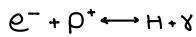


## Cosmology

### Very early universe:

electron, proton, photon plasma

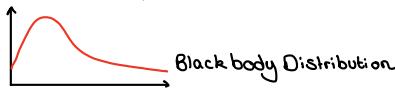


Cools under expansion.

5% Normal Matter

Can observe primordial plasma.

24% Dark matter



71% Dark Energy

### Cosmological Principle:

On Large Scales the Universe is homogeneous and isotropic

Homogeneous: The Universe has the same composition and structure everywhere when averaged over sufficiently large distances.

Isotropic: Means the Universe looks the same in every direction with no preferred orientation.

Therefore: treat our Universe as 'small' perturbations

on top of a completely smooth background.

### Statistical approach to fluctuations

#### Gaussian

initial amplitudes described by a power spectrum  $P(k)$

$R \rightarrow k$ : Spatial homogeneity and isotropy.

#### Mathematically:

Observed CMB: field on the Sphere  $T(l)$

Function drawn from a distribution with Kernel  $C(l, j)$

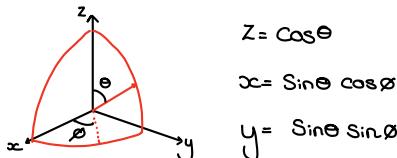
Constructed from 'angular power Spectrum'  $C_l$  of Cosmological model.

Cosmological parameters  $\approx$  Hyperparameters of the gaussian process.

### Power Spectrum



### Functions on a Sphere:



$$Z = \cos \theta$$

$$x = \sin \theta \cos \phi$$

$$y = \sin \theta \sin \phi$$

### Basic Functions

A set of functions that one can describe any other functions as a linear combination of them

But can be used for value on a sphere's surface

Take the form:

$$T(\underline{i}) = \sum_m a_{lm} Y_{lm}(\underline{i})$$

$\underline{i}$  - Vectors ( $\theta_i, \phi_i$ )

$a_{lm}$  - Coefficients

$$a_{lm} = \int d^2i T(\underline{i}) Y_{lm}^*(\underline{i})$$

$Y_{lm}$  - basis functions

Due to:

$$\text{Orthogonality: } \int d^2i Y_{lm}(\underline{i}) Y_{l'm'}^*(\underline{i}) = S_{ll'} S_{mm'}$$

$$\text{Completeness: } \sum_m Y_{lm}(\underline{i}) Y_{l'm}^*(\underline{i}) = \delta^{(2)}(\underline{i} - \underline{j})$$

$Y_{lm}$ 's are eigenfunctions of the partial differential equations on the surface of the sphere

Form of  $Y_{lm}$ :

$$Y_{lm}(\theta, \phi) \sim P_{lm}(\theta) e^{im\phi}$$

$P_{lm}$  - Associated Legendre Polynomials

$l = 0, 1, 2, \dots$

$m = -l, -l+1, \dots, l-1, l$  (2L+1 values)

The Addition Theorem:

$$\sum_m Y_{lm}(\underline{i}) Y_{lm}^*(\underline{j}) = \frac{(2l+1)}{4\pi} P_l(\underline{i} \cdot \underline{j}) \quad P_l(\cos(\theta)) \text{ is a Legendre Polynomial.}$$

The Kernel is Stationary:

Statistical properties only dependent on separations.

Statistically Isotropy for a Sphere

↳ No relations between points on the sphere should explicitly depend on the coordinates

But:

For two points: Any properties between two points can only depend on arc-length / angle between

For three points: three arc lengths between.

Two Point Function:

$$\langle T(\underline{i}) T(\underline{j}) \rangle = f(\underline{i}, \underline{j}) \quad \text{in harmonic space:}$$

angle between

$$\langle a_{lm} a_{l'm'}^* \rangle = C_l S_{ll'} \delta_{mm'}$$

For a Statistically Isotropic Gaussian Field on the Sphere:

↳ all of its properties are determined by the angular power spectrum  $C_l$ .

As Temperature is Real:

$$T^* = T$$

$$\sum_m a_{lm}^* Y_{lm}(\underline{i}) = \sum_m a_{lm} Y_{lm}(\underline{i})$$

As  $Y_{lm} = P_{lm}(\theta) e^{im\phi}$

But:  $Y_{lm}^*(\underline{i}) = (-1)^m Y_{l-m}(\underline{i})$

$$\sum_m a_{lm}^* (-1)^m Y_{l-m}(\underline{i}) = \sum_m a_{lm} Y_{lm}(\underline{i})$$

$$\sum_m a_{l-m}^* (-1)^m Y_{l-m}(\underline{i}) = \sum_m a_{lm} Y_{lm}(\underline{i})$$

Thus  $a_{l-m} = (-1)^m a_{lm}^*$

$m=0$ :

$$a_{l,0} = a_{l,0}^* \rightarrow a_{l,0} \text{ - real.}$$

$m > 0$ :

Sums are given in terms of  $m > 0$  which are complex.

2L

For a given L:

2L+1

Over all L:

$$\sum_{l=0}^{l_{\max}} 2l+1 = (l_{\max}+1)^2 \rightarrow \langle a_{lm} a_{l'm'}^* \rangle = C_l \delta_{ll'} \delta_{mm'}$$

$C_l$  - Set of variance of effectively  $2L+1$  quantities

Multipole moments distributed:

$2L+1$  real variables, i.e.  $y_i$ , with the same variance  $C$ .

Sampling distribution

$$\prod_i \frac{1}{2\pi\sqrt{C}} e^{-\frac{y_i^2}{2C}}$$

→ function of  $\sum y_i^2$

→  $\frac{1}{2L+1} \sum_i y_i^2$  Sufficient Statistic for learning about  $C$ :

Power Spectrum Estimator:

$$\tilde{C}_l = \frac{1}{2L+1} \sum_m |a_{lm}|^2$$

$$\langle \tilde{C}_l \rangle = C_l$$

$$\langle \langle \tilde{C}_l^2 \rangle \rangle = \langle \tilde{C}_l \rangle - \langle \tilde{C}_l \rangle^2$$

$$= \frac{2C_l^2}{2L+1} \quad \begin{array}{l} \text{(cosmic variance)} \\ \text{Irreducible} \end{array}$$

Visualisation:

$$l_{\max} \approx 2 n_{\text{side}}$$

$$\text{DoF: } (l_{\max}+1)^2 \text{ vs } N_{\text{pix}} = 12 n_{\text{side}}^2$$

Data: Planck Legacy Archive

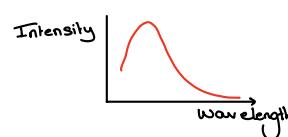
2018 'PR3' Legacy Archive

GHz:

Maps are made from different detectors looking at the microwave

Sky at different frequencies.

CMB: Almost perfect body



## Foreground Contamination:

### ↳ Galactic

Galactic Synchrotron emission ( $e^-$  in magnetic field) — Stronger low freq

Galactic Dust emission — Stronger high freq

### Extragalactic

EG Point Sources

Cosmic Infrared Background (CIB) ('continuous point sources')

Thermal and Kinetic Sunyaev Zeldovich effects.

## Methods for Handling This:

Remove (Subtraction)

Mask (Removal of regions)

Model (Dust maps + Syncrotron + bayesian inference)

## Correcting Power Spectrum to account for masking:

$$\langle \tilde{C}_L \rangle = M_{LL} C_L + N.$$

## Mask Deconvolved power Spectrum:

$$\hat{C}_L = M_{LL}^{-1} (\tilde{C}_L - N)$$

$$\text{Unbiased: } \langle \tilde{C}_L \rangle = C_L$$

$$\hat{C}_L \approx \frac{\tilde{C}_L - N}{f_{\text{sky}}} \quad f_{\text{sky}} - \text{Sky fraction: mean of the mask over the sky} \quad f_{\text{sky}} = \bar{W}^2$$

Can model the contribution of residual foregrounds at power spectrum level

$$C_L^{fg} = A L^\gamma \quad A, \gamma - \text{parameters}$$

Also must deal with:

### Beam Response:

Each detector sees light in a general direction convolved with telescope's beam response.

$W_L$  - the square of beam transfer function  $B_L$

$$\hat{C}_L \rightarrow \frac{\hat{C}_L}{W_L}$$

## Pixel Window function Correction:

↳ Finite Sized Pixels (tiny Smoothing Kernel) - especially at high multipoles  $L$ .

Can be corrected for using `healpy.pixwin`

## Binning Power Spectrum

Smooths cosmic variance from raw power spectrum

Averaging adjacent multipoles values improving visual clarity

### Implementation:

Straight average in  $D_L$  over an odd no. multipoles.

$\Delta L$  of 31 or 61 typical

## Summary

Foreground

Masking

Beam (Instrumental)

Pixel window functions

More in detail on masking:

Investigating the effect of masking on the Power Spectrum.

Theoretical Power Spectrum:  $C_L$

Expected  $\tilde{C}_L$  post masking:

$$\tilde{C}_L = \frac{1}{2L+1} \sum_m \tilde{\alpha}_{lm} \tilde{\alpha}_{lm}^*$$

Spherical harmonic transforms

$$\tilde{\alpha}_{lm} = \sum_i w_i T(i) Y_{lm}(i)$$

Temperature Any harmonic func  
Mask function (0-1)

$$\tilde{\alpha}_{lm}^* = \sum_j w_j T(j) Y_{lm}(j)$$

$$\tilde{C}_L = \frac{1}{2L+1} \sum_m \sum_i w_i T(i) Y_{lm}^*(i) \sum_j w_j T(j) Y_{lm}(j)$$

$$\langle \tilde{C}_L \rangle = \frac{1}{2L+1} \sum_m w_i w_j Y_{lm}^*(i) Y_{lm}(j) \underbrace{\langle T(i) T(j) \rangle}_{\sum_l C_L Y_{lm}(i) Y_{lm}^*(j)} \text{Theoretical (Signal + Noise)}$$

$$= \underbrace{\frac{1}{2L+1} \sum_{m \neq m'} w_i w_j Y_{lm}^*(i) Y_{lm}(j) Y_{l'm'}^*(i) Y_{l'm'}(j) C_{l'}}_{\text{Signal}} + \underbrace{\frac{1}{2L+1} \sum_{m \neq m'} w_i w_j Y_{lm}^*(i) Y_{lm}(j) N_{ij}}_{\text{Noise}}$$

$$= \frac{1}{2L+1} \sum_l \left( \sum_{m \neq m'} w_i w_j Y_{lm}^*(i) Y_{lm}(j) Y_{l'm'}^*(i) Y_{l'm'}(j) \right) C_l + \text{Noise term}$$

$$\langle \tilde{C}_L \rangle = \sum_l M_{ll} C_l + \langle \tilde{N}_l \rangle$$

' $M_{ll}$ ' - Matrix operation - Coupling matrix

↳ Expectation masked power spectrum - Matrix  $\times$  Theory Power Spectrum

Deconvolved Power Spectrum,  $\hat{C}_L$

$$\hat{C}_L = M_{ll}^{-1} (\tilde{C}_L - \langle \tilde{N}_l \rangle)$$

$$\begin{aligned} \langle \hat{C}_L \rangle &= M_{ll}^{-1} (\langle \tilde{C}_L \rangle - \langle \tilde{N}_l \rangle) \\ &= M_{ll}^{-1} (M_{ll} C_l + \langle \tilde{N}_l \rangle - \langle \tilde{N}_l \rangle) \end{aligned}$$

$$\langle C_L \rangle = C_L \quad \text{Unbiased estimator.}$$

Further Approximations

$$\hat{C}_L \approx \frac{\tilde{C}_L - \langle \tilde{N}_l \rangle}{f_{\text{sky}}}$$

Sky fraction  $f_{\text{sky}} = \overline{W^2}$  - mean of the square of the mask over the sky.

Variance of masked  $\tilde{C}_L$

↳ The mask couples modes so the covariance is no longer diagonal:

$$\langle \langle \tilde{C}_L \tilde{C}_{L'} \rangle \rangle \neq 0 \text{ for } L \neq L'$$

Each element of matrix is a multidimensional

quadratic function of the theory power spectrum - depending on all the  $C_L$ s

Can obtain  $\langle \langle \hat{C}_l \hat{C}_{l'} \rangle \rangle$  from  $\langle \langle \tilde{C}_l \tilde{C}_{l'} \rangle \rangle$  by matrix multiplying on one side with inverse of Coupling matrix and on other side with the transpose.

## Theory and Data Comparison

### Likelihood:

A pixel based likelihood: (Impossible due to Invertibility)

$$\frac{1}{\sqrt{12\pi(C_l+N_l)}} e^{\left(-\frac{1}{2} \left[ \begin{matrix} \tilde{C}_l \\ C_l+N_l \end{matrix} \right]^{-1} \left[ \begin{matrix} \tilde{C}_l \\ C_l+N_l \end{matrix} \right] \right)}$$

Pixel data (50 million)

Too large to invert

Covariance matrix + noise from model

Alternative: In an idealised Circumstance

↳ Angular power Spectrum is a Sufficient Statistic?

