for non-orientable topologies they have a mixture of anti-phased and phased correlations.

The statistic has a range over the interval $[-1, 1]$. Circles that are perfectly matched have $S = 1$, while uncorrelated circles will have a mean value of $S = 0$. Although the statistic can also take negative values for the temperature anisotropy generated by the Doppler term (Bielewicz et al. 2012), anticorrelated circles are not expected for the total temperature anisotropy considered in this work. To find matched circles for each radius α , the maximum value $S_{\text{max}}^\pm(\alpha) = \max_{i,j,\phi_*} S_{i,j}^\pm(\alpha, \phi_*)$ is determined.

Because general searches for matched circles are computationally very intensive, we restrict our analysis to a search for pairs of circles centered around antipodal points, so called back-to-back circles. As described above, the maps were also downgraded to $N_{\text{side}} = 512$, which greatly speeds up the computations required, but with no significant loss of discriminatory power, as seen in Sect. 5.3.1. More details on the numerical implementation of the algorithm can be found in the paper by Bielewicz & Banday (2011).

As mentioned in Sect. 3.1, the constraints we will derive concern topologies that predict matching pairs of back-to-back circles. However, the constraints do not apply to those universes for which the orientation of the matched circles is impossible to detect due to partial masking on the sky. Because of the larger sky fraction removed by the *Planck* common mask than for WMAP

this probability is larger for the analysis of the *Planck* maps. Moreover, the smaller fraction of the sky used in the search of matched circles results in a false detection level larger with our $f_{\text{sky}} = 0.76$ mask than for the $f_{\text{sky}} = 0.78$ 7-year KQ85 WMAP mask. As a result we obtain weaker – but more conservative – constraints on topology than for similar analyses of WMAP data (Bielewicz & Banday 2011).

To draw any conclusions from an analysis based on the statistic $S_{\text{max}}^{\pm}(\alpha)$, it is very important to correctly estimate the threshold for a statistically significant match of circle pairs. We used 300 Monte Carlo simulations of CMB maps, described in detail in Sect. 5.3.1, to establish the threshold such that fewer than 1% of simulations would yield a false event.

5.2. Bayesian analyses

The second set of methods take advantage of the fact that the underlying small-scale physics is unchanged in both anisotropic and topological models compared to the standard cosmology, and thus a Gaussian likelihood function will still describe the statistics of temperature and polarization on the sky, albeit no longer with isotropic correlations. When considering specific topologies, these likelihood methods instead calculate the pixel-pixel correlation matrix. This has been done for various torus topologies (which are a continuous family of possibilities) in the flat Universe as well as for locally hyperbolic and spherical geometries (which have a discrete set of possibilities for a given value of the curvature). More general likelihood-based techniques have been developed for generic mild anisotropies in the initial power spectrum (Hanson & Lewis 2009), which may have extension to other models. For the Bianchi setting, an isotropic zero-mean Gaussian likelihood is recovered by subtracting a deterministic Bianchi component from the data, where the cosmological covariance matrix remains diagonal in harmonic space but masking introduces non-diagonal structure that must be taken into account.

Because these methods use the likelihood function directly, they can take advantage of any detailed noise correlation information that is available, including any correlations induced by the foreground-removal process. We denote the data by the vector \mathbf{d} , which may be in the form of harmonic coefficients d_{lm} or pixel temperatures d_p or, in general, coefficients of the temperature expansion in any set of basis functions. We denote the model under examination by the discrete parameter M , which can take on the appropriate value denoting the usual isotropic case, or the Bianchi case, or one of the possible multiply-connected universes. The continuous parameters of model M are given by the vector Θ , which for this case we can partition into Θ_C for the cosmological parameters shared with the usual isotropic and simply-connected case, and Θ_A which denotes the parameters for the appropriate anisotropic case, be it a topologically non-trivial universe or a Bianchi model. Note that all of the anisotropic cases contain “nuisance parameters” which give the orientation of either the fundamental domain or the Bianchi template which we can marginalize over as appropriate.

Given this notation, the posterior distribution for the parameters of a particular model, M , is given by Bayes’ theorem:

$$P(\Theta|\mathbf{d}, M) = \frac{P(\Theta|M)P(\mathbf{d}|\Theta, M)}{P(\mathbf{d}|M)}. \quad (10)$$

Here, $P(\Theta|M) = P(\Theta_C, \Theta_A|M)$ is the joint prior probability of the standard cosmological parameters Θ_C and those describing the anisotropic universe Θ_A , $P(\mathbf{d}|\Theta, M) \equiv \mathcal{L}$ is the likelihood,

and the normalizing constant $P(\mathbf{d}|M)$ is the Bayesian evidence, which can be used to compare the models to one another.

We will usually take the priors to be simple “non-informative” distributions (e.g., uniform over the sphere for orientations, uniform in length for topology scales, etc.) as appropriate. The form of the likelihood function will depend on the anisotropic model: for multiply-connected models, the topology induces anisotropic correlations, whereas for the Bianchi model, there is a deterministic template, which depends on the Bianchi parameters, in addition to the standard isotropic cosmological perturbations. We will assume that any other non-Gaussian signal (either from noise or cosmology) is negligible (Planck Collaboration XXIII 2014; Planck Collaboration XXIV 2014) and use an appropriate multivariate Gaussian likelihood.

Given the signal and noise correlations, and a possible Bianchi template, the procedure is similar to that used in standard cosmological-parameter estimation, with a few complications. Firstly, the evaluation of the likelihood function is computationally expensive and usually limited to large angular scales. This means that in practice the effect of the topology on the likelihood is usually only calculated on those large scales. Secondly, the orientation of the fundamental domain or Bianchi template requires searching (or marginalizing) over three additional parameters, the Euler angles.

5.2.1. Topology

In topological studies, the parameters of the model consist of Θ_C , the set of cosmological parameters for the fiducial best-fit flat cosmological model, and Θ_T , the topological parameters which include the set of compactification lengths L_x, L_y, L_z for flat toroidal model or the curvature parameter Ω_K for curved spaces, and a choice of compactification T . In our studies we keep Θ_C fixed, and vary Θ_T for a select choice of compactifications listed in Sect. 3.1. These parameters define the predicted two-point signal correlation matrix $C_{pp'}$ for each model, which are precomputed. Additional internal parameters, including the amplitude of the signal A and the angles of orientation of the fundamental domain of the compact space relative to the sky φ (e.g., parameterized by a vector of the three Euler angles), are maximized and/or marginalized over during likelihood evaluation.

The likelihood, i.e., the probability to find a temperature data map \mathbf{d} with associated noise matrix \mathbf{N} given a certain topological model is then given by

$$P(\mathbf{d}|\mathbf{C}[\Theta_C, \Theta_T, T], A, \varphi) \propto \frac{1}{\sqrt{|\mathbf{AC} + \mathbf{N}|}} \times \exp\left\{-\frac{1}{2}\mathbf{d}^*(\mathbf{AC} + \mathbf{N})^{-1}\mathbf{d}\right\}. \quad (11)$$

Working with a cut-sky, it is often easier to start the analysis with data and a correlation matrix given in pixel space. However, especially in the realistic case of negligible noise on large scales, the matrix $\mathbf{C} + \mathbf{N}$ is poorly conditioned in pixel space, and pixel space evaluation of the likelihood is, as a rule, not robust. Indeed, there are typically more pixels than independent modes that carry information about the signal (e.g., even in the standard isotropic case, sub-arcminute pixels would not be useful due to beam-smoothing; with anisotropic correlations and masked regions of the sky, more complicated linear combinations of pixels even on large scales may have very little signal content). Therefore in general we expand the temperature map d_p , the theoretical correlation matrix $C_{pp'}$, and the noise covariance matrix $N_{pp'}$ in a discrete set of mode functions $\psi_n(p)$,

orthonormal over the pixelized sphere, possibly with weights $w(p)$, $\sum_p w(p)\psi_n(p)\psi_{n'}^*(p) = \delta_{nn'}$, obtaining the coefficients of expansion

$$\begin{aligned} d_n &= \sum_p d_p \psi_n^*(p) w(p); \\ C_{nn'} &= \sum_p \sum_{p'} C_{pp'} \psi_n(p) \psi_{n'}^*(p') w(p) w(p'); \\ N_{nn'} &= \sum_p \sum_{p'} C_{pp'} \psi_n(p) \psi_{n'}^*(p') w(p) w(p'). \end{aligned} \quad (12)$$

Next we select N_m such modes for comparison and consider the likelihood marginalized over the remainder of the modes

$$p(\mathbf{d}|\mathbf{C}[\Theta_C, \Theta_T, T], \varphi, A) \propto \frac{1}{\sqrt{|\mathbf{C} + \mathbf{N}|_M}} \times \exp\left\{-\frac{1}{2} \sum_{n=1}^{N_m} d_n^* (\mathbf{C} + \mathbf{N})_{nn'}^{-1} d_{n'}\right\}, \quad (13)$$

where \mathbf{C} and \mathbf{N} are restricted to the $N_m \times N_m$ block of chosen modes. Flexibility in choosing mode functions and their number N_m is used to achieve the compromise between the robust invertibility of the projected $\mathbf{C} + \mathbf{N}$ matrix on the one hand, and the amount of discriminating information retained in the data on the other. The weights $w(p)$ can be used to improve the accuracy of transforms on a pixelized sky.

For full-sky analysis the natural choice of the mode functions is the set of ordinary spherical harmonics $Y_{lm}(p)$ which leads to standard harmonic analysis with modes limited to a suitably chosen ℓ_{\max} . Here, where we focus on masked data, we have made a somewhat different choice. As a mode set for comparison we use the $N_m = 837$ largest eigenvectors of the $C_{pp'}$ matrix, restricted to the masked sky, for the fiducial flat isotropic model with best-fit parameters Θ_C . We emphasize that the correlation matrix computed for this reduced dataset has fewer modes, but contains no additional assumptions beyond those of the original $C_{pp'}$.

Since computation of $C_{pp'}$ matrices for a range of topological models is expensive, we do not aim to determine the full Bayesian evidence $P(\mathbf{d}|T)$ which would require marginalization over all parameters Θ_C, Θ_T , an overall amplitude of the correlation matrix A (proportional to the physical amplitude σ_8 or the scalar amplitude A_s), and orientation (Euler angles) φ , and would in addition be sensitive to the prior probabilities assumed for the size of the fundamental domain. Instead we directly compare the likelihood along the changing set of Θ_T that has as its limit the flat fiducial model defined by Θ_C . In case of toroidal topology such a limit is achieved by taking compactification lengths to infinity, while for curved models we vary Ω_K in comparison to the flat limit $\Omega_K = 0$. In the latter case, for the spherical spaces we change Ω_Λ and H_0 together with Ω_K to track the CMB geometrical degeneracy line in which the recombination sound speed, initial fluctuations, and comoving distance to the last scattering surface are kept constant (e.g., [Bond et al. 1997](#); [Zaldarriaga & Seljak 1997](#); [Stompor & Efstathiou 1999](#)), and for hyperbolic spaces we vary Ω_K while keeping H_0 and $\Omega_\Lambda - \Omega_m$ fixed to fiducial values. Note that hyperbolic multi-connected spaces, in contrast to tori and the single-action positive curvature manifolds considered in this paper, are not only anisotropic but also inhomogeneous. Therefore, the likelihood is expected to be dependent on the position of the observer. We do not study this dependence here.

For each parameter choice, we find the likelihood at the best orientation φ of the topology with respect to the sky after marginalizing over the amplitude A of the signal (hence, this can be considered a *profile likelihood* with respect to the orientation parameters). This likelihood is compared both with the fiducial model applied to the observed temperature map and with the likelihood of the topological model applied to the simulated realization of the isotropic map drawn from the fiducial model. Such a strategy is optimized for the detection of topological signatures. For non-detections, the marginalized likelihood can be a better probe of the overall power of the data to reject a non-trivial topology, and so for real data below, we also show the likelihood marginalized over the orientations φ . We estimate the marginalized likelihood from the random sample of 10 000 orientations, drawn statistically uniformly on the S^3 sphere of unit quaternions representing rotations of the fundamental domain relative to the observed sky.

5.2.2. Bianchi

For the Bianchi analysis the posterior distribution of the parameters of model M is given by Bayes' Theorem, specified in Eq. (10), similar to the topological setting. The approach of [McEwen et al. \(2013\)](#) is followed, where the likelihood is made explicit in the context of fitting a deterministic Bianchi template embedded in a stochastic CMB background, defined by the power spectrum $C_\ell(\Theta_C)$ for a given cosmological model with parameters Θ_C . The Bianchi VII_h parameters are denoted Θ_B . The corresponding likelihood is given by

$$P(\mathbf{d}|\Theta_B, \Theta_C) \propto \frac{1}{\sqrt{|\mathbf{X}(\Theta_C)|}} \exp[-\chi^2(\Theta_C, \Theta_B)/2], \quad (14)$$

where

$$\chi^2(\Theta_C, \Theta_B) = [\mathbf{d} - \mathbf{b}(\Theta_B)]^\dagger \mathbf{X}^{-1}(\Theta_C) [\mathbf{d} - \mathbf{b}(\Theta_B)] \quad (15)$$

and $\mathbf{d} = \{d_{\ell m}\}$ and $\mathbf{b}(\Theta_B) = \{b_{\ell m}(\Theta_B)\}$ are the spherical harmonic coefficients of the data and Bianchi template, respectively, considered up to the harmonic band-limit ℓ_{\max} . A band-limit of $\ell_{\max} = 32$ is considered in the subsequent analysis for computational tractability and since this is sufficient to capture the structure of the CMB temperature fluctuations induced in Bianchi VII_h models in the vicinity of the best-fit model found in WMAP data (see, e.g., [McEwen et al. 2006](#)). The likelihood is computed in harmonic space where rotations of the Bianchi template can be performed efficiently ([McEwen et al. 2006](#)).

The covariance matrix $\mathbf{X}(\Theta_C)$ depends on whether the full-sky or partial-sky masked setting is considered. In the full-sky setting $\mathbf{X}(\Theta_C) = \mathbf{C}(\Theta_C)$ as first considered by [Bridges et al. \(2007\)](#), where $\mathbf{C}(\Theta_C)$ is the diagonal CMB covariance matrix with entries $C_\ell(\Theta_C)$ on the diagonal. In the case of a zero Bianchi component, Eq. (14) then reduces to the likelihood function used commonly to compute parameter estimates from the power spectrum estimated from CMB data (e.g., [Verde et al. 2003](#)). In the masked setting considered subsequently, the situation is a little more involved.

In order to handle a mask in the harmonic space analysis of Bianchi models we follow the approach of [McEwen et al. \(2013\)](#), where *masking noise* is added to the data to effectively marginalise over the pixel values of the data in the masked region. The masking noise m is chosen to be zero-mean and large in the masked region of the data, and zero elsewhere. Consequently, the masking noise is anisotropic over the sky but

may be chosen to be uncorrelated, and may thus be defined by its covariance

$$\langle m(\omega_i) m^*(\omega_j) \rangle = \delta_{ij} \sigma_m^2(\omega_i), \quad (16)$$

where δ_{ij} is Kronecker delta symbol, ω_i denotes the angular coordinate of pixel i , and the variance of the noise for pixel i is given by a constant value in the masked regions $\sigma_m^2(\omega_i) = \Sigma_m^2$ and zero elsewhere. By synthetically adding masking noise that is much larger than the original data in the masked region of the sky, we effectively marginalise over the pixel values of the data in this region. The noisy mask introduces coupling in harmonic space that must be accounted for in the analysis. The covariance matrix of the resultant data is given by $\mathbf{X}(\Theta_C) = \mathbf{C}(\Theta_C) + \mathbf{M}$, where \mathbf{M} is the non-diagonal mask covariance matrix:

$$\mathbf{M}_{\ell m \ell' m'}^{\ell' m'} = \langle m_{\ell m} m_{\ell' m'} \rangle \simeq \sum_{\omega_i} \sigma_m^2(\omega_i) Y_{\ell m}^*(\omega_i) Y_{\ell' m'}(\omega_i) \Omega_i^2, \quad (17)$$

and Ω_i is the area of pixel i (see McEwen et al. 2013 for further details).

The χ^2 of the likelihood for the Bianchi case hence differs from the topology case by the nonzero Bianchi template \mathbf{b} and the use of a correlation matrix \mathbf{M} to account for the presence of the mask.

In the most physically motivated scenario, the Bianchi and cosmological parameters are coupled (e.g., the total density of the Bianchi and standard cosmological model are identical). However, it is also interesting to consider Bianchi templates as phenomenological models with parameters decoupled from the standard cosmological parameters, particularly for comparison with previous studies. Both scenarios are considered in the subsequent analysis. In the decoupled scenario a flat cosmological model is considered, whereas in the coupled scenario an open cosmological model is considered to be consistent with the Bianchi VII_h model; we label these models the flat-decoupled-Bianchi model and the open-coupled-Bianchi model, respectively.

To determine whether the inclusion of a Bianchi component better describes the data the Bayesian evidence is examined, as given by

$$E = P(\mathbf{d}|\mathbf{M}) = \int d\Theta P(\mathbf{d}|\Theta, \mathbf{M}) P(\Theta|\mathbf{M}). \quad (18)$$

Using the Bayesian evidence to distinguish between models naturally incorporates Occam's razor, trading off model simplicity and accuracy. In the absence of any prior information on the preferred model, the Bayes factor given by the ratio of Bayesian evidences (i.e., E_1/E_2) is identical to the ratio of the model probabilities given the data. The Bayes factor is thus used to distinguish models. The Jeffreys scale (Jeffreys 1961) is often used as a rule-of-thumb when comparing models via their Bayes factor. The log-Bayes factor $\Delta \ln E = \ln(E_1/E_2)$ (also called the log-evidence difference) represents the degree by which the model corresponding to E_1 is favoured over the model corresponding to E_2 , where: $0 \leq \Delta \ln E < 1$ is regarded as inconclusive; $1 \leq \Delta \ln E < 2.5$ as significant; $2.5 \leq \Delta \ln E < 5$ as strong; and $\Delta \ln E \geq 5$ as conclusive (without loss of generality we have assumed $E_1 \geq E_2$). For reference, a log-Bayes factor of 2.5 corresponds to odds of 1 in 12, approximately, while a factor of 5 corresponds to odds of 1 in 150, approximately.

The ANICOSMO⁶ code (McEwen et al. 2013) is used to perform a Bayesian analysis of Bianchi VII_h models, which in turn

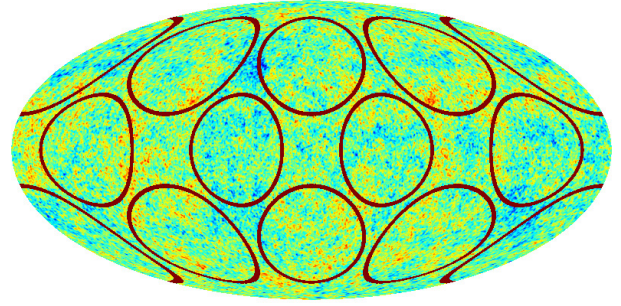


Fig. 4. A simulated map of the CMB sky in a universe with a $T[2, 2, 2]$ toroidal topology. The dark circles show the locations of the same slice through the last scattering surface seen on opposite sides of the sky. They correspond to matched circles with radius $\alpha \approx 24^\circ$.

uses the public MultiNest⁷ code (Feroz & Hobson 2008; Feroz et al. 2009) to sample the posterior distribution and compute evidence values by nested sampling (Skilling 2004). We sample the parameters describing the Bianchi VII_h model and those describing the standard cosmology simultaneously.

5.3. Simulations and Validation

5.3.1. Topology

Circles-in-the-Sky Before beginning the search for pairs of matched circles in the *Planck* data, we validate our algorithm using simulations of the CMB sky for a universe with 3-torus topology for which the dimension of the cubic fundamental domain is $L = 2H_0^{-1}$, and with cosmological parameters corresponding to the Λ CDM model (see Komatsu et al. 2011, Table 1) determined from the 7-year WMAP results combined with the measurements of the distance from the baryon acoustic oscillations and the Hubble constant. We performed simulations computing directly the $a_{\ell m}$ coefficients up to the multipole of order $\ell_{\max} = 500$ as described in Bielewicz & Banday (2011) and convolving them with the same smoothing beam profile as used for the data, i.e., a Gaussian beam with 30' FWHM. In particular, we verified that our code is able to find all pairs of matched circles in such a map. The map with marked pairs of matched circles with radius $\alpha \approx 24^\circ$ and the statistic $S_{\max}(\alpha)$ for the map are shown in Figs. 4 and 5, respectively. Note that the peak amplitudes in the statistic, corresponding to the temperature correlation for matched circles, decrease with radius of the circles. Cornish et al. (2004) noted that this is primarily caused by the Doppler term, which becomes increasingly anticorrelated for circles with radius smaller than 45° .

The intersection of the peaks in the matching statistic with the false detection level estimated for the CMB map corresponding to the simply-connected universe defines the minimum radius of the correlated circles which can be detected for this map. The height of the peak with the smallest radius seen in Fig. 5 indicates that the minimum radius is about $\alpha_{\min} \approx 20^\circ$.

For the Monte Carlo simulations of the CMB maps for the simply-connected universe we used the same cosmological parameters as for the multi-connected universe, i.e., corresponding to the Λ CDM model determined from the 7-year WMAP results. The maps were also convolved with the same beam profile as for the simulated map for the 3-torus universe and data, as well as masked with the same cut used for the analysis of data. The false

⁶ <http://www.jasonmcewen.org/>

⁷ <http://www.mrao.cam.ac.uk/software/multinest/>

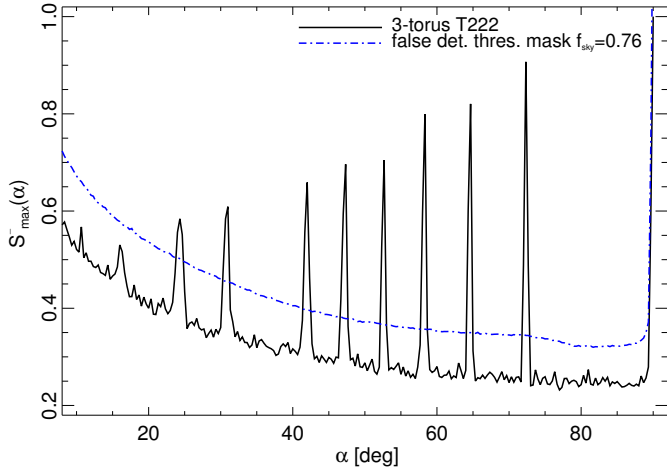


Fig. 5. An example of the S_{\max}^{-} statistic as a function of circle radius α for a simulated CMB map (shown in Fig. 4) of a universe with the topology of a cubic 3-torus with dimensions $L = 2H_0^{-1}$ (solid line). The dash-dotted line show the false detection level established such that fewer than 1% out of 300 Monte Carlo simulations of the CMB map, smoothed and masked in the same way as the data, would yield a false event.

detection threshold was established such that fewer than 1% of 300 Monte Carlo simulations would yield a false event.

