Cosmological Image Processing

Jason McEwen www.jasonmcewen.org @jasonmcewen

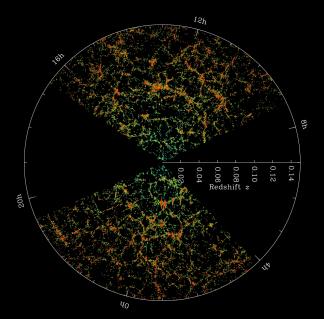
Mullard Space Science Laboratory (MSSL) University College London (UCL)



Image and Vision Computing New Zealand 2013



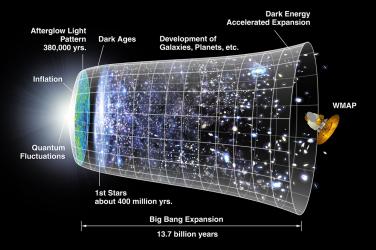








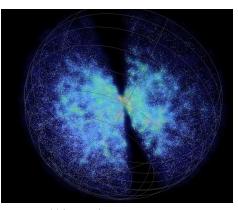
We have entered an era of concordance cosmology.



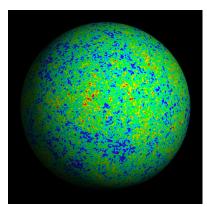




Cosmological observations



(a) Large-scale structure [Credit: SDSS]



(b) Cosmic microwave background [Credit: WMAP]

Figure: Cosmological observations





Cosmic microwave background (CMB)

Origin of CMB

- Temperature of early Universe sufficiently hot that photons had enough energy to ionise hydrogen.
- Universe opaque photon-baryon fluid.
- As Universe expanded it cooled, until photons no longer had sufficient energy to ionise hydrogen.
- Opening Photons decoupled from baryons and the Universe became transparent to radiation.

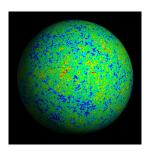


Figure: CMB

- CMB is highly uniform over the celestial sphere, however it contains small fluctuations



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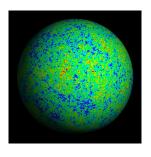


Figure: CMB

- Recombination occurred when temperature of Universe dropped to 3000K, about 400,000 years after the Big Bang.
- Photons then free to propagate largely unhindered and observed today on celestial sphere as CMB radiation.
- CMB is highly uniform over the celestial sphere, however it contains small fluctuations at a relative level of 10^{-5} due to acoustic oscillations in the early Universe.



Telescopes and satellites



Figure: LSS observations



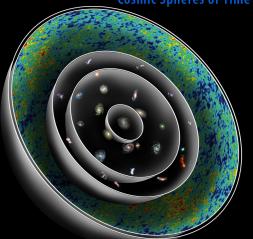
Figure: Full-sky CMB observations





Observations made on the celestial sphere

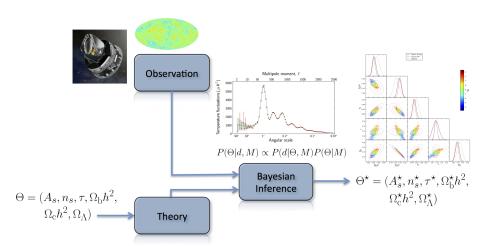
Cosmic Spheres of Time



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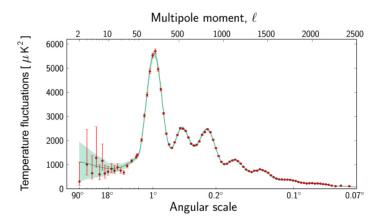
Precision cosmology Case study: CMB







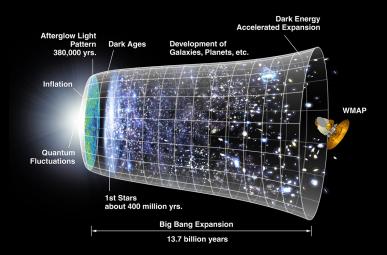
Precision cosmology Case study: CMB







Outstanding questions







Outline

- Cosmology
 - Cosmological concordance
 - Observational probes
 - Precision cosmology
 - Outstanding questions
- Dark energy
 - ISW effect
 - Continuous wavelets on the sphere
 - Detecting dark energy
- Cosmic strings
 - String physics
 - Scale-discretised wavelets on the sphere
 - String estimation
- Anisotropic cosmologies
 - Bianchi models
 - Bayesian analysis of anisotropic cosmologies
 - Planck results





- Universe consists of ordinary baryonic matter, cold dark matter and dark energy.
- Dark energy represents energy density of empty space, which acts as a repulsive force.
- Strong evidence for dark energy exists but we know very little about its nature and origin.
- A consistent model in the framework of particle physics lacking.

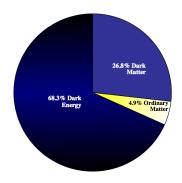


Figure: Content of the Universe [Credit: Planck]





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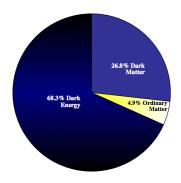


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Integrated Sachs Wolfe Effect Analogy

(no dark energy)

(with dark energy)

(a) No dark energy

(b) With dark energy

Figure: Analogy of ISW effect





Integrated Sachs Wolfe Effect Correlation between CMB and LSS

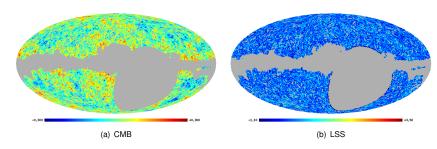


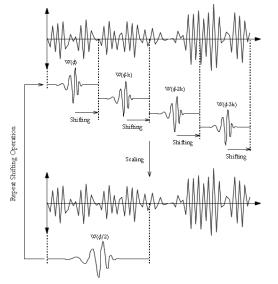
Figure: Constraining dark energy through any correlation between the CMB and LSS.





Cosmology Dark Energy Cosmic Strings Anisotropy ISW Effect Continuous Wavelets Detection

Recall wavelet transform in Euclidean space







- One of the first natural wavelet construction on the sphere was derived in the seminal work of Antoine and Vandergheynst (1998) (reintroduced by Wiaux 2005).
- Construct wavelet atoms from affine transformations (dilation, translation) on the sphere of a mother wavelet.
- The natural extension of translations to the sphere are rotations. Rotation of a function f on the sphere is defined by

$$\left[[\mathcal{R}(\rho)f](\omega) = f(\rho^{-1} \cdot \omega), \quad \omega = (\theta, \varphi) \in \mathbb{S}^2, \quad \rho = (\alpha, \beta, \gamma) \in \mathrm{SO}(3) \right]$$
 translation

- How define dilation on the sphere?
- The spherical dilation operator is defined through the conjugation of the Euclidean dilation and stereographic projection Π:

$$\mathcal{D}(a) \equiv \Pi^{-1} d(a) \Pi.$$





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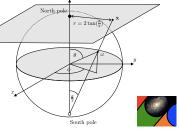
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Forward transform (i.e. analysis)

ullet Wavelet family on the sphere constructed from rotations and dilations of a mother wavelet Ψ :

$$\{\Psi_{a,\rho} \equiv \mathcal{R}(\rho)\mathcal{D}(a)\Psi : \rho \in SO(3), a \in \mathbb{R}_*^+\}.$$

The forward wavelet transform is given by

$$\boxed{ \begin{array}{c} W_{\Psi}^f(a,\rho) = \langle f, \Psi_{a,\rho} \rangle \\ \\ \text{projection} \end{array} } \equiv \int_{\mathbb{S}^2} \, \mathrm{d}\Omega(\omega) \, f(\omega) \, \Psi_{a,\rho}^*(\omega) \; ,$$

where $d\Omega(\omega) = \sin \theta d\theta d\varphi$ is the usual invariant measure on the sphere.

• Wavelet coefficients live in $SO(3) \times \mathbb{R}_*^+$; thus, directional structure is naturally incorporated.





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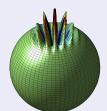


Continuous wavelets on the sphere Fast algorithms

- Fast algorithms essential (for a review see Wiaux, McEwen & Vielva 2007)
 - Factoring of rotations: McEwen et al. (2007), Wandelt & Gorski (2001), Risbo (1996)
 - Separation of variables: Wiaux et al. (2005)

FastCSWT code

http://www.fastcswt.org



Fast directional continuous spherical wavelet transform algorithms McEwen et al. (2007)

- Fortran
- Supports directional and steerable wavelets





Mother wavelets

- Correspondence principle between spherical and Euclidean wavelets (Wiaux et al. 2005).
- Mother wavelets on sphere constructed from the projection of mother Euclidean wavelets defined on the plane:

$$\Psi = \Pi^{-1} \Psi_{\mathbb{R}^2}$$

where $\Psi_{\mathbb{R}^2} \in L^2(\mathbb{R}^2, d^2x)$ is an admissible wavelet on the plane.





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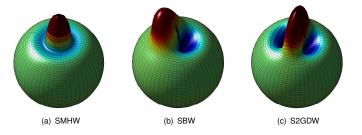


Figure: Spherical wavelets at scale a, b = 0.2.



Inverse transform (i.e. synthesis)

The inverse wavelet transform given by

where $d\rho(\rho) = \sin \beta d\alpha d\beta d\gamma$ is the invariant measure on the rotation group SO(3).

$$0 < \widehat{C}_{\Psi}^{\ell} \equiv \frac{8\pi^2}{2\ell + 1} \sum_{m=-\ell}^{\ell} \int_0^{\infty} \frac{\mathrm{d}a}{a^3} \mid (\Psi_a)_{\ell m} \mid^2 < \infty, \quad \forall \ell \in \mathbb{N}$$





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$$f(\omega) = \underbrace{\int_0^\infty \frac{\mathrm{d}a}{a^3} \, \int_{\mathrm{SO}(3)} \mathrm{d}\varrho(\rho)}_{\text{Sum' contributions}} \underbrace{W_\Psi^f(a,\rho) \, [\mathcal{R}(\rho) \widehat{L}_\Psi \Psi_a](\omega)}_{\text{weighted basis functions}},$$

where $d\varrho(\rho) = \sin \beta \, d\alpha \, d\beta \, d\gamma$ is the invariant measure on the rotation group SO(3).

Perfect reconstruction iff wavelets satisfy admissibility property:

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Wavelet coefficient correlation

- Compute wavelet correlation of CMB and LSS data (McEwen et al. 2007, McEwen et al. 2008).
- Compare to 1000 Monte Carlo simulations.





Detecting dark energy Wavelet coefficient correlation

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- Compare to 1000 Monte Carlo simulations.
- Correlation detected at 99.9% significance.
 - Independent evidence for the existence of dark energy!

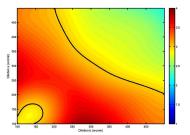


Figure: Wavelet correlation N_{σ} surface. Contours are shown at 3σ .





Constraining cosmological models

- Use positive detection of the ISW effect to constrain parameters of cosmological models:
 - Energy density Ω_{Λ} .
 - Equation of state parameter w relating pressure and density of cosmological fluid modelling dark energy, i.e. $p = w\rho$.
- Parameter estimates of $\Omega_{\Lambda} = 0.63^{+0.18}_{-0.17}$ and $w = -0.77^{+0.35}_{-0.36}$

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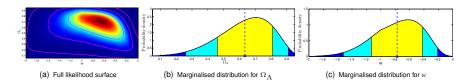


Figure: Likelihood for dark energy parameters.



Cosmology Dark Energy Cosmic Strings Anisotropy Strings Discrete Wavelets Estimation

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- Cosmology
 - Cosmological concordance
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Cosmic strings

- Symmetry breaking phase transitions in the early Universe → topological defects.
- Cosmic strings well-motivated phenomenon that arise when axial or cylindrical symmetry is broken
 → line-like discontinuities in the fabric of the Universe.
- We have not yet observed cosmic strings but we have observed string-like topological defects in other media

The detection of cosmic strings would open a new window into the physics of the Universe!





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Figure: Optical microscope photograph of a thin film of freely suspended nematic liquid crystal after a temperature quench. [Credit: Chuang et al. (1991).]

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Observational signatures of cosmic strings

Conical Spacetime

- Spacetime about a cosmic string is conical, with a three-dimensional wedge removed (Vilenkin 1981).
- Strings moving transverse to the line of sight induce line-like discontinuities in the CMB (Kaiser & Stebbins 1984).
- The amplitude of the induced contribution scales with the string tension $G\mu$.

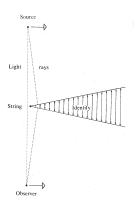


Figure: Spacetime around a cosmic string. [Credit: Kaiser & Stebbins 1984, DAMTP.]





Observational signatures of cosmic strings

CMB contribution

- Make contact between theory and data using high-resolution simulations.
- Search for a weak string signal s embedded in the CMB c, with observations d given by

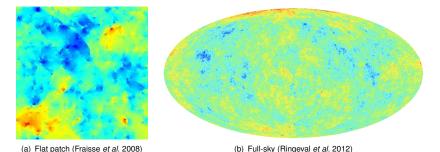


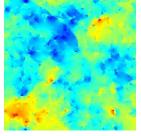
Figure: Cosmic string simulations.



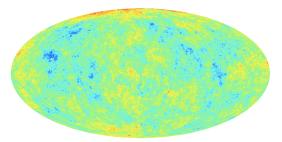
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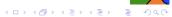


(a) Flat patch (Fraisse et al. 2008)



(b) Full-sky (Ringeval et al. 2012)

Figure: Cosmic string simulations.



Scale-discretised wavelets on the sphere

Wavelet construction

Exact reconstruction not feasible in practice with continuous wavelets!

- Exact reconstruction with directional wavelets on the sphere
 Wiaux, McEwen, Vandergheynst, Blanc (2008)
- Dilation performed in harmonic space [cf. McEwen et al. (2006), Sanz et al. (2006)]
 - Scale-discretised wavelet $\Psi^j \in L^2(\mathbb{S}^2, d\Omega)$ defined in harmonic space:

$$\Psi^{j}_{\ell m} \equiv \kappa^{j}(\ell) s_{\ell m} .$$

$$\frac{|\Phi_{\ell 0}|^2}{|\Phi_{\ell 0}|^2} + \sum_{j=0}^J \sum_{m=-\ell}^\ell \frac{|\Psi^j_{\ell m}|^2}{\text{wavelet}} = 1 \;, \quad \forall \ell \;.$$



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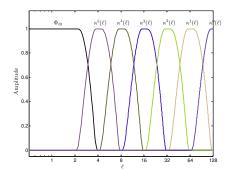


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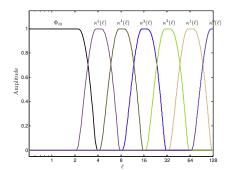


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$$\frac{|\Phi_{\ell 0}|^2}{\text{scaling function}} + \sum_{j=0}^J \sum_{m=-\ell}^\ell \frac{|\Psi_{\ell m}^j|^2}{\text{wavelet}} = 1 \;, \quad \forall \ell \;.$$



Scale-discretised wavelets on the sphere Wavelets

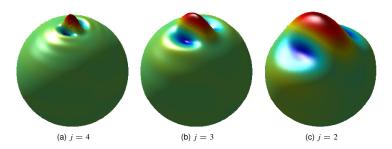


Figure: Scale-discretised wavelets on the sphere.



Scale-discretised wavelets on the sphere

Forward and inverse transform (i.e. analysis and synthesis)

The scale-discretised wavelet transform is given by the usual projection onto each wavelet:

$$\boxed{ \begin{array}{c} \textcolor{red}{W^{\Psi^j}(\rho) = \langle f, \, \mathcal{R}_{\rho} \Psi^j \rangle} \\ \text{projection} \end{array}} = \int_{\mathbb{S}^2} \, \mathrm{d}\Omega(\omega) f(\omega) (\mathcal{R}_{\rho} \Psi^j)^*(\omega) \; .$$

$$f(\omega) = \boxed{2\pi \int_{\mathbb{S}^2} \mathrm{d}\Omega(\omega') W^{\Phi}(\omega') (\mathcal{R}_{\omega'} L^{\mathrm{d}} \Phi)(\omega)} + \sum_{j=0}^{\prime} \boxed{\int_{\mathrm{SO}(3)} \mathrm{d}\varrho(\rho) W^{\Psi^j}(\rho) (\mathcal{R}_{\rho} L^{\mathrm{d}} \Psi^j)(\omega) \ .}$$





Scale-discretised wavelets on the sphere Forward and inverse transform (i.e. analysis and synthesis)

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The original function may be recovered exactly in practice from the wavelet (and scaling) coefficients:

$$f(\omega) = \boxed{2\pi \int_{\mathbb{S}^2} \mathrm{d}\Omega(\omega') W^\Phi(\omega') (\mathcal{R}_{\omega'} L^\mathrm{d}\Phi)(\omega)} + \boxed{\sum_{j=0}^J \int_{\mathrm{SO}(3)} \mathrm{d}\varrho(\rho) W^{\Psi^j}(\rho) (\mathcal{R}_\rho L^\mathrm{d}\Psi^j)(\omega) \ .}$$
 scaling function contribution

finite sum





Scale-discretised wavelets on the sphere

Exact and efficient computation

Wavelet analysis can be posed as an inverse Wigner transform on SO(3):

$$\boxed{ W^{\Psi^j}(\rho) = \sum_{\ell=0}^{L-1} \sum_{m=-\ell}^{\ell} \sum_{n=-\ell}^{\ell} \frac{2\ell+1}{8\pi^2} \left(W^{\Psi^j} \right)_{mn}^{\ell} D_{mn}^{\ell*}(\rho) \,, } \quad \text{where} \quad \left(W^{\Psi^j} \right)_{mn}^{\ell} = \frac{8\pi^2}{2\ell+1} f_{\ell m} \Psi_{\ell n}^{j*} \,. }$$

which can be computed efficiently via a factoring of rotations (Risbo 1996, Wandelt & Gorski 2001, McEwen et al. 2007).

Wavelet synthesis can be posed as an forward Wigner transform on SO(3):

$$f(\omega) \sim \sum_{j=0}^J \int_{\mathrm{SO}(3)} \, \mathrm{d}\varrho(\rho) W^{\Psi^j}(\rho) (\mathcal{R}_\rho L^{\mathrm{d}} \Psi^j)(\omega) = \sum_{j=0}^J \sum_{\ell mn} \frac{2\ell+1}{8\pi^2} \left(W^{\Psi^j}\right)_{mn}^\ell \Psi^j_{\ell n} Y_{\ell m}(\omega) \,,$$

where

$$\left(\boldsymbol{W}^{\Psi^{j}}\right)_{mn}^{\ell} = \langle \boldsymbol{W}^{\Psi^{j}}, \, D_{mn}^{\ell*} \rangle = \int_{\mathrm{SO}(3)} \, \mathrm{d}\varrho(\rho) \boldsymbol{W}^{\Psi^{j}}(\rho) D_{mn}^{\ell}(\rho) \,,$$

which can be computed efficiently via a factoring of rotations (Risbo 1996, Wiaux, McEwen *et al.* 20 and exactly by employing the Driscoll & Healy (1994) or McEwen & Wiaux (2011) sampling theorem.





Scale-discretised wavelets on the sphere

Exact and efficient computation

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Scale-discretised wavelets on the sphere

Exact and efficient computation

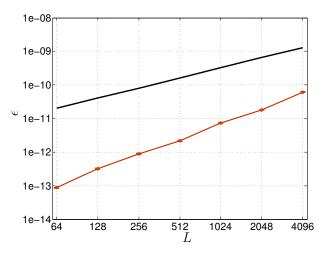


Figure: Numerical accuracy.





Scale-discretised wavelets on the sphere

Exact and efficient computation

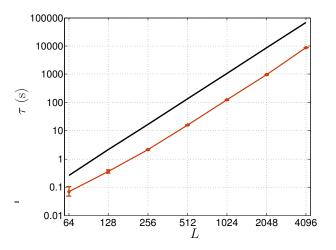


Figure: Computation time.





Scale-discretised wavelets on the sphere Codes

S2DW code

http://www.s2dw.org

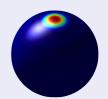


Exact reconstruction with directional wavelets on the sphere Wiaux, McEwen, Vandergheynst, Blanc (2008)

- Fortran
- Parallelised
- Supports directional and steerable wavelets

S2LET code

http://www.s2let.org



S2LET: A code to perform fast wavelet analysis on the sphere Leistedt, McEwen, Vandergheynst, Wiaux (2012)

- C. Matlab, IDL, Java
- Supports only axisymmetric wavelets at present
- Future extensions planned (directional and steerable wavelets, faster algos, spin wavelets)

Scale-discretised wavelets on the sphere

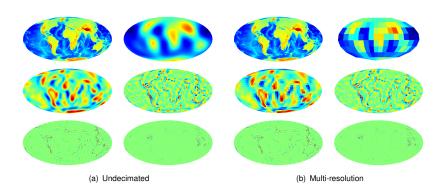


Figure: Scale-discretised wavelet transform of a topography map of the Earth.



Motivation for using wavelets to detect cosmic strings

Denote the wavelet coefficients of the data d by

$$W_{j\rho}^d = \langle d, \Psi_{j\rho} \rangle$$

for scale $j \in \mathbb{Z}^+$ and position $\rho \in SO(3)$.

ullet Consider an even azimuthal band-limit N=4 to yield wavelet with odd azimuthal symmetry.

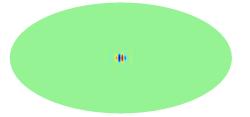


Figure: Example wavelet matched to the expected string contribution.





Motivation for using wavelets to detect cosmic strings

ullet Wavelet transform yields a sparse representation of the string signal o hope to effectively separate the CMB and string signal in wavelet space.

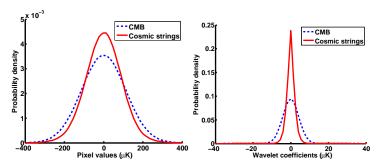


Figure: Distribution of CMB and string signal in pixel (left) and wavelet space (right).





Learning the statistics of the CMB and string signals in wavelet space

Wavelet-Bayesian approach to estimate the string tension and map:

- Need to determine statistical description of the CMB and string signals in wavelet space.
- Calculate analytically the probability distribution of the CMB in wavelet space.
- Fit a generalised Gaussian distribution (GGD) for the wavelet coefficients of a string training map (cf. Wiaux et al. 2009):

$$P_j^s(W_{j\rho}^s \mid G\mu) = \frac{\upsilon_j}{2G\mu\nu_i\Gamma(\upsilon_i^{-1})} e^{\left(-\left|\frac{W_{j\rho}^s}{G\mu\nu_j}\right|^{\upsilon_j}\right)},$$

with scale parameter ν_i and shape parameter v_i .





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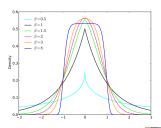
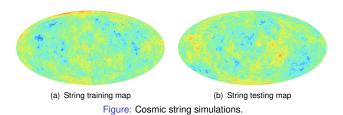


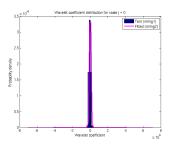
Figure: GGD

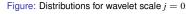


Learning the statistics of the CMB and string signals in wavelet space



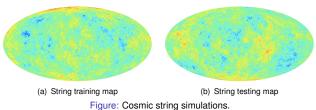
Distributions in close agreement.







Learning the statistics of the CMB and string signals in wavelet space



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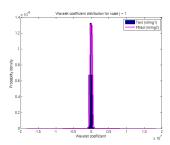
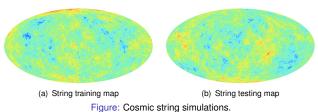


Figure: Distributions for wavelet scale j=1



Learning the statistics of the CMB and string signals in wavelet space



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Distributions in close agreement.

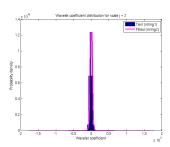


Figure: Distributions for wavelet scale j = 2.



Learning the statistics of the CMB and string signals in wavelet space

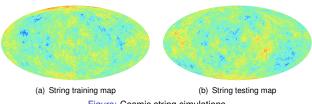


Figure: Cosmic string simulations.

Distributions in close agreement.

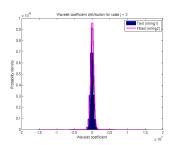
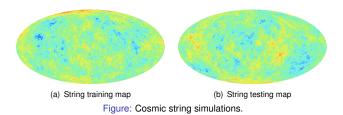
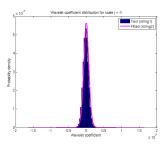


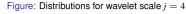
Figure: Distributions for wavelet scale j = 3.

Learning the statistics of the CMB and string signals in wavelet space



- Distributions in close agreement.
- Accurately characterised statistics of string signals in wavelet space.







- Perform Bayesian string tension estimation in wavelet space.
- For each wavelet coefficient the likelihood is given by

$$\mathbb{P}(W^d_{j\rho} \,|\: G\mu) = \mathbb{P}(W^s_{j\rho} + W^c_{j\rho} \,|\: G\mu) = \int_{\mathbb{R}} \,\mathrm{d}W^s_{j\rho} \,\, \mathbb{P}^c_j(W^d_{j\rho} - W^s_{j\rho}) \,\, \mathbb{P}^s_j(W^s_{j\rho} \,|\: G\mu) \;.$$

The overall likelihood of the data is given by

$$P(W^d \mid G\mu) = \prod_{j,\rho} P(W^d_{j\rho} \mid G\mu) ,$$

where we have assumed independence for numerical tractability.





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• Compute the string tension posterior $P(G\mu \mid W^d)$ by Bayes theorem:

$$P(G\mu \mid W^d) = \frac{P(W^d \mid G\mu) P(G\mu)}{P(W^d)} \propto P(W^d \mid G\mu) P(G\mu) .$$

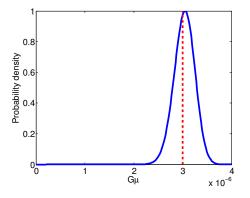


Figure: Posterior distribution of the string tension (true $G\mu = 3 \times 10^{-6}$).





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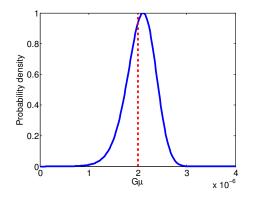


Figure: Posterior distribution of the string tension (true $G\mu = 2 \times 10^{-6}$).





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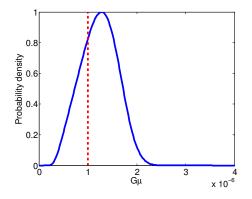


Figure: Posterior distribution of the string tension (true $G\mu = 1 \times 10^{-6}$).





Bayesian evidence for strings

- Compute Bayesian evidences to compare the string model M^s to the alternative model M^c
 that the observed data is comprised of just a CMB contribution.
- The Bayesian evidence of the string model is given by

$$E^s = \mathrm{P}(W^d \,|\, \mathrm{M}^s) = \int_{\mathbb{R}} \, \mathrm{d}(G\mu) \, \mathrm{P}(W^d \,|\, G\mu) \, \mathrm{P}(G\mu) \;.$$

The Bayesian evidence of the CMB model is given by

$$E^c = P(W^d \mid M^c) = \prod_{j,\rho} P_j^c(W_{j\rho}^d).$$

Compute the Bayes factor to determine the preferred model:

$$\Delta \ln E = \ln(E^s/E^c) \ .$$





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Table: Tension estimates and log-evidence differences for simulations.

$G\mu/10^{-6}$	0.7	0.8	0.9	1.0	2.0	3.0
$\widehat{G\mu}/10^{-6}$	1.1	1.2	1.2	1.3	2.1	3.1
$\Delta {\rm ln} E$	-1.3	-1.1	-0.9	-0.7	5.5	29





Recovering string maps

- Inference of the wavelet coefficients of the underlying string map encoded in posterior probability distribution $P(W^s_{j\rho} \mid W^d)$.
- Estimate the wavelet coefficients of the string map from the mean of the posterior distribution:

$$\overline{W}_{j\rho}^{s} = \int_{\mathbb{R}} dW_{j\rho}^{s} \ W_{j\rho}^{s} \ P(W_{j\rho}^{s} \mid W^{d})$$

- Recover the string map from its wavelets (possible since the scale-discretised wavelet transform on the sphere supports exact reconstruction).
- Work in progress...





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Outline

- Cosmology
 - Cosmological concordance
 - Observational probes
 - Precision cosmology
 - Outstanding questions
- Dark energy
 - ISW effect
 - Continuous wavelets on the sphere
 - Detecting dark energy
- Cosmic strings
 - String physics
 - Scale-discretised wavelets on the sphere
 - String estimation
- Anisotropic cosmologies
 - Bianchi models
 - Bayesian analysis of anisotropic cosmologies
 - Planck results





Bianchi VII_h cosmologies

Test fundamental assumptions on which modern cosmology is based, e.g. isotropy.

- Relax assumptions about the global structure of spacetime by allowing anisotropy about each point in the universe, i.e. rotation and shear.
- ullet Yields more general solutions to Einstein's field equations o Bianchi cosmologies
- Induces a characteristic subdominant, deterministic signature in the CMB, which is embedded in the usual stochastic anisotropies (Collins & Hawking 1973, Barrow et al. 1985).





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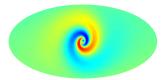


Figure: Bianchi CMB contribution.





Bianchi VII_h cosmologies

Parameters

• Models described by the parameter vector:

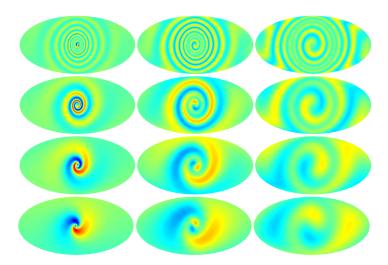
$$\Theta_{\mathrm{B}} = (\Omega_{\mathrm{m}}, \, \Omega_{\Lambda}, \, x, \, (\omega/H)_{0}, \, \alpha, \beta, \gamma) \; .$$

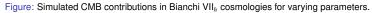
- Free parameter, x, describing the comoving length-scale over which the principal axes of shear and rotation change orientation, i.e. 'spiralness'.
- Amplitude characterised by the dimensionless vorticity $(\omega/H)_0$, which influences the amplitude of the induced temperature contribution only and not its morphology.
- The orientation and handedness of the coordinate system is also free.





Bianchi VII_h cosmologies **Simulations**







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Bayesian analysis of Bianchi VII_h cosmologies

Parameter estimation

- Perform Bayesian analysis of McEwen et al. (2013).
- Consider open and flat cosmologies with cosmological parameters:

$$\Theta_{\rm C}=(A_s,\,n_s,\,\tau,\,\Omega_{\rm b}h^2,\,\Omega_{\rm c}h^2,\,\Omega_{\Lambda},\,\Omega_k).$$

Recall Bianchi parameters:

$$\Theta_{\rm B} = (\Omega_{\rm m}, \, \Omega_{\Lambda}, \, x, \, (\omega/H)_0, \, \alpha, \beta, \gamma).$$

Likelihood given by

$$| P(\mathbf{d} \mid \Theta_{\mathrm{B}}, \Theta_{\mathrm{C}}) \propto \frac{1}{\sqrt{|\mathbf{X}(\Theta_{\mathrm{C}})|}} e^{\left[-\chi^2(\Theta_{\mathrm{C}}, \Theta_{\mathrm{B}})/2\right]} ,$$

$$\chi^2(\Theta_{\rm C},\Theta_{\rm B}) = \left[\mathbf{d} - \mathbf{b}(\Theta_{\rm B}) \right]^{\dagger} \mathbf{X}^{-1}(\Theta_{\rm C}) \left[\mathbf{d} - \mathbf{b}(\Theta_{\rm B}) \right].$$





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Bayesian analysis of Bianchi VII_h cosmologies

Covariance

- Bianchi VII_h templates can be computed accurately and rotated efficiently in harmonic space \rightarrow consider harmonic space representation, where $d = \{d_{\ell m}\}$ and $b(\Theta_{\rm B}) = \{b_{\ell m}(\Theta_{\rm B})\}$.
- Partial-sky analysis that handles in harmonic space a mask applied in pixel space.
- Add masking noise in order to marginalise the pixel values of the data contained in the masked region, with variance for pixel i given by $\sigma_m^2(\omega_i)$.
- The covariance is then given by

$$\mathbf{X}(\Theta_{\mathbf{C}}) = \mathbf{C}(\Theta_{\mathbf{C}}) + \mathbf{M}$$
,

- $C(\Theta_C)$ is the diagonal CMB covariance defined by the power spectrum $C_{\ell}(\Theta_C)$;
- M is the non-diagonal noisy mask covariance matrix defined by

$$\mathbf{M}_{\ell m}^{\ell'm'} = \langle m_{\ell m} \, m_{\ell'm'}^* \rangle \simeq \sum_{\omega_i} \sigma_m^2(\omega_i) Y_{\ell m}^*(\omega_i) \, Y_{\ell'm'}(\omega_i) \, \Omega_{\text{pix}}^2 \,.$$





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- Bianchi VII_h templates can be computed accurately and rotated efficiently in harmonic space \rightarrow consider harmonic space representation, where $d = \{d_{\ell m}\}$ and $b(\Theta_B) = \{b_{\ell m}(\Theta_B)\}$.
- Partial-sky analysis that handles in harmonic space a mask applied in pixel space.
- Add masking noise in order to marginalise the pixel values of the data contained in the masked region, with variance for pixel i given by $\sigma_m^2(\omega_i)$.
- The covariance is then given by

$$\mathbf{X}(\Theta_{\mathbf{C}}) = \mathbf{C}(\Theta_{\mathbf{C}}) + \mathbf{M}$$

- $\mathbb{C}(\Theta_{\mathbb{C}})$ is the diagonal CMB covariance defined by the power spectrum $C_{\ell}(\Theta_{\mathbb{C}})$;
- M is the non-diagonal noisy mask covariance matrix defined by

$$\mathbf{M}_{\ell m}^{\ell' m'} = \langle m_{\ell m} m_{\ell' m'}^* \rangle \simeq \sum_{\omega_i} \sigma_m^2(\omega_i) Y_{\ell m}^*(\omega_i) Y_{\ell' m'}(\omega_i) \Omega_{\mathrm{pix}}^2.$$





Bayesian analysis of Bianchi VII_h cosmologies

Model selection

Compute the Bayesian evidence to determine preferred model:

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

- Use MultiNest to compute the posteriors and evidences via nested sampling (Feroz & Hobson 2008, Feroz et al. 2009).
- Consider two models
 - Flat-decoupled-Bianchi model: Θ_C and Θ_B fitted simultaneously but decoupled \to phenomenological
 - Open-coupled-Bianchi model: Θ_C and Θ_B fitted simultaneously and coupled \to physical





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Bayesian analysis of Bianchi VII_h cosmologies

Validation with simulations

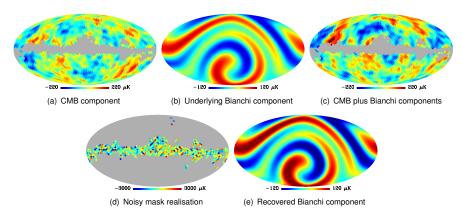


Figure: Partial-sky simulation with embedded Bianchi VII_h component at L=32.



Bayesian analysis of Bianchi VII_h cosmologies

Validation with simulations

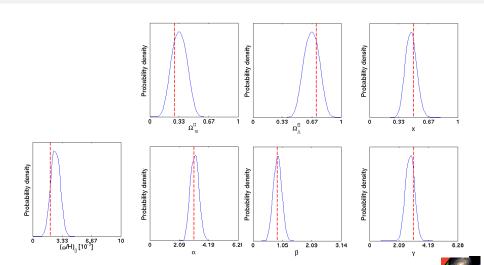
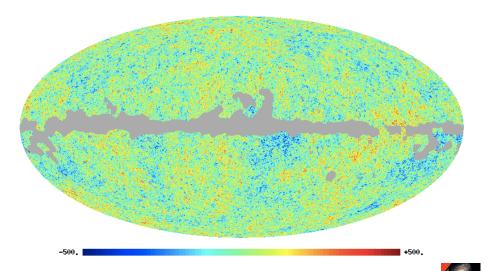


Figure: Marginalised posterior distributions recovered from partial-sky simulation at L=32.



Planck results







Planck results: flat-decoupled-Bianchi model

Bayesian evidence

Table: Bayes factor relative to equivalent ΛCDM model (positive favours Bianchi model).

Model	$\Delta { m ln} E$		
	SMICA	SEVEM	
Flat-decoupled-Bianchi (left-handed) Flat-decoupled-Bianchi (right-handed)	2.8 ± 0.1 0.5 ± 0.1	1.5 ± 0.1 0.5 ± 0.1	
1 (0 /			

- On the Jeffreys (1961) scale, evidence for the inclusion of a Bianchi VII_h component would be termed strong (significant) for SMICA (SEVEM) component-separated data.
- A log-Bayes factor of 2.8 corresponds to an odds ratio of approximately 1 in 16.

Planck data favour the inclusion of a phenomenological Bianchi VII_h component!





Planck results: flat-decoupled-Bianchi model

Best-fit Bianchi component

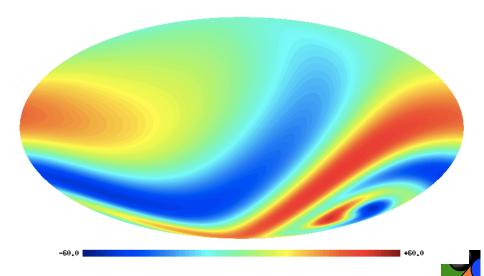


Figure: Best-fit template of flat-decoupled-Bianchi VII_n model found in Planck SMICA component-separated data.

BUT the flat-Bianchi-decoupled model is phenomenological and **not physical!**

Parameter estimates are not consistent with concordance cosmology.





Planck results: open-coupled-Bianchi model

Bayesian evidence

Table: Bayes factor relative to equivalent Λ CDM model (positive favours Bianchi model).

Model	$\Delta { m ln} E$		
	SMICA	SEVEM	
Open-coupled-Bianchi (left-handed) Open-coupled-Bianchi (right-handed)	$0.0 \pm 0.1 \\ -0.4 \pm 0.1$	0.0 ± 0.1 -0.4 ± 0.1	

 In the physical setting where the standard cosmological and Bianchi parameters are coupled,

Planck data do not favour the inclusion of a Bianchi VII_h component.

• Find no evidence for Bianchi VII_h cosmologies and constrain vorticity to:







Planck results: open-coupled-Bianchi model

Bayesian evidence

Table: Bayes factor relative to equivalent Λ CDM model (positive favours Bianchi model).

Model	$\Delta { m ln} E$		
	SMICA	SEVEM	
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 In the physical setting where the standard cosmological and Bianchi parameters are coupled,

Planck data do not favour the inclusion of a Bianchi VII_h component.

• Find no evidence for Bianchi VII_h cosmologies and constrain vorticity to:

$$(\omega/H)_0 < 8.1 \times 10^{-10}$$

95% confidence level



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