If the Instrument noise is gaussian and isotropic, all that changes is that  
 $C_l \rightarrow C_l + N_l$

$$P(\hat{C}_l | C_l) \propto \frac{1}{\sqrt{2\pi(C_l+N_l)^{2L+1}}} e^{-\frac{1}{2} \frac{(2L+1)\hat{C}_l}{C_l+N_l}}$$

In the real world case:

Gaussian likelihood: full power Spectra

$$\langle \hat{C}_l \rangle = C_l + N_l$$

$$\langle \langle \hat{C}_l \hat{C}_{l'} \rangle \rangle = \dots$$

$$\frac{1}{\sqrt{12\pi \langle \langle C_l C_{l'} \rangle \rangle}} e^{-\frac{1}{2} \left\{ (\tilde{C}_l - C_l - N_l) \langle \langle C_l C_{l'} \rangle \rangle (\tilde{C}_{l'} - C_{l'} - N_{l'})^T \right\}}$$

→ Every model must recompute

### Fudicial Gaussian

→ Scatter/Covariance matrix kept constant - does not adjust for each model adjustment

$$\langle \langle C_l C_{l'} \rangle \rangle_F$$

Advantage of 'unbiased':

$$\frac{1}{\sqrt{2\pi \langle \langle C_l C_{l'} \rangle \rangle_F}} e^{-\frac{1}{2} \left\{ (\tilde{C}_l - C_l - N_l) \langle \langle C_l C_{l'} \rangle \rangle_F (\tilde{C}_{l'} - C_{l'} - N_{l'})^T \right\}}$$

must better understand noise modelling:

↳ In the 'numerator' - we have  $\hat{C}_l - C_l - C_l^{1g} - N_l$

$N_l$  factor Subtracts the noise off the model

Not known perfectly → residual trend will be fitted by the Cosmology  
 'biasing' the inference.

### Cross Spectra:

↳ Rather than Power Spectrum of one map - Cross power Spectrum of two maps

If noise in the maps is independent → noise contribution to cross spectrum is on average zero.

$$\tilde{C}_l = \frac{1}{2L+1} \sum_m \tilde{a}_{lm} \tilde{a}_{lm}^* \longrightarrow \tilde{C}_l^{IJ} = \frac{1}{2L+1} \sum_m \tilde{a}_{lm}^I \tilde{a}_{lm}^{J*} \longrightarrow I, J \text{ two different maps}$$

Methods of Splitting by:

Frequency ( $143 \times 217$ )

Detector ( $143-d_{s1} \times 143-d_{s2}$ )

Time ( $143\text{hm1} \times 143\text{hm2}$ )

Result: Can make a likelihood out of multiple Spectra.

Hybrid Approach

Exact pixel base likelihood at low  $L$   $L=2-29$

Approx Power Spectrum based likelihood at high  $L$   $L=30-2500$ .

Gravitational Lensing: Statistical Challenges

Cosmology background:

Distances / Redshift:

The Universe is expanding:

$$z = \frac{\lambda_{\text{obs}} - \lambda_{\text{emitted}}}{\lambda_{\text{emitted}}} \quad (\text{redshift}) \quad \text{tells us how much the universe has expanded}$$

Since light was emitted before reaching us.

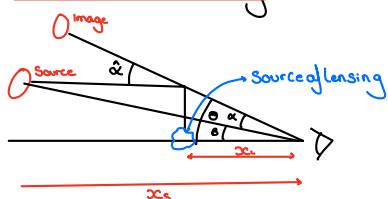
Comoving Distance,  $X(z)$  — denotes distance.

Standard Cosmological Model: (~6 free params)

$\Omega_m$ : Matter density     $\Omega_k$ : Dark Energy density     $A_s$ : Strength of initial perturbations

$\Omega_b$ : Baryon density     $H_0$ : Hubble Constant     $n_s$ : Tilt of initial power spectrum

Gravitational Lensing



$$\text{Simple Case: Deflection Angle } \hat{\alpha} = \frac{4GM}{c^2 b}$$

\* In a flat universe

$$\text{Change in angular position } \alpha = \frac{x_s - x_l}{x_s} \frac{4GM}{c^2 b}$$

Evidence for GR:

Deflection is twice that in Newtonian gravity

Generalised to total deflection from gradient of grav. potential along photon path

$$\alpha = \frac{2}{c^2} \int_0^{x_s} dx_i \frac{x_s - x_i}{x_s} \nabla_\perp \Phi(x_i, \theta, x_i)$$

$$= -\frac{2}{c^2} \nabla \Psi(\theta, x_i)$$

Image transformed according to lens equation.

$$\vec{\beta} = \vec{\theta} - \frac{2}{c^2} \nabla \psi$$

Source and Image Coordinates mapping

$$I(\vec{\theta}) = I^{(s)}[\vec{\beta}(\vec{\theta})] \quad \text{Coordinate transformation.}$$

$\vec{\beta}$  - unlensed / source plane coords

$\vec{\theta}$  - coords in lensed / 'image' plane.

In weak lensing regime (linearise mapping approx)

$$I(\vec{\theta}) = I^{(s)} \left[ \vec{\theta}_0 + \frac{\partial \vec{\beta}}{\partial \vec{\theta}} \cdot (\vec{\theta} - \vec{\theta}_0) \right]$$

$$\frac{\partial \vec{\beta}_i}{\partial \theta_j} = \gamma_{ij} - \gamma_i \gamma_j \kappa \quad (\text{from lens equation})$$

$$\begin{pmatrix} 1 - \kappa - \gamma_1 & -\gamma_2 \\ -\gamma_2 & 1 - \kappa - \gamma_1 \end{pmatrix} \quad \begin{matrix} \gamma = \text{'Shear'} \quad \gamma_1, \gamma_2 \\ \kappa = \text{Convergence} \end{matrix}$$

Jacobian lensing Equation

Effects on a Circular isophotes:  $I^s(\theta_1, \theta_2)$  for  $(\theta_1, \theta_2) = (R \cos \phi, R \sin \phi)$

$$I(\vec{\theta}) = I^s[\vec{\beta}(\vec{\theta})] = I^s[\vec{\theta}_0 + A \cdot (\vec{\theta} - \vec{\theta}_0)]$$

= 1 when  $A \cdot \vec{\theta} = (R \cos \phi, R \sin \phi)$

$$\vec{\theta} = A^{-1} \cdot (R \cos \phi, R \sin \phi)$$

Special Cases:

$$\gamma_1 = \gamma_2 = 0$$

$$(\theta_1, \theta_2) = (1 + \kappa) (R \cos \phi, R \sin \phi)$$

$$\gamma_2 = 0$$

$$(\theta_1, \theta_2) = ((1 - \kappa - \gamma_1)^{-1} R \cos \phi, (1 - \kappa + \gamma_1)^{-1} R \sin \phi)$$

Ellipticity

$$|\varepsilon| = \frac{a-b}{a+b}$$

$$\varepsilon = |\varepsilon| e^{2i\phi}$$

$$\varepsilon_1 = |\varepsilon| \cos 2\phi$$

$$\varepsilon_2 = |\varepsilon| \sin 2\phi$$

In weak lensing regime:  $\kappa, \gamma \ll 1$ :  $\rightarrow$  Initial shape of galaxy.

$$\varepsilon_{1,2}^{\text{obs}} = \varepsilon_{1,2}^{\text{intrinsic}} + \gamma_{1,2}$$

Orientation ellipse globally = 0 (although may be ellipses - no preferred orientation).

$$\langle \varepsilon_{1,2}^{\text{intrinsic}} \rangle = 0 \rightarrow \langle \varepsilon_{1,2}^{\text{obs}} \rangle = \gamma_{1,2}$$

$\downarrow$  Allows it  $\gamma_{1,2}$  to be inferred on large scales.

Shear of an image:

Quantify ellipticity via quadrupole moments:

$$\varepsilon_1 = \frac{Q_{xx} - Q_{yy}}{Q_{xx} + Q_{yy} + 2(Q_{xx}Q_{yy} - Q_{xy}^2)^{1/2}}$$

$$\varepsilon_2 = \frac{Q_{xy}}{Q_{xx} + Q_{yy} + 2(Q_{xx}Q_{yy} - Q_{xy}^2)^{1/2}}$$

# CMB Lensing (Blake Sherwin)

## Basics:

Galaxies: Gas: < 20%

Dark Matter: > 80%

Cosmic structure grows from gravitational collapse of dark matter and gas. (Runaway collapse).

Sensitive to initial conditions - theory of gravity etc.

Large Scale Dark Matter Structure:

## Gravitational Lensing on CMB Radiation $\Theta(\vec{r})$

CMB: Oldest source - most distant source of light - passes through all of universe.

Distribution of Dark Matter deflect CMB light that it passes through (Preserves Surface Brightness).

↳ Original, unlensed, CMB fluctuations

↳ Very well understood statistical properties, e.g. isotropy

>Add lensing

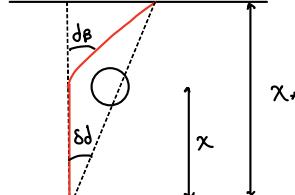
↳ Causes magnification / shearing.

## Approximate Quantification:

$$\tilde{\Theta}(\vec{r}) = \Theta(\vec{r} + d(\vec{r})) \quad \text{Lensing deflection field: } d$$

↳ Described by remapping by vector field:  $d$ .

## CMB Last Scattering Surface:



CMB is deflected by mass:

Described by i component of geodesic equation:

$$\frac{dp^i}{d\lambda} + \Gamma^i_{jk} p^j p^k = 0$$

Derive by  $\delta B = -2 \delta x \nabla_\perp \Phi$

Deflection Angle

Distance

Grav. Potential

$$x_* \delta d = (x_* - x) \delta B, \quad \delta d = \frac{x_* - x}{x_*} \delta B$$

Hence:

observed deflection angle

$$\delta d = -2 \frac{x_* - x}{x_*} \delta x \nabla_\perp \Phi$$

Perpendicular Gradient.

The lensing is related to an integral of the potential:

$$\text{Total Deflection} \quad d(\vec{r}) = - \int_0^{x_*} 2 dx \frac{x_* - x}{x_*} \nabla_\perp \Phi(x, n_0 - x)$$

Small angle approx  $\nabla_\perp \Phi \approx \nabla_\perp \Phi / x$

$$d(\vec{r}) = \nabla \phi$$

$$\phi(\vec{r}) = -2 \int_0^x dx \frac{x_* - x}{x_*} \Phi$$

### Lensing Convergence:

$$K = -\frac{1}{2} \nabla^2 \phi = \int_0^{x_0} \frac{x_0 - x}{x_0 x} x^2 \nabla_{\perp}^2 \Phi dx$$

$\nabla_{\perp}^2 \Phi = \frac{3}{2a} \Omega_{m,0} H_0^2 \delta$

Overall:  $K = \Omega_{m,0} \int_0^x dx \frac{3H_0^2}{2a(x)} \frac{(x_0 - x)x}{x_0} \delta$

### Overall point:

↳ Lensing probes the projected total mass density in each direction (of which most is dark matter)

### The Power Spectrum of a lensing map: $C_L^{\phi\phi}$

$$\langle \phi(L) \phi^*(L') \rangle = (2\pi)^2 C_L^{\phi\phi} \delta(L-L')$$

### Flat Sky Approximation:

Neglect the Curvature of the Sky:

$$\langle a_{lm}^{\phi} (a_{l'm'}^{\phi})^* \rangle = C_L^{\phi\phi} \delta_{ll'} \delta_{mm'}$$

$\boxed{\langle \phi(L) \phi^*(L') \rangle = (2\pi)^2 C_L^{\phi\phi} \delta(L-L')}$

### What does $C_L^{\phi\phi}$ tell us:

$\delta$ -density fluctuation

Lensing power spectrum probes Square density fluctuation  $\langle \delta^2 \rangle$

↳  $\sim \sigma_8 \sim \sqrt{\langle \delta^2 \rangle} \approx$  'Clumpiness of parameter'

Must also consider neutrino mass.

### Lensing Reconstruction:

2-point Correlation function (describes Gaussian Random Field).

↳  $\xi(x, x') = \langle \Theta(x) \Theta(x') \rangle = \int \Theta(\underline{x}) \Theta(\underline{x}') \Pr[\Theta]$

Require Statistical Isotropy / Translation Invariance:

$$\xi(x, x') = \xi(x+a, x'+a)$$

Fourier Transform:  $\Theta(\underline{x}) = \int \frac{d^3 L}{(2\pi)^3} \Theta(L) e^{i L \cdot \underline{x}}$