Bayesian Analysis Because of the expense of the calculation of the correlation matrix, we wish to limit the number of three-dimensional wavevectors \mathbf{k} we consider, as well as the number of spherical harmonic modes ℓ , and finally the number of different correlation matrices as a whole. We need to ensure that the full set of matrices $C_{\ell\ell'}^{mm'}$ that we calculate contains all of the available information on the correlations induced by the topology in a sufficiently fine-grained grid. For this purpose, we consider the Kullback-Leibler (KL) divergence as a diagnostic (see, e.g., [Kunz et al. 2006, 2008](#), for applications of the KL divergence to topology). The KL divergence between two probability distributions $p_1(x)$ and $p_2(x)$ is given by

$$d_{\text{KL}} = \int p_1(x) \ln \frac{p_1(x)}{p_2(x)} dx. \quad (19)$$

If the two distributions are Gaussian with correlation matrices C_1 and C_2 , this expression simplifies to

$$d_{\text{KL}} = -\frac{1}{2} \left[\ln |C_1 C_2^{-1}| + \text{Tr} (I - C_1 C_2^{-1}) \right], \quad (20)$$

and is thus a measure of the discrepancy between the correlation matrices. The KL divergence can be interpreted as the ensemble average of the log-likelihood-ratio $\Delta \ln \mathcal{L}$ between realizations of the two distributions. Hence, they enable us to probe the ability to tell if, on average, we can distinguish realizations of p_1 from a fixed p_2 without having to perform a brute-force Monte Carlo integration. Thus, the KL divergence is related to ensemble averages of the likelihood-ratio plots that we present for simulations (Fig. 10) and real data (Sect. 6), but does not depend on simulated or real data.

We first use the KL divergence to determine the size of the fundamental domain which we can consider to be equivalent to the simply-connected case (i.e., the limit in which all dimensions of the fundamental domain go to infinity). We note that in our standard Λ CDM model, the distance to the surface of

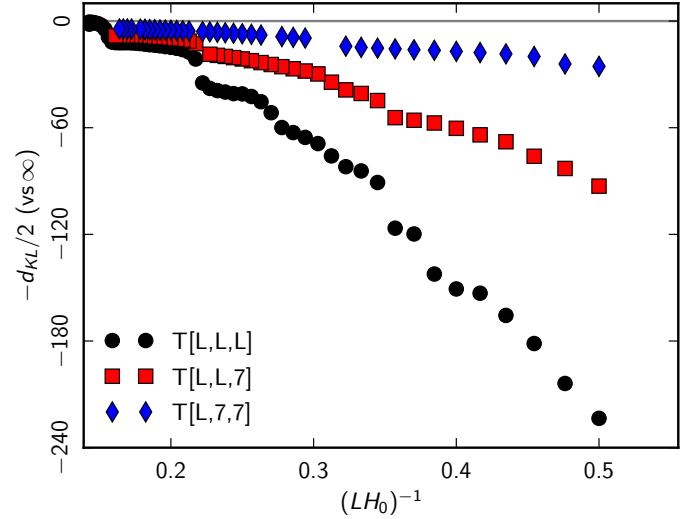


Fig. 6. KL divergence computed for torus models as a function of the (inverse) length of a side of the cube. $T[L_1, L_2, L_3]$ refers to a torus with edge lengths L_i .

last scattering is $\chi_{\text{rec}} \approx 3.1416(H_0)^{-1}$. We would naively expect that as long as the sphere enclosing the last scattering surface can be enclosed by the fundamental domain ($L = 2\chi_{\text{rec}}$), we would no longer see the effects of non-trivial topology. However, because the correlation matrix includes the full three-dimensional correlation information (not merely the purely geometrical effects of completely correlated points) we would see some long-scale correlation effects even for larger fundamental domains. In Fig. 6 we show the KL divergence (as a function of $(LH_0)^{-1}$ so that the simply-connected limit $L \rightarrow \infty$ is at a finite position) for the $T[L, L, L]$ (cubic), $T[L, L, 7]$ (chimney) and $T[L, 7, 7]$ (slab) spaces and show that it begins to level off for $(LH_0)^{-1} \lesssim 1/5$, although these topologies are still distinguishable from the $T[7, 7, 7]$ torus which is yet closer to the value for a simply-connected universe $d_{\text{KL}}[7, 7, 7] \approx 1.1$. These figures indicate that a length of $L = 7H_0^{-1}$ is an acceptable proxy for the simply-connected infinite Universe. The figures, as well as the likelihoods computed on simulations and data, show steps and other structures on a variety of scales corresponding to the crossing of the different length scales of the fundamental domain \mathcal{R}_u , \mathcal{R}_m , and \mathcal{R}_i crossing the last scattering surface; smaller fundamental domains with longer intersections with the last scattering surface are easier to detect.

Computational limitations further prevent us from calculating the likelihood at arbitrary values of the fundamental domain size parameters. We must therefore ensure that our coarse-grained correlation matrices are sufficient to detect a topology even if it lies between our gridpoints. In Fig. 7 we show the KL divergence as a function of the size of the fundamental domain, relative to various models, both aligned with our grid ($LH_0 = 4.5$) and in between our grid points ($LH_0 = 5.25$). We see that the peak is wide enough that we can detect a peak within $\delta LH_0 \sim 0.1$ of the correct value. We also show that we can detect anisotropic fundamental domains even when scanning through cubic tori: we show a case which approximates a ‘‘chimney’’ universe with one direction much larger than the distance to the last scattering surface.

Because our topological analyses do not simultaneously vary the background cosmological parameters along with those describing the topology, we also probe the sensitivity to the cosmology. In Fig. 8 we show the effect of varying the fiducial

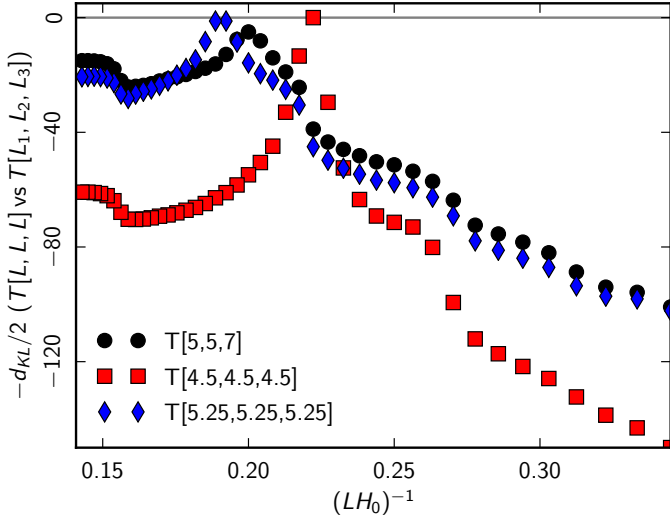


Fig. 7. KL divergence between a supposed correct model and other models. We show differences of cubic tori with respect to models with $(LH_0)^{-1} = 1/4.5 \approx 0.22$ (aligned with our grid of models), $(LH_0)^{-1} = 1/5.25 \approx 0.19$ (in between the gridpoints) and a $T[5, 5, 7]$ chimney model with $(LH_0)^{-1} = 1/5$ in two directions and $(LH_0)^{-1} = 1/7 \approx 0.14$ in the third.

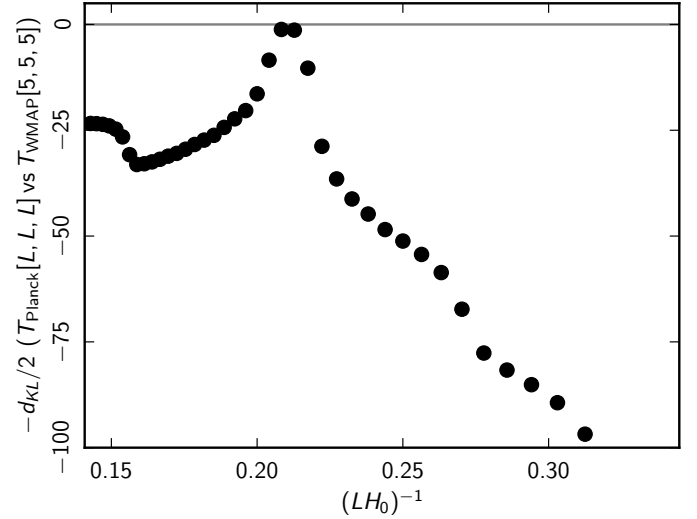


Fig. 8. KL divergence between a model generated with the WMAP best-fit cosmological parameters as a background cosmology and a $T[5, 5, 5]$ cubic torus topology with respect to a *Planck* best-fit cosmology and a varying cubic topology.

cosmology from the [Planck Collaboration XVI \(2014\)](#) best-fit values to those reported by WMAP ([Komatsu et al. 2011](#))⁸. We see that this induces a small bias of $\delta LH_0 \approx 0.2$ but does not hinder the ability to detect a non-trivial topology. This indicates that small deviations from the correct background cosmology do not hinder our ability to detect (or rule out) topological signals.

We have also directly validated the topological Bayesian techniques with simulations. In [Fig. 9](#) we show the log-likelihood for the above $T[2, 2, 2]$ simulations as a function of two of the Euler angles, maximized over the third. We find a strong peak at the correct orientation, with a multiplicity due to the degenerate orientations corresponding to the faces of the cube (there are peaks at the North and South poles, which are difficult to see in this projection). Note that the peaks correspond to ratios of more than $\exp(700)$ compared to the relatively smooth minima elsewhere.

In [Fig. 10](#) we also test the ability of the Bayesian likelihood technique to detect the compactification of the space in the simulated temperature realizations drawn from the dodecahedral closed model. For curved geometries, the size of the fundamental domain is fixed with respect to the varying curvature scale (R_0), whereas the distance to the last scattering χ_{rec} is constant. Hence we plot the likelihood as a function of χ_{rec}/R_0 , inversely proportional to the scale of the fundamental domain.

Two multiply-connected realizations of the sky were tested: one corresponding to the space in which the last scattering sphere can be just inscribed into the fundamental domain, $\chi_{\text{rec}} = R_i$, when just the first large angle correlations appear, and the second drawn from a somewhat smaller space for which $\chi_{\text{rec}} = R_e$. We see detections in both cases, stronger as the fundamental domain shrinks relative to χ_{rec} . We also calculate the likelihood for a model known to be simply-connected.

Note that the likelihood taken at the best orientations of the compact models generically shows a slight increase relative to that for the limiting simply connected space as one brings the

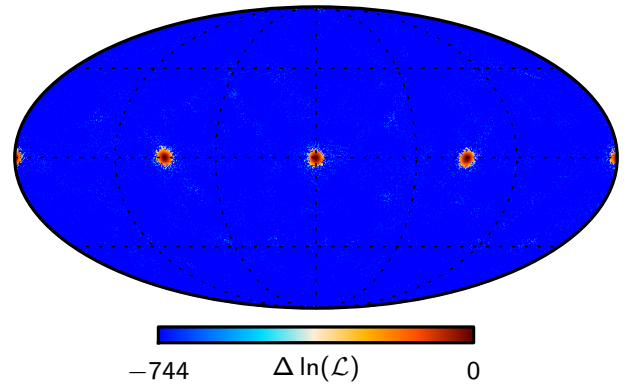


Fig. 9. Log-likelihood with respect to the peak as a function of the orientation of the fundamental $T[2, 2, 2]$ torus domain for the simulations. The third Euler angle is marginalized over. We see peaks at the orientations corresponding to the six faces of the cubic fundamental domain (there are peaks at the North and South poles, which are difficult to see in this projection).

size of the fundamental domain down to the size of the last scattering surface ($\chi_{\text{rec}} \approx R_i$), followed, in the absence of signal in the map, by a rapid drop as soon as the models smaller than χ_{rec} are applied. This small increase is also present in the fiducial exactly isotropic sky, a single realization of which is shown in the figure, but is a generic feature irrespective of the topology being tested (occurring also in models with $R < R_i$), and thus should not be taken as an indication for compact topology. The reason for the increase is the possibility of aligning the model with a weak anisotropic correlation feature with chance patterns of a single sky realization. However the fit drastically worsens as soon as the correlation features in a model become pronounced. Moreover, the feature becomes considerably less significant when the likelihood is marginalized over the orientation (Euler angles) of the fundamental domain.

All of these results (KL divergences and likelihoods) were computed with $\ell_{\text{max}} = 40$, corresponding approximately to $N_{\text{side}} = 16$, indicating that this is more than adequate for detecting even relatively small fundamental domains such as the

⁸ We use the `wmap7+bao+h0` results from <http://lambda.gsfc.nasa.gov>.

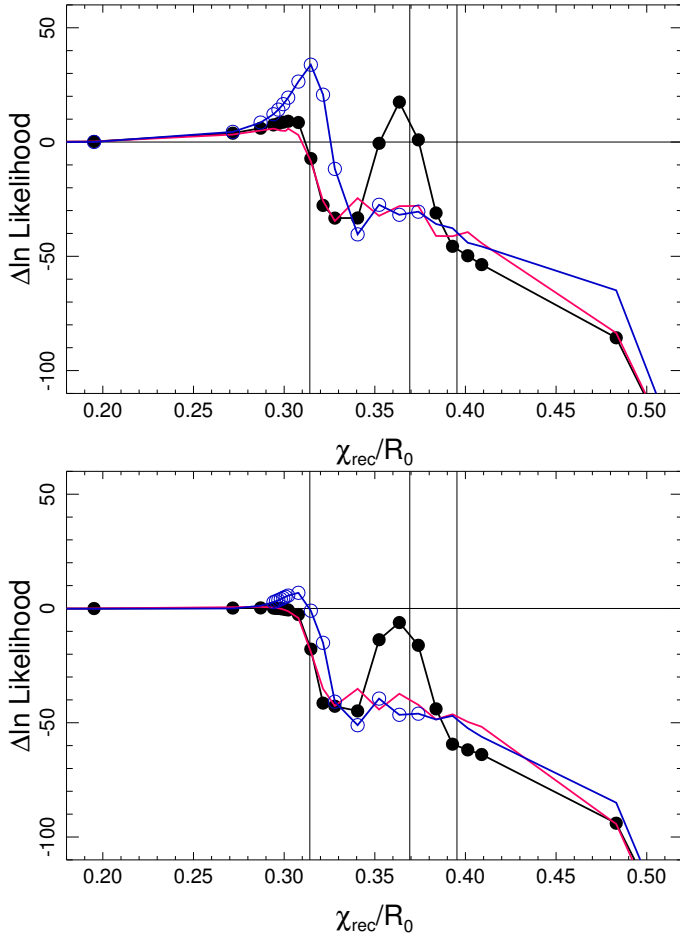


Fig. 10. Test for likelihood detectability of compactified space for the example of a dodecahedral (I^*) closed universe. The vertical axis shows the log-likelihood relative to the largest model considered. Values are given for the orientations of the models which maximize the likelihood (*top*) and marginalized over the orientations (*bottom*). Different size models are tested against two HEALPix $N_{\text{side}} = 16$ temperature realizations drawn from the model with $\chi_{\text{rec}}/R_0 = 0.314 = \mathcal{R}_i$ (blue) and $\chi_{\text{rec}}/R_0 = 0.361$ (black). No noise is added and the common mask has been applied. Dots mark the positions of the models for which the likelihoods were computed. The vertical lines show characteristic scales of the fundamental domain of the models in the units of curvature, from smaller to larger, \mathcal{R}_i/R_0 , \mathcal{R}_m/R_0 and \mathcal{R}_u/R_0 . The variable χ_{rec}/R_0 gives the size of the last scattering surface in the same units. The $R_0 \rightarrow \infty$ limit corresponds to the flat simply-connected space. Both maximized and marginalized likelihoods show a detection relative to the isotropic sky realization drawn from the fiducial flat infinite universe (red) with the detection stronger for smaller spaces. However only the maximized likelihood unambiguously distinguishes the correct compact model from spaces that exceed the last-scattering diameter, which shows that the likelihood for small models is narrowly peaked at the correct orientation and suppressed otherwise.

$T[2, 2, 2]$ case simulated above. We also calculate d_{KL} between the correlation matrices for the $T[7, 7, 7]$ torus (as a proxy for the simply-connected case) and the $T[5, 5, 5]$ torus, as a function of the maximum multipole ℓ_{max} used in the calculation of the correlation matrix: we find that d_{KL} continues to increase beyond $\ell_{\text{max}} = 60$. Thus, higher-resolution maps (as used by the matched-circles methods) contain more information, but with the very low level of noise in the *Planck* CMB maps, $\ell_{\text{max}} = 40$ would nonetheless give a robust detection of a multiply-connected topology, even with the conservative foreground masking we apply.

We note that it is difficult to compress the content of these likelihood figures down to limits upon the size of the fundamental domain. This arises because it is difficult to provide a physically-motivated prior distribution for quantities related to the size of the fundamental domain. Most naive priors would diverge towards arbitrarily large fundamental domain sizes or would otherwise depend on arbitrary limits to the topological parameters.

5.3.2. Bianchi

The ANICOSMO code (McEwen et al. 2013) is used to perform a Bayesian analysis of Bianchi VII_h models, which has been extensively validated by McEwen et al. (2013) already; we briefly summarise the validation performed for the masked analysis. In McEwen et al. (2013) a CMB map is simulated, in which a simulated Bianchi temperature map with a large vorticity (i.e., amplitude) is embedded, before applying a beam, adding isotropic noise and applying a mask. Both the underlying cosmological and Bianchi parameters used to generate the simulations are well recovered. For this simulation the coupled Bianchi model is favoured over ΛCDM , with a log-Bayes factor of $\Delta \ln E \sim 50$. As expected, one finds that the log-Bayes factor favours ΛCDM in simulations where no Bianchi component is added. For further details see McEwen et al. (2013).

6. Results

We now discuss the results of applying the circles-in-the-sky and likelihood methods to *Planck* data to study topology and Bianchi VII_h cosmologies.

6.1. Topology

Neither the circles-in-the-sky search nor the likelihood method find evidence for a multiply-connected topology. We show the matched circle statistic in Fig. 11. We do not find any statistically significant correlation of circle pairs in any map. As seen in Fig. 5, the minimum radius at which the peaks expected for the matching statistic are larger than the false detection level is $\alpha_{\text{min}} \approx 20^\circ$. Thus, we can exclude at the confidence level of 99% any topology that predicts matching pairs of back-to-back circles larger than this radius, assuming that relative orientation of the fundamental domain and mask allows its detection. This implies that in a flat universe described otherwise by the *Planck* fiducial ΛCDM model, a 99% confidence-limit lower bound on the size of the fundamental domain is $L/2 \gtrsim \chi_{\text{rec}} \cos(\alpha_{\text{min}}) = 0.94\chi_{\text{rec}} = 13.2\text{Gpc}$. This is better than the limits from the marginalized likelihood ratios below for the tori and octahedron topologies and slightly worse than the limits for the dodecahedron and truncated cube. However, this constraint is not limited only to these few topologies. The frequentist analysis provides constraints upon a much wider class of topologies than those explicitly considered in the Bayesian likelihood approach; it concerns all topologies listed in Sect. 3.1.

The likelihood method also show no evidence of a multiply connected universe. We present the likelihood for various models. In Fig. 12 we show the likelihood (marginalized over amplitude and maximized over orientation of the fundamental domain) for the cubic torus, fixing the background cosmology to the best-fit flat Universe *Planck* model (Planck Collaboration XVI 2014). We see that this is maximized for $L > 2\chi_{\text{rec}}$, i.e., showing no evidence for non-trivial topology.

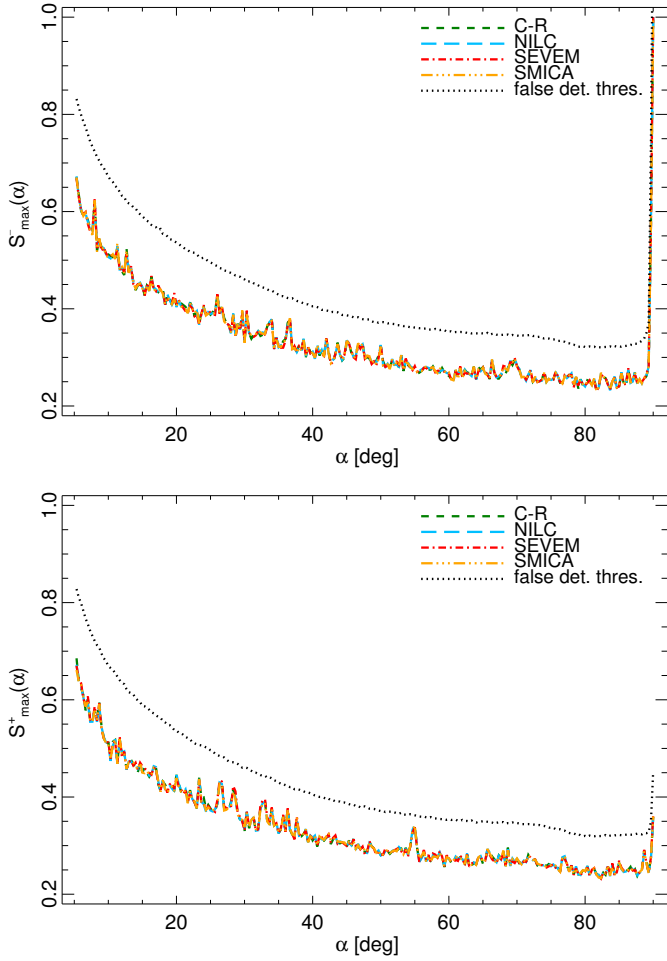


Fig. 11. The S_{\max}^- (upper) and S_{\max}^+ (lower) statistics as a function of circle radius α for the *Planck* CMB maps estimated using Commander-Ruler (short-dashed green line), NILC (blue long dashed line), SEVEM (dot-dashed red line) and SMICA (orange three dots-dashed line). Dotted line shows the false detection level established such that fewer than 1% out of 300 Monte Carlo simulations of the CMB map, smoothed and masked in the same way as the data, would yield a false event. The peak at 90° corresponds to a match between two copies of the same circle of radius 90° centered around two antipodal points.

Note that the likelihood shows mild features as the size goes through the other scales associated with the topology, in particular a small increase in the likelihood when the scale of the inscribed sphere \mathcal{R}_i is crossed. However, the same increase is found when the toroidal model is compared to a single realization of a strictly isotropic fiducial sky, and thus, should not be interpreted as a detection of multi-connected topology. The origin of this likelihood behaviour at best fit angles is that the freedom of orientation can be used to align small enhancements in large-angle correlations in the anisotropic $L \approx 2\mathcal{R}_i$ model with random features in the given single realization of the sky. When marginalized over all possible orientations the effect is significantly reduced; the slight rise is $\Delta \ln \mathcal{L} \approx 1.9$ from a likelihood of $P = 650$, which is comparable to the numerical noise inherent in our stochastic integration. For even smaller spaces, more extensive correlations of the temperature can no longer be accommodated and for $L < 2\mathcal{R}_m$ the likelihood of the T3 cubic toroidal model drops quickly, although not strictly monotonically.

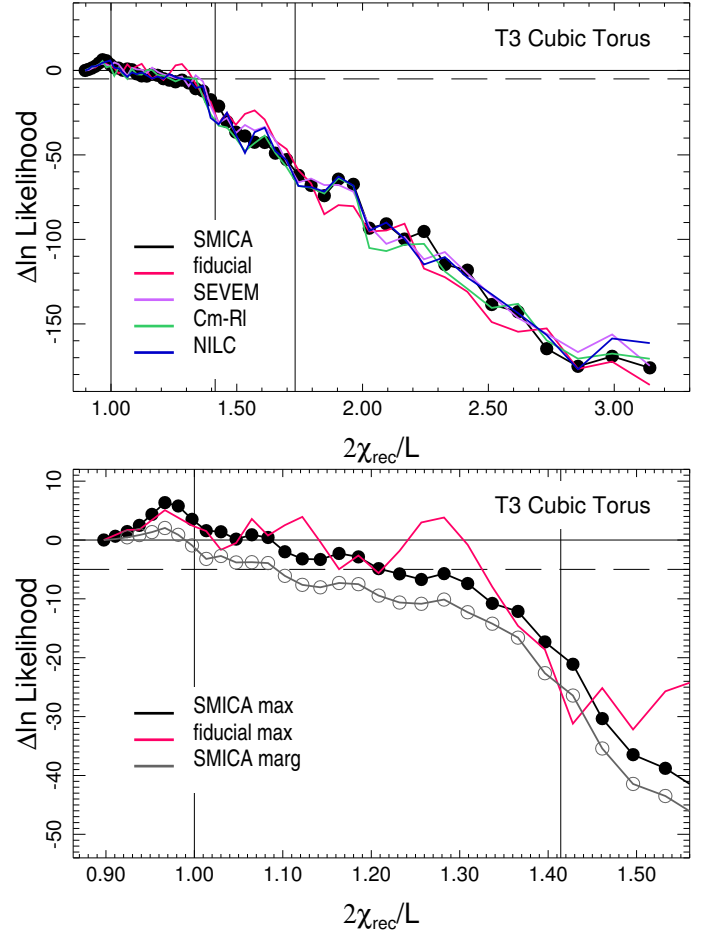


Fig. 12. *Top:* likelihood as a function of the length of an edge of the fundamental domain L for a cubic-torus topology. In this figure, χ_{rec} gives the distance to the surface of recombination. The data are component-separated CMB temperature maps degraded to HEALPix $N_{\text{side}} = 16$ resolution and smoothed with an $FWHM = 660'$ Gaussian filter. The common mask of $f_{\text{sky}} = 0.78$ is used. The likelihood is marginalized over the amplitude of fluctuations, but maximized over the orientation of the fundamental domain. Lines for different estimates of the CMB temperature from *Planck* data are black: SMICA; magenta: SEVEM; green: Commander-Ruler; blue: NILC. The red line is for a simulated isotropic sky from a fiducial flat simply-connected model. Noise has been accounted for but is negligible at $N_{\text{side}} = 16$. The likelihoods are normalized to match the likelihood obtained with the common mask in the $R_0 \rightarrow \infty$ isotropic flat limit. The vertical lines mark the positions where χ_{rec} is equal to the characteristic sizes of the fundamental domain, from left to right, $\mathcal{R}_i = L/2$, $\mathcal{R}_m = \sqrt{2}L/2$ and $\mathcal{R}_u = \sqrt{3}L/2$. Dots, superimposed onto the SMICA curve, designate the discrete set of models studied. *Bottom:* zoom into the transitional region near $\chi_{\text{rec}} \approx \mathcal{R}_i$. Black *Planck* SMICA and red fiducial curves are the same as in the top panel. The grey curve (open circles) is the likelihood marginalized over the orientations for the *Planck* SMICA map. Only \mathcal{R}_i and \mathcal{R}_m are within the scale range shown.

In Figs. 13 and 14 we show the likelihood for the $T[L, L, 7]$ chimney and $T[L, 7, 7]$ slab topologies, which are also maximized in the simply-connected limit. The T2 chimney, with only two compact dimensions, is less constrained than the T3 cube, and the T1 slab, with one compact dimension, even less so.

We find similar limits for the topologies allowed in a closed universe with a locally spherical geometry. In Fig. 15 we show the likelihood for the dodecahedral fundamental domain, in

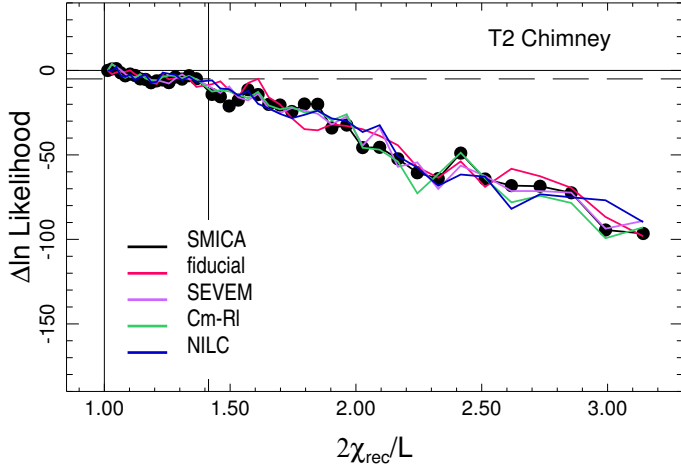


Fig. 13. Same as Fig. 12, but for a toroidal space with one large dimension fixed at $7H_0^{-1}$ and two short dimensions of equal size L (approximating the “chimney” space). \mathcal{R}_i and \mathcal{R}_m are marked while $\mathcal{R}_u = \infty$

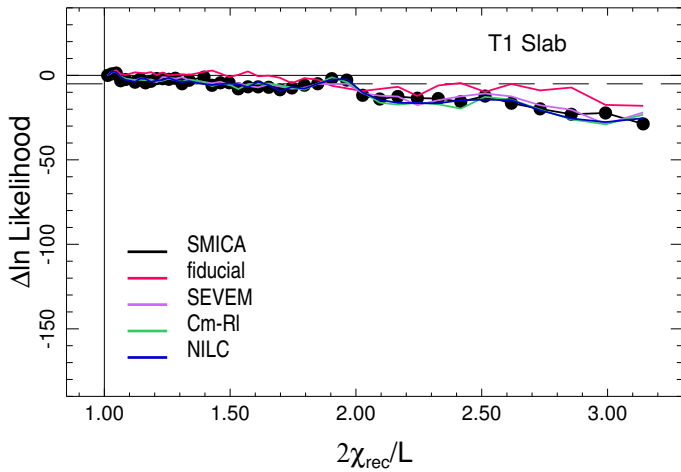


Fig. 14. Same as Fig. 12, but for a toroidal space with two large dimensions fixed at $7H_0^{-1}$ and one short dimension of variable L (approximating the “slab” space). \mathcal{R}_i is marked while $\mathcal{R}_m = \mathcal{R}_u = \infty$.

Fig. 16 for the truncated cube, and in Fig. 17 for the octahedron. In this case, we do not fix the background cosmological model, but rather account for the geometrical degeneracy line which links H_0 and Ω_Λ with Ω_K . The degeneracy relations are approximated as $\Omega_\Lambda = 0.691 + 2.705\Omega_K$ and $H_0 = 67.8 + 388\Omega_K + 1200\Omega_K^2$. As in the toroidal case, there is no detection of a small space at the level expected from the simulations of Sect. 5. Fundamental domains larger than the last scattering diameter are preferred for the dodecahedral and truncated cube spaces with somewhat weaker restriction for the octahedral case. Note that an observationally motivated prior on H_0 or Ω_K would be yet more restrictive on the fundamental domain size. For all three topologies, again as in the toroidal case, the maximum of the likelihood at best fit orientation is detected for the finite volume spaces with $\chi_{\text{rec}} \approx \mathcal{R}_i$ at the level $\Delta \ln \mathcal{L} \approx +4$ relative to the fiducial flat simply-connected model. Since this feature is seen in the isotropic fiducial sky as well, we cannot take it as an indication of a detection of a multi-connected space. In the case of curved spaces we see that this mild increase disappears when we consider the likelihood marginalized over orientations.

We present numerical limits for these flat and positively curved spaces in Table 2. Because of the one-sided nature of these limits, we characterize the shape of the likelihood by the

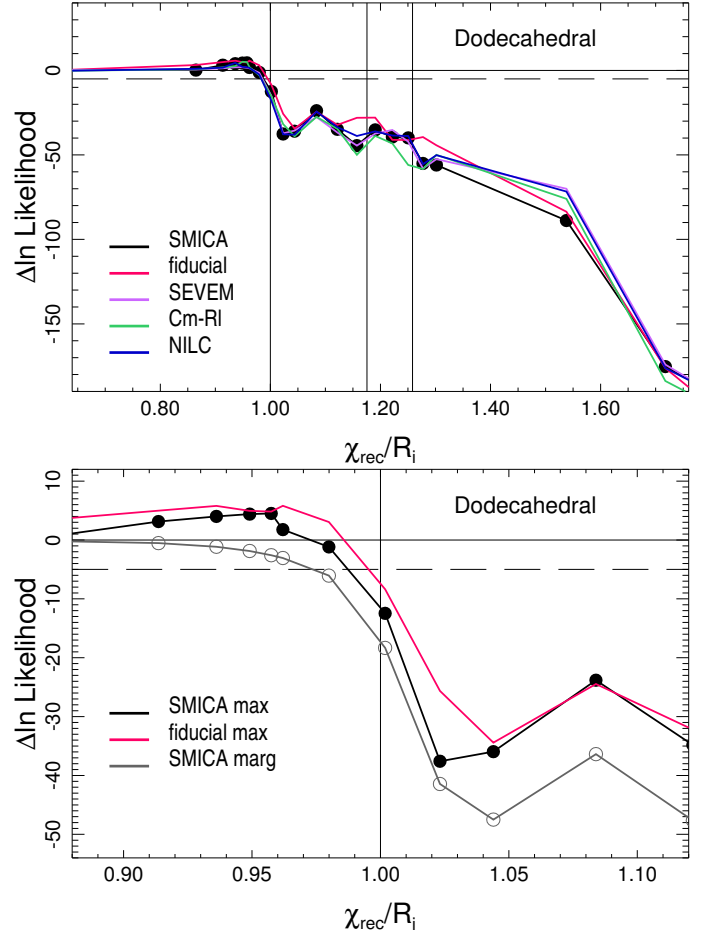


Fig. 15. *Top:* likelihood as a function of the distance to last scattering surface in curvature units for a locally spherical multiply-connected universe with a dodecahedral (I^*) fundamental domain with $\mathcal{R}_i = 0.31R_0$. Lines are for different estimates of the CMB temperature from *Planck* data as in Fig. 12. In this figure, the χ_{rec}/R_0 parameterizes the position of the model on the geometrical degeneracy line which links H_0 and Ω_Λ with Ω_K . The degeneracy relations are approximated as $\Omega_\Lambda = 0.691 + 2.705\Omega_K$ and $H_0 = 67.8 + 388\Omega_K + 1200\Omega_K^2$. The red reference curve is for the random isotropic realization of a fiducial flat model. Vertical lines mark when χ_{rec} equals each of \mathcal{R}_i , \mathcal{R}_m , and \mathcal{R}_u , the characteristic scales of the fundamental domain. *Bottom:* zoom into the transitional region near $\chi_{\text{rec}} \approx \mathcal{R}_i$. Both the likelihood at the best orientation of the domain versus the sky (black for the *Planck* SMICA CMB map and red for the fiducial realization, as in the top panel) and the likelihood marginalized over the orientations for *Planck* SMICA map (gray curve, open circles) are shown.

steepness of its fall from the value as the scale of the fundamental domain goes to infinity (i.e., the simply-connected limit). Hence, we show limits for $\Delta \ln \mathcal{L} < -5$, (roughly equivalent to a $3\sigma - 99\%$ confidence limit – fall for a Gaussian; because of the very steep gradient, the 2σ limits are very similar) and $\Delta \ln \mathcal{L} < -12.5$ (5σ). Note that the limits differ depending on whether we marginalize or maximize the likelihood over the orientation angles. We show lower limits on the quantity \mathcal{R}_i ($L/2$ for a torus with edge length L) in units of the last scattering distance χ_{rec} (in conventional units, $\chi_{\text{rec}} \approx 14$ Gpc for the fiducial *Planck* parameters; [Planck Collaboration XVI 2014](#)). In most cases, the limits are roughly $\mathcal{R}_i \gtrsim \chi_{\text{rec}}$ – the scale of the fundamental domain must be greater than that of the last scattering surface. We place the most restrictive limits on the dodecahedron with $\mathcal{R}_i > 1.03\chi_{\text{rec}}$ using marginalized values for $\Delta \ln \mathcal{L} < -5$.

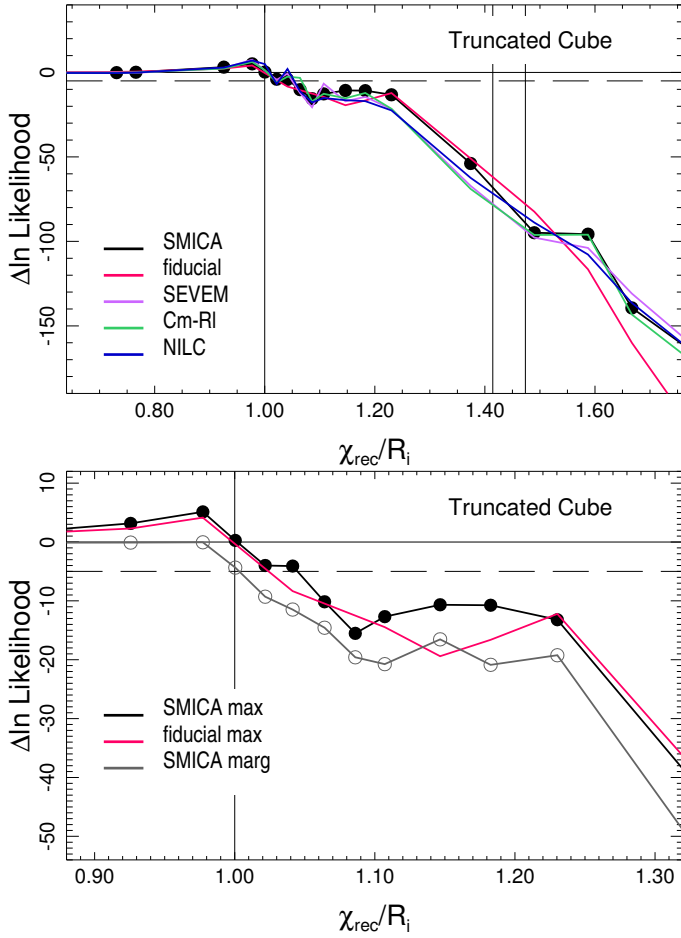


Fig. 16. Likelihood for a constant positive curvature multiply-connected universe with a truncated cube (O^*) fundamental domain with $R_i = 0.39R_0$. Notation is the same as in Fig. 15.

Conversely, the chimney and slab spaces are less constrained as the expected correlations are weaker in one or two directions; for the slab space, we only constrain $R_i = L/2 \gtrsim 0.5\chi_{\text{rec}}$.

In Fig. 18 we show the likelihood for the two hyperbolic models listed in Table 1, which also show no detection of the multi-connected topology. In the hyperbolic case we space the range of space sizes by varying Ω_K while keeping $\Omega_\Lambda - \Omega_m$ as well as H_0 constant at fiducial values.

All of these results show at least some increase in the likelihood for certain orientations when one of the characteristic scales of the fundamental domain (\mathcal{R}_u , \mathcal{R}_m , or \mathcal{R}_i) just exceed the surface of last scattering, and so no longer produces matched patterns, but induces extra correlations at large angular separations. Chance patterns can then mimic these correlations, and this is exacerbated by our conservative sky masks, which allow arbitrary patterns in the masked regions.

6.2. Bianchi

Masked *Planck* data are analysed for evidence of a Bianchi VII_h component, where the prior parameter ranges adopted are the same as those specified by McEwen et al. (2013). The analysis is performed on the SMICA component-separated map, using the mask defined for this method, and is repeated on the SEVEM component-separated map for validation purposes (using the mask defined for the SEVEM method). The Bayes factors for the various Bianchi VII_h models and the equivalent standard cosmological models are shown in Table 3.

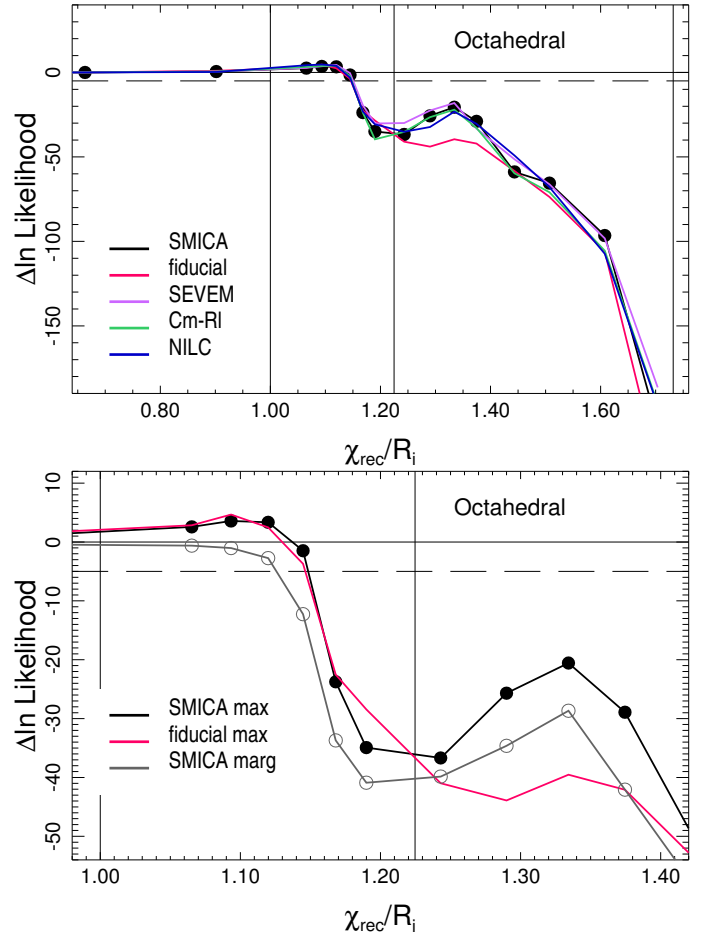


Fig. 17. Likelihood for a constant positive curvature multiply-connected universe with an octahedral (T^*) fundamental domain with $R_i = 0.45R_0$. Notation is the same as in Fig. 15.

Table 2. Lower limits on the size of the fundamental domain for different multiply-connected spaces, in units of the distance to the last scattering surface, χ_{rec} .

Space	Quantity	$\Delta \ln \mathcal{L} < -5$		$\Delta \ln \mathcal{L} < -12.5$	
		max	marg	max	marg
T3 cubic torus	$L/(2\chi_{\text{rec}})$	0.83	0.92	0.76	0.83
T2 chimney	$L/(2\chi_{\text{rec}})$	0.71	0.71	0.63	0.67
T1 slab	$L/(2\chi_{\text{rec}})$	0.50	0.50
Dodecahedron	$\mathcal{R}_i/\chi_{\text{rec}}$	1.01	1.03	1.00	1.01
Truncated cube	$\mathcal{R}_i/\chi_{\text{rec}}$	0.95	1.00	0.81	0.97
Octahedron	$\mathcal{R}_i/\chi_{\text{rec}}$	0.87	0.89	0.87	0.88

Notes. For the torus, slab, and chimney, we present limits on the quantity $L/2$; in curved spaces, limits are on the inscribed-sphere topology scale \mathcal{R}_i . For the columns labelled “max”, we maximize the probability over the orientation of the fundamental domain; for “marg”, we marginalize over orientation.

For the phenomenological flat-decoupled-Bianchi model, evidence in support of a left-handed Bianchi template is found. On the Jeffreys scale (Jeffreys 1961), evidence for this model would be referred to as strong for the SMICA map and significant for the SEVEM map. For both SMICA and SEVEM component-separated data, recovered posterior distributions for the flat-decoupled-Bianchi model are shown in Fig. 19a, where similar

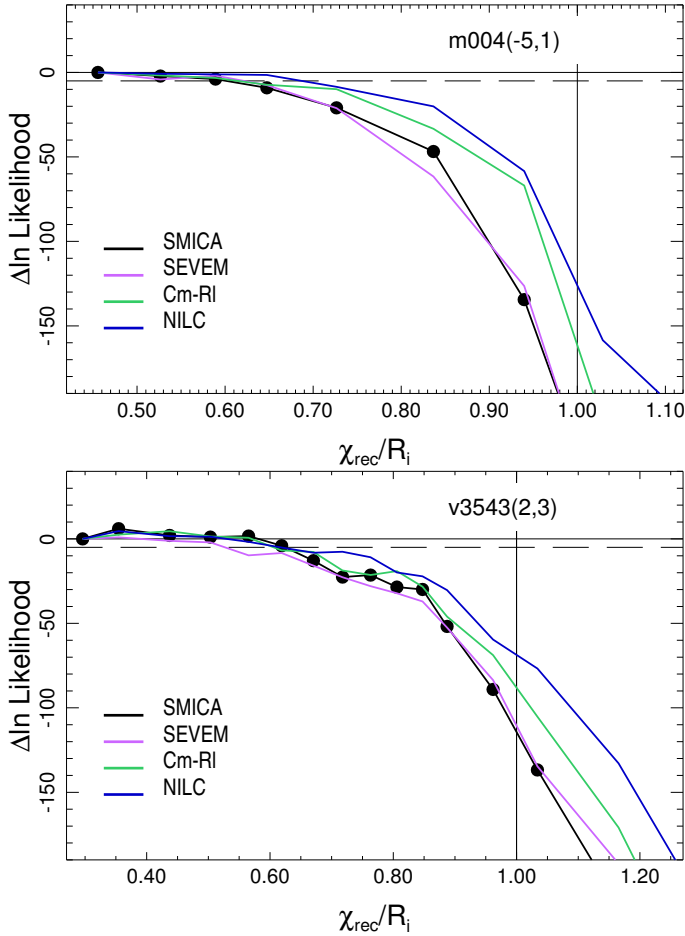


Fig. 18. Likelihood for two constant negative curvature multiply-connected universe, *top*: m004(-5,1); *bottom*: v3543(2,3). Notation is as in Fig. 15 except that only \mathcal{R}_i/R_0 is shown by vertical lines.

Table 3. Log-Bayes factor relative to equivalent Λ CDM model (positive favours Bianchi model).

Model	SMICA	SEVEM
Flat-decoupled-Bianchi (left-handed)	2.8 ± 0.1	1.5 ± 0.1
Flat-decoupled-Bianchi (right-handed)	0.5 ± 0.1	0.5 ± 0.1
Open-coupled-Bianchi (left-handed)	0.0 ± 0.1	0.0 ± 0.1
Open-coupled-Bianchi (right-handed)	-0.4 ± 0.1	-0.4 ± 0.1

posterior distributions are recovered for both component separation methods. Recall that the Bianchi parameters are decoupled from the standard cosmology in the flat-decoupled-Bianchi model, hence for this model Ω_m^B and Ω_Λ^B are specific to the Bianchi model and should not be compared with standard values. The maximum a posteriori (MAP) best-fit template found for SMICA component-separated data is shown in Fig. 20b, with the difference between this template and the template found in WMAP 9-year data (McEwen et al. 2013) shown in Fig. 21. Note that the template found in *Planck* data is very similar to the template found in WMAP 9-year data (McEwen et al. 2013), which in turn is similar to the template first found by Jaffe et al. (2005). However, the template found in WMAP 9-year data (McEwen et al. 2013) is only significant in full-sky data, but not when the 9-year KQ75 WMAP mask (Bennett et al. 2013) is applied. Since the *Planck* SMICA and SEVEM masks

are less conservative than the KQ75 mask, these findings suggest data near the Galactic plane may be playing a considerable role in supporting a Bianchi component in *Planck* data. The SMICA CMB map and a Bianchi-subtracted version of this map are also shown in Fig. 20. The best-fit parameters of the templates found in *Planck* SMICA and SEVEM component-separated data are displayed in Table 4, for both the MAP and mean-posterior estimates. The analysis was also performed on a SMICA component-separated Gaussian simulation, yielding a null detection (i.e., no evidence for a Bianchi component), as expected.

For the most physically motivated open-coupled-Bianchi model where the Bianchi VII_h model is coupled to the standard cosmology, there is no evidence in support of a Bianchi contribution. Recovered posterior distributions for the open-coupled-Bianchi model are shown in Fig. 19b for both SMICA and SEVEM component-separated data. Although the cosmological Bianchi parameters agree reasonably well between these different component-separated data, the posterior distributions recovered for the Euler angles differ. For SEVEM data, an additional mode of the posterior distribution is found; the mode found with SMICA data is still present in SEVEM data but is not dominant. Consequently, the best-fit estimates for the Euler angles differ between the SMICA and SEVEM component-separated data. Note that the additional mode found in SEVEM data is also present in WMAP 9-year data (McEwen et al. 2013). The resulting best-fit parameters for the open-coupled-Bianchi model are displayed in Table 5, while the corresponding MAP best-fit maps are shown in Fig. 22. Nevertheless, for both SMICA and SEVEM data the Bayes factors computed (Table 3) do not favour the inclusion of any Bianchi component for the open-coupled-Bianchi model. *Planck* data thus do not provide evidence in support of Bianchi VII_h cosmologies. However, neither is it possible to conclusively discount Bianchi VII_h cosmologies in favour of Λ CDM cosmologies. The constraints $(\omega/H)_0 < 7.6 \times 10^{-10}$ (95% confidence level) on the vorticity of the physical coupled Bianchi VII_h left-handed models and $(\omega/H)_0 < 8.1 \times 10^{-10}$ (95% confidence level) for right-handed models are recovered from SMICA component-separated data.

7. Discussion

We have used the *Planck* temperature anisotropy maps to probe the large-scale structure of spacetime. We have calculated the Bayesian likelihood for specific topological models in universes with locally flat, hyperbolic and spherical geometries, all of which find no evidence for a multiply-connected topology with a fundamental domain within the last scattering surface. After calibration on simulations, direct searches for matching circles resulting from the intersection of the fundamental topological domain with the surface of last scattering also give a null result at high confidence. These results use conservative masks of the sky, unlike previous WMAP results, which used full-sky internal linear combination maps (not originally intended for cosmological studies) or less conservative foreground masks. Hence, the results presented here, while corroborating the previous non-detections, use a single, self-consistent, and conservative dataset. The masked sky also increases the possibility of chance patterns in the actual sky mimicking the correlations expected for topologies with a characteristic scale near that of the last scattering surface.

Depending on the shape of the fundamental domain, we find $\mathcal{R}_i \gtrsim \chi_{\text{rec}}$ (Table 2) with detailed 99% confidence limits (considering the likelihood marginalized over the orientation of the fundamental domain) varying from $0.9\chi_{\text{rec}}$ for the cubic

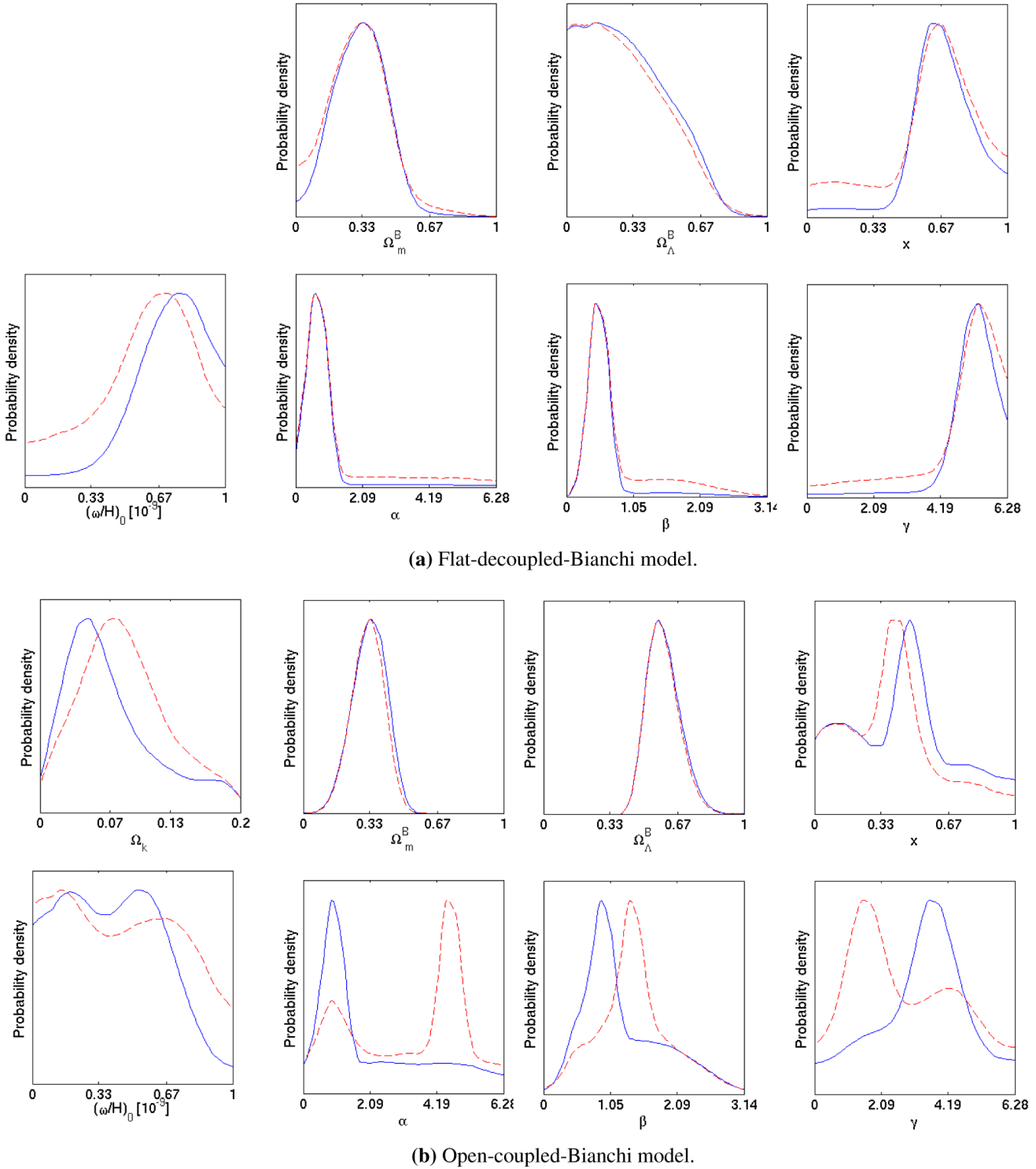


Fig. 19. Posterior distributions of Bianchi parameters recovered from *Planck* SMICA (solid curves) and SEVEM (dashed curves) component-separated data for left-handed models. *Planck* data provide evidence in support of a Bianchi component in the phenomenological flat-decoupled-Bianchi model (panel a)) but not in the physical open-coupled-Bianchi model (panel b)).

torus in a flat universe to $1.03\chi_{\text{rec}}$ for the dodecahedron in a positively curved universe, with somewhat weaker constraints for poorly-proportioned spaces that are considerably larger along some directions. In the case of the torus and octahedron topologies, a tighter constraint of $0.94\chi_{\text{rec}}$ comes from the matched circles method (albeit with a somewhat different interpretation of frequentist and Bayesian limits). The constraint derived using this method applies to a wide class of topologies,

listed in Sect. 3.1, predicting matching pairs of back-to-back circles.

Note that the results derived using the likelihood method make use of the expected pixel-space correlations as a unique signal of non-trivial topology. Hence, although a small fundamental domain will suppress power on the largest scales of the CMB, observation of such low power on large scales as observed by COBE (Bond et al. 2000c), and confirmed by WMAP

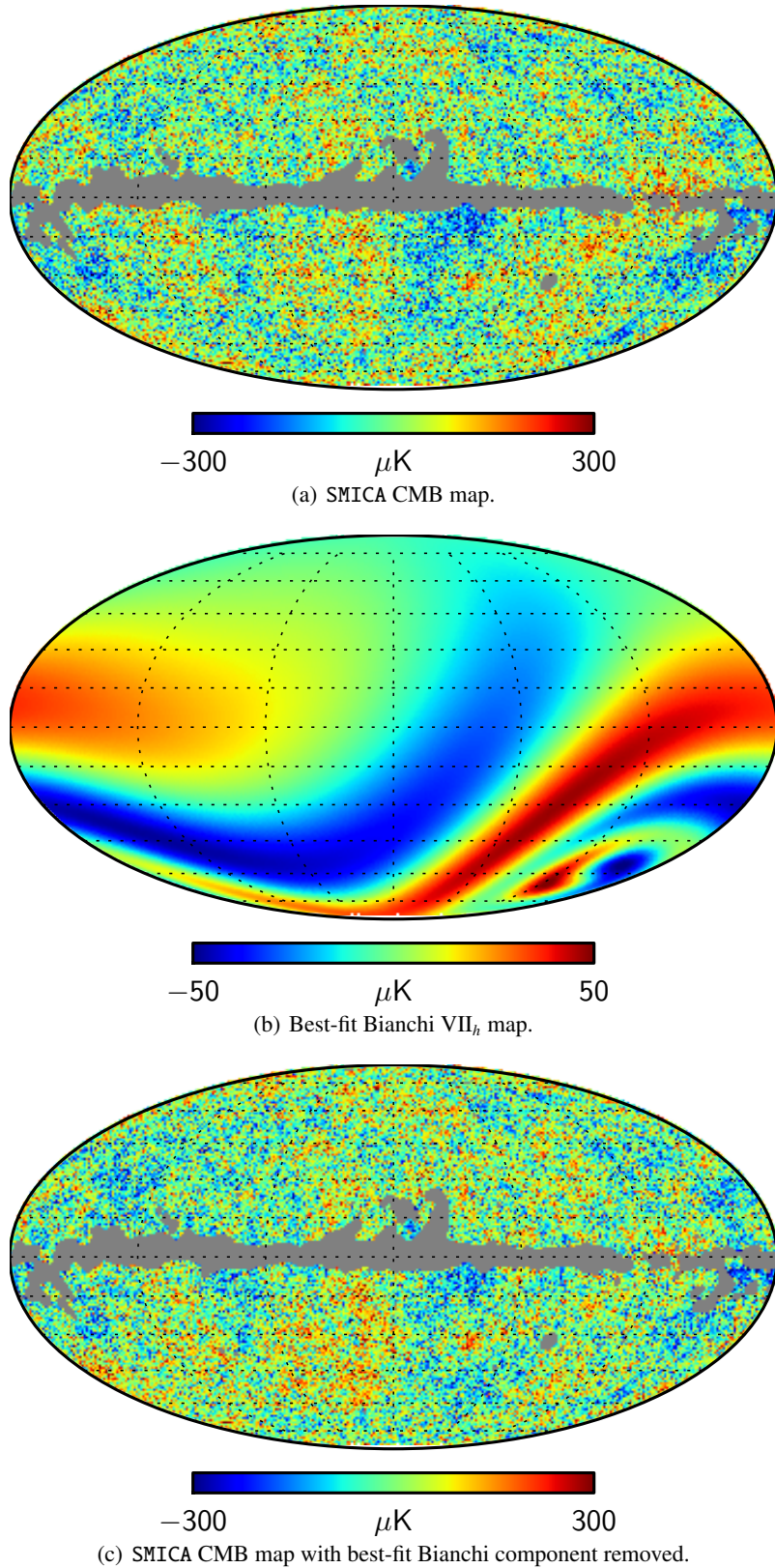


Fig. 20. Best-fit template of left-handed flat-decoupled-Bianchi VII_h model subtracted from *Planck* SMICA component-separated data. Before subtraction, the peak-to-peak variation is $\pm 594 \mu\text{K}$, reduced to $\pm 564 \mu\text{K}$ after subtraction.

(Luminet et al. 2003), is not sufficient for the detection of topology. Conversely, because our methods search directly for these correlations (and indeed marginalize over the amplitude of fluctuations), a slight modification of the background FRW

cosmology by lowering power in some or all multipoles (Planck Collaboration XV 2014) will not affect the ability to detect the correlations induced by such topologies.

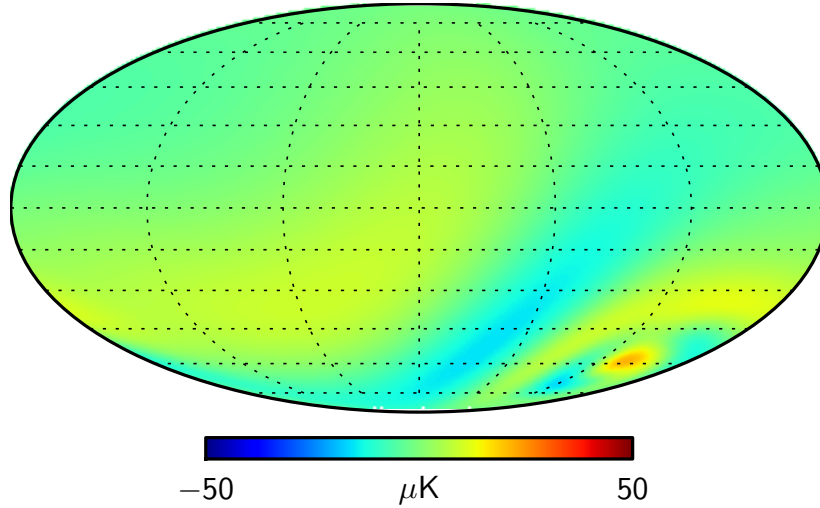


Fig. 21. Difference between best-fit template of flat-decoupled-Bianchi VII_h model recovered from WMAP 9-year data and from *Planck* SMICA component-separated data.

Table 4. Parameters recovered for left-handed flat-decoupled-Bianchi model.

Bianchi parameter	SMICA		SEVEM	
	MAP	Mean	MAP	Mean
Ω_m^B	0.38	0.32 ± 0.12	0.35	0.31 ± 0.15
Ω_Λ^B	0.20	0.31 ± 0.20	0.22	0.30 ± 0.20
x	0.63	0.67 ± 0.16	0.66	0.62 ± 0.23
$(\omega/H)_0$	8.8×10^{-10}	$(7.1 \pm 1.9) \times 10^{-10}$	9.4×10^{-10}	$(5.9 \pm 2.4) \times 10^{-10}$
α	$38^\circ 8$	$51^\circ 3 \pm 47^\circ 9$	$40^\circ 5$	$77^\circ 4 \pm 80^\circ 3$
β	$28^\circ 2$	$33^\circ 7 \pm 19^\circ 7$	$28^\circ 4$	$45^\circ 6 \pm 32^\circ 7$
γ	$309^\circ 2$	$292^\circ 2 \pm 51^\circ 9$	$317^\circ 0$	$271^\circ 5 \pm 80^\circ 7$

Notes. *Planck* data favour the inclusion of a Bianchi component in this phenomenological model.

Table 5. Parameters recovered for left-handed open-coupled-Bianchi model.

Bianchi parameter	SMICA		SEVEM	
	MAP	Mean	MAP	Mean
Ω_k	0.05	0.07 ± 0.05	0.09	0.08 ± 0.04
Ω_m^B	0.41	0.33 ± 0.07	0.41	0.32 ± 0.07
Ω_Λ^B	0.55	0.60 ± 0.07	0.50	0.59 ± 0.07
x	0.46	0.44 ± 0.24	0.38	0.39 ± 0.22
$(\omega/H)_0$	5.9×10^{-10}	$(4.0 \pm 2.4) \times 10^{-10}$	9.3×10^{-10}	$(4.5 \pm 2.8) \times 10^{-10}$
α	$57^\circ 4$	$122^\circ 5 \pm 960^\circ 0$	$264^\circ 1$	$188^\circ 6 \pm 98^\circ 7$
β	$54^\circ 1$	$70^\circ 8 \pm 35^\circ 5$	$79^\circ 6$	$81^\circ 1 \pm 31^\circ 7$
γ	$202^\circ 6$	$193^\circ 5 \pm 77^\circ 4$	$90^\circ 6$	$160^\circ 4 \pm 91^\circ 1$

Notes. *Planck* data do not favour the inclusion of a Bianchi component in this model and some parameters are not well constrained.

Similarly, using a Bayesian analysis we find no evidence for a physical, anisotropic Bianchi VII_h universe. However, *Planck* data do provide evidence supporting a phenomenological Bianchi VII_h component, where the parameters of the Bianchi component are decoupled from standard cosmology. The resulting best-fit Bianchi VII_h template found in *Planck* data is similar to that found in WMAP data previously (Jaffe et al. 2005; McEwen et al. 2013). However, although this Bianchi component can produce some of the (possibly anisotropic) temperature patterns seen on the largest angular scales (see also Planck Collaboration XXIII 2014), there is no set of cosmological

parameters which can simultaneously produce these patterns and the observed anisotropies on other scales. Moreover, the parameters of the best-fit Bianchi VII_h template in the decoupled setting are in strong disagreement with other measurements of the cosmological parameters.

These results are expected from previous measurements from COBE and WMAP, but *Planck*'s higher sensitivity and lower level of foreground contamination provides further confirmation. We have shown that the results are insensitive to the details of the preparation of the temperature maps (in particular, the method by which the cosmological signal is separated

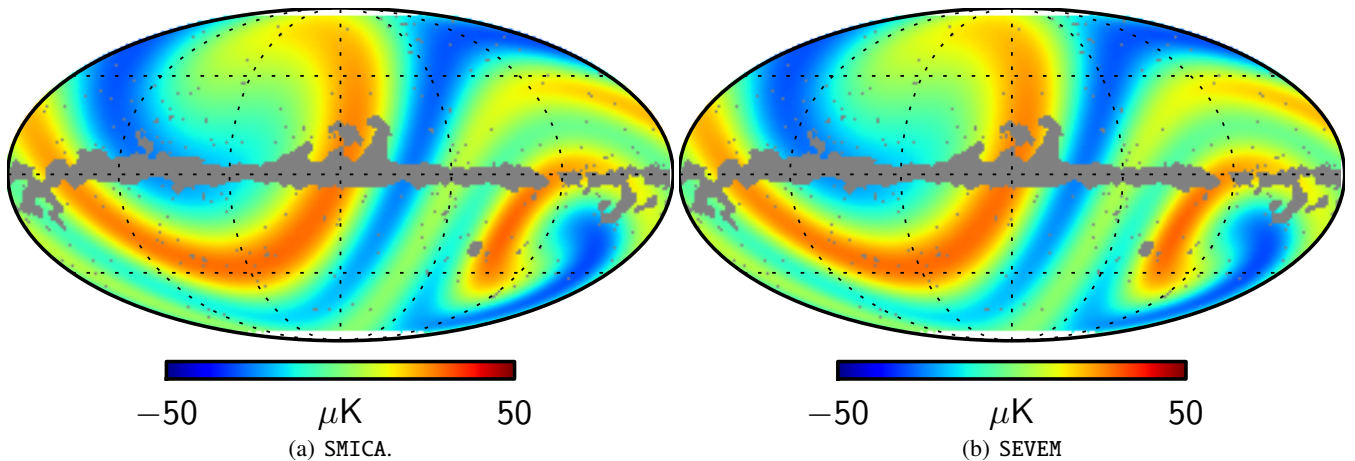


Fig. 22. Best-fit templates of left-handed open-coupled-Bianchi VII_h model recovered from *Planck* SMICA and SEVEM component-separated data. The Bayes factors for this model indicate that *Planck* data do not favour the inclusion of these Bianchi maps.

from astrophysical foreground contamination). Future *Planck* measurement of CMB polarization will allow us to further test models of anisotropic geometries and non-trivial topologies and may provide more definitive conclusions, for example allowing us to moderately extend the sensitivity to large-scale topology (Bielewicz et al. 2012).

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