

Delensing, Lensing, and Neural Networks – a perspective from the CMB W.L. Kimmy Wu KICP Fellow University of Chicago Kavli Institute for Cosmological Physics

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### Outline

- Background
- \* Delensing of BICEP/Keck B-mode maps, with Planck + SPTpol E-mode maps
- Lensing reconstruction from SPTpol
- Convolutional Neural Networks on CMB lensing
- Summary and outlook





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**BICEP/Keck Array** 

### The CMB is sensitive to our standard parameters and more!



Background Figure: BICEP/Keck collaboration

CMB at recombination is sensitive to photon, matter, and baryon densities, initial \* conditions (A\_s, n\_s)

And more! Primordial gravitational waves, dark radiation, ... \* Kimmy Wu, UChicago 4

### $T(\hat{n}) \ (\pm 350 \mu K)$





(no primordial B-modes)

unlensed

### $T(\hat{n}) \ (\pm 350 \mu K)$



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(no primordial B-modes)

lensed

### Effect of lensing on the spectra



Planck's all sky CMB

temperature map

scale ±500 µK



### Are the effect of lensing limiting our parameter constraints?



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### Delensing for *r* : a BICEP/Keck example



- We can fit lensing model + *r* simultaneously, but limited by sample variance of lensing
- **Delensing** B-modes: using the *realization-specific* lensing B-mode sky to reduce lensing sample variance
- Especially important if observing a small sky patch



Telescope and Mount Focal Plane Beams on Sky

### Delensing: the idea

1. Use Phi tracer and lensed E map to get estimate of lensing B modes



2. Cross-correlate the lensing B template with observed B mode map to quantify how much lensing B modes are in the observed map



B template





### CIB as a $\phi$ tracer

- \$\overline{\phi}: can reconstruct from CMB, but S/ N rather low currently (Future will be better!)
- Cosmic infrared background (CIB) from dusty star-forming dusty galaxies with redshift distribution peaked between z~1 and 2.
- CMB lensing potential's redshift kernel peaks between 1 < z < 3</li>
- Cross-correlation can be as high as ~80%



### Lensing template construction





2. Difference the pre- and post-deflected map





Feed the Q/U map through a B-estimator to get the power spectra as inputs to the multicomponent analysis.

# Connecting delensing to $\sigma(r)$

BICEP/Keck analysis framework:

how is delensing incorporated

### BK multicomponent analysis (no delensing)

• Input maps to multicomponent analysis that extracts constraints on *r* 



Maps from BICEP/Keck (95/150GHz)



Maps from Planck



### BK multicomponent analysis (no delensing)

- Take the auto- and cross-spectra of the BICEP/Keck and WMAP/Planck maps
- To calculate the likelihood, compare the data bandpowers against the model expectation values of lensing BB, *r*, and 7 parameter foreground model:

 $A_{\text{dust}}, \alpha_{\text{dust}}, \beta_{\text{dust}}, A_{\text{sync}}, \alpha_{\text{sync}}, \beta_{\text{sync}}$ dust/sync correlation



### BK15 constraints



### BK multicomponent analysis (+ delensing)

• Input maps to multicomponent analysis that extracts constraints on *r* 



#### Maps from BICEP/Keck (95/150GHz)





#### Maps from Planck



### Inputs to BK lensing template

- Phi tracer: Planck's CIB \* map
- Q/U maps: combination of \* BICEP/Keck, SPTpol, and Planck maps





### Incorporating lensing template to likelihood

- Use same model: lensing BB, *r*, and 7 parameter foreground model:  $A_{\text{dust}}, \alpha_{\text{dust}}, \beta_{\text{dust}}, A_{\text{sync}}, \alpha_{\text{sync}}, \beta_{\text{sync}}$ , and dust/sync correlation
- Adding the lensing template increases the total auto/cross BB spectra from 66 to 78

$$\mathcal{C}(\theta|d) \propto \frac{1}{\sqrt{|\mathbf{C}(\theta)|}} \exp\left(-\frac{1}{2}(d-\mu(\theta))^{\dagger} [\mathbf{C}(\theta)]^{-1}(d-\mu(\theta))\right)$$
  
more data bandpowers

In the BK analysis, we use the HL likelihood. The gaussian likelihood is for illustration purposes.

### Lensing template as input in multicomponent analysis

The covariance matrix that enters the likelihood has information of the covariance between the lensing BB spectrum and the observed BB spectrum -> reducing  $\sigma(r)$ .



### How much do we improve $\sigma(r)$ ?



- With perfect φ map (no decorrelation, no noise), adding a lensing template to the BK14 data set improves σ(r) from 0.025 to 0.018
- Using CIB phi tracer to form the lensing template, σ(r) improves by ~10% from BK14

### Current limitation to delensing

- B mode variance is dominated by galactic foregrounds; even with perfect delensing we do not improve σ(r) very significantly
- CIB map we use has cross-correlation with underlying φ at 60-80%; need better cross-correlation to improve towards perfect delensing
- \* CMB-derived φ from next-generation CMB experiments can provide that!