### Translational Invariance effect on Fourier Coefficients

$$\begin{aligned} \xi(x, x') &= \langle \Theta(x), \Theta(x') \rangle = \int \frac{d^3 L}{(2\pi)^3} \int \frac{d^3 L'}{(2\pi)^3} e^{i L \cdot x + i L' \cdot x'} \langle \Theta(L) \Theta(L') \rangle \\ &= \xi(x+a, x'+a) = \int \frac{d^3 L}{(2\pi)^3} \int \frac{d^3 L'}{(2\pi)^3} e^{i L \cdot x + i L' \cdot x' + i(L+L') \cdot a} \langle \Theta(L) \Theta(L') \rangle \end{aligned}$$

Thus  $\langle \Theta(L) \Theta(L') \rangle \propto \delta^{(D)}(L+L')$

Implies:

$$\langle \Theta(l) \Theta(l') \rangle = (2\pi)^2 C_{ll} \delta^{(0)}(l+l')$$

$\Theta^*(l) = \Theta(-l)$  ↗ C power Spectrum of CMB

Complex Conjugate:

$$\langle \Theta(l) \Theta^*(l') \rangle = (2\pi)^2 C_{ll'} \delta^{(0)}(l-l')$$

Different modes are uncorrelated

$$\langle \Theta(l) \Theta^*(l-l') \rangle = 0 \text{ without lensing}$$

Lensing breaks this translational invariance:

$$\langle \tilde{\Theta}(l) \tilde{\Theta}^*(l-l') \rangle \sim \phi(l)$$

measure lensing by looking for these new correlations in the CMB two-point functions

Proof:  $\tilde{\Theta}(x) = \Theta(x+d) = \Theta(x) + \nabla \phi \cdot \nabla \Theta + \dots$

$$\tilde{\Theta}(l) = \Theta(l) - \int \frac{d^2 l'}{(2\pi)^2} [l' \Theta(l')] \cdot (l-l') \phi(l-l')$$

$$(2\pi)^2 C_l \delta(l+l')$$

$$\langle \tilde{\Theta}(l) \tilde{\Theta}^*(l-l') \rangle = - \int \frac{d^2 l'}{(2\pi)^2} [l' \underbrace{\langle \Theta(l) \Theta(l') \rangle}_{(2\pi)^2 C_l \delta(l+l')} (l-l-l-l') \phi(l-l-l-l')]$$

$$= [l C_l] l \phi(l) = \underbrace{K(l, l)}_{\text{Response Function.}} \phi(l)$$

Quadratic Lensing Estimator:

$$\hat{\phi}(l) = \int \frac{d^2 l}{(2\pi)^2} j(l, l) \Theta(l) \Theta(l-l)$$

$$\hat{\phi}(l) \sim \frac{\Theta(l) \Theta(l-l)}{K(l, l)}$$

Say have  $i$  measurements of quantity  $x, \hat{x}$

$$\langle \hat{x}_i \rangle = x$$

$$\text{different } \sigma(x_i) = \sigma_i$$

$$\text{Best estimate of } x: \hat{x} = \frac{\sum w_i \hat{x}_i}{\sum w_i} \quad w = \frac{1}{\sigma_i^2}$$

Combine with inverse-variance weight:

"Optimal" Quadratic Estimator:

$$\hat{\phi}(l) = N(l) l \cdot \int \frac{d^2 l}{(2\pi)^2} \frac{l C_l}{C_l^{\text{noisy}}} \Theta(l) \frac{1}{C_{l-l}^{\text{noisy}}} \Theta^*(l-l)$$

Sum over all pairs of modes

A fast approach:

Using Convolution theorem  $\text{IFT}(A * B) = \text{IFT}(A) \times \text{IFT}(B)$

(IFT - Inverse Fourier Transform)

Thus:

$$A * B = \text{FT}(\text{IFT}(A) \times \text{IFT}(B)) \rightarrow \text{Scales } N \log N \text{ vs } N^2$$

$$\hat{\phi}(l) = N(l) l \cdot \int \frac{d^2 l}{(2\pi)^2} \left[ \frac{l C_l}{C_l^{\text{noisy}}} \Theta(l) \right] \left[ \frac{1}{C_{l-l}^{\text{noisy}}} \Theta^*(l-l) \right]$$

$$l F_2(l) \qquad F_1(l-l)$$

$$= N(l) l \cdot \text{FT} \left[ \text{IFT}(F_1) \times \text{IFT}(l F_2) \right] = -N(l) i l \int d^2 x e^{-il \cdot x} F_1(x) \nabla F_2(x)$$

### Reconstruction Noise:

$$\hat{C}_L^{\phi\phi} = \langle \hat{\phi}(l) \hat{\phi}^*(l) \rangle \sim \langle TTTT \rangle \sim C_L^{\phi\phi} + \langle T\tau \rangle \langle \tau T \rangle$$

### Cross Correlation:

$$\langle T_1 T_2 \rangle = \langle T T \rangle + \langle n_1 \rangle \langle T \rangle + \langle n_2 \rangle \langle T \rangle + \langle n_1 \rangle \langle n_2 \rangle$$

### Challenge 2: Fore grounds

$$T(l) = T^{\text{CMB}}(l) + T^{\text{fg}}(l)$$

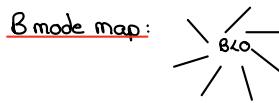
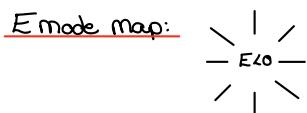
### Effect on 4 point function:

$$\begin{aligned} \hat{C}_L^{\phi\phi} \sim & \langle TTTT \rangle \sim \langle T^{\text{CMB}} T^{\text{CMB}} T^{\text{CMB}} T^{\text{CMB}} \rangle \\ & + \langle T^{\text{fg}} T^{\text{fg}} T^{\text{fg}} T^{\text{fg}} \rangle \\ & + \langle T^{\text{CMB}} T^{\text{CMB}} T^{\text{fg}} T^{\text{fg}} \rangle + \dots \end{aligned} \quad \left. \right\} \text{foreground bias terms.}$$

Correct for: frequency based Separation.

Delensing (ie extracting original CMB Map before lensing effect).

### CMB Polarization Basics

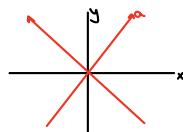


B-mode: Contains signals from inflation.

Describe with Q and U fields defined via electric fields:

Stokes Q, U parameters:

$$\begin{aligned} Q &\propto E_x^2 - E_y^2 \\ U &\propto E_x^2 + E_y^2 \end{aligned}$$



### CMB Polarization:

$$Q(\vec{x}) = \int \frac{d^2 L}{(2\pi)^2} (E_1 \cos 2\phi_1 - B_1 \sin 2\phi_1) e^{i \vec{L} \cdot \vec{x}}$$

aligned/orientated with wavevector.

$$U(\vec{x}) = \int \frac{d^2 L}{(2\pi)^2} (E_1 \sin 2\phi_1 + B_1 \cos 2\phi_1) e^{i \vec{L} \cdot \vec{x}}$$

B mode polarization Channel  $\rightarrow$  Primordial GW from inflation.

However:

Lensing also produced B-mode polarization by  
Converting E to B modes.

For CMB Polarisations:

$$\tilde{B}(l) = \int d\ell W(l, l') E(l') \phi(l - l') \quad W(l, l') = l' \cdot (l - l') \sin(2\phi_{l, l'})$$

If I have a lensing estimate:

$\hookrightarrow$  Delensing: Construct Lensing  $\sim E\phi$  map from measured  $\phi$  and  $E$   
and Subtract B-Lensing

## Control Variance:

Say have an estimator  $m$  with  $\langle m \rangle = N$ , unknown.

wish to lower variance:

Have another Statistic  $t$  with  $\langle t \rangle = T$

New unbiased estimator:  $m^* = m - C(t - T)$

Minimise Variance in  $m^*$  for  $C = \frac{\text{Cov}(m, t)}{\text{Var}(t)}$

$$\text{Variance}(m^*) = (1 - \rho_{m,t}^2) \text{Var}(m)$$

$\swarrow$  Correlation Coefficient.

## Posterior Analysis:

The evidence  $P(d) = \int d\theta P(d|\theta) P(\theta)$ .

### Affine Invariant Sampler

#### Affine transformation:

$$\alpha' = A \cdot \alpha + b$$

#### Markov Process:

$$\alpha_{t+1} = R(\alpha_t, \xi, P(\alpha)) \rightarrow \text{update algorithm}$$

### Affine Invariant Sampler:

$$R(\alpha \cdot c + b, \xi, P) = A R(\alpha, \xi, P) \cdot c + b$$

## Example from Cosmology

CMB lensing distorts the CMB:

$$T_{\text{obs}}(\vec{r}) = T_{\text{lens}}(\vec{r}) + \nabla \phi(\vec{r})$$

Estimate of  $\phi$   $\hat{\phi} \sim \sum T_{\text{obs}} T^{\dagger}_{\text{obs}}$