Key take-aways:

Incorporated delensing into a likelihood analysis for *r*;
 Delensing reduces σ(r) by ~10% for the BK14 dataset.

### Getting a better $\phi$ tracer: CMB lensing reconstruction

Lensing correlates CMB modes across angular scales These off-diagonal correlations  $\propto \phi(L)$ ; can use the correlations to measure  $\phi$ !

 $T_{\text{len}}(\hat{n}) = T_{\text{unl}}(\hat{n} + \nabla \phi)$ 

 $(Q \pm iU)_{\text{len}}(\hat{n}) = (Q \pm iU)_{\text{unl}}(\hat{n} + \nabla\phi)$ 

### Lensing reconstruction: quadratic estimator

In equations,  $\phi(L)$  can be estimated as follows:



The estimated  $\phi(L)$  is a weighted sum of the products of Fourier modes from X and Y for all the pairs of  $\ell$  and  $\ell'$  where  $L = \ell + \ell'$ ;

It picks out the correlations in the lensed CMB maps introduced by  $\phi(L)$ .



## The South Pole Telescope (SPT)

10-meter sub-mm quality wavelength telescope 95, 150, 220 GHz and 1.6, 1.2, 1.0 arcmin resolution

### 2007: SPT-SZ

960 detectors 95,150,220 GHz



2016: SPT-3G ~16,000 detectors 95,150,220 GHz +Polarization









## SPT surveys



### SPTpol lensing map





Work led by M. Mocanu

## Cosmology from the $Cl^{\phi\phi}$ spectrum



neutrino mass

\*  $\sigma_8/A_{lens}/\Omega_M$  (cross-correlate with/compare against optical surveys)

### Lensing map noise



High S/N per lensing mode measurement in the SPTpol patch important for delensing 30

## Delensing efficiency

Cross-correlation of tracer and  $\phi$ -field  $\rho_{\ell} = \frac{C_l^{\text{tracer-}\phi}}{\sqrt{C_l^{\text{tracer-tracer}}C_l^{\phi\phi}}}$ For CMB reconstructed  $\phi$  $\rho_{\ell} = \sqrt{\frac{C_{\ell}^{\phi\phi}}{C_{\ell}^{\phi\phi} + N_{\ell}^{\phi\phi}}}$ 



~scales of lenses that source most lensing B-modes

 In the limit that the E-mode noise is small, the correlation between the φ tracer and the underlying phi field determines how well the lensing B-modes are estimated —> delensing efficiency

### Forecasts (SPT-3G / CMB-S4)



CMB reconstructed  $\phi$  will soon be the best lensing potential tracer for B-mode delensing

### **BICEP Array + SPT-3G**



Assuming BK15 foreground model:  $A_{dust}$ ,  $\alpha_{dust}$ ,  $\beta_{dust}$ ,  $A_{sync}$ ,  $\alpha_{sync}$ ,  $\beta_{sync}$   $\sigma(r)$  saturates without delensing even with the addition of 30/40 GHz and 220/280 GHz receivers in BICEP Array

#### Key takeaways:

- CMB lensing map from SPTpol survey has S/N > 1 measurements for L < 250 modes;</li>
- 2) Precise measurement of CMB lensing amplitude (~6%) and will provide relevant constraints for cosmological parameters
- 3) CMB reconstructed phi will soon be competitive for delensing.
- 4) BICEP Array + SPT-3G delensing is projected to give σ(r) ~ 0.003.

## Neural networks for CMB lensing reconstruction

(arXiv: 1810.01483)

### Why neural network?



- network needs to perform transformation from one set of images to another; seem to be a good fit for neural networks
- Real need of beyond quadratic estimators to get optimal lensing reconstruction

### It works!



solid lines: neural net dashed lines: maximum likelihood approx. 1uK-arcmin: ~CMB-S4 level noise

### Network architecture



• 11200 sets of Q/U, E/K sim maps; 80:10:10 training:validation:test sets

### Convolution layer





 $A_i^j$  is what training determines.

There are 2x128<sup>2</sup> of these 5x5 grids from one pair of Q/U maps. It outputs a 1x128<sup>2</sup> image.

> For this first step, we create 64 versions of the 1x128<sup>2</sup> image

 $2 \times 128^2$ 

 $64 \times 128^2$ 

 $(\tilde{Q}, \tilde{U})$ 

### The first 12 of the 64 outputs at the first step



### Network architecture



- The loss function is MSE (mean square error) between the output E/K and the true E/K; choose  $A_i^j$  to minimize MSE.
- Residual UNet; "residual connections" at before each dimension changing step; "skip connections" across the layers with same dimensions

### Lensing field recovery



### Unlensed E-mode recovery



### Compare to "physics-ful" methods



solid lines: neural net dashed lines: maximum likelihood approx. 1uK-arcmin: ~CMB-S4 level noise

- The NN approach doesn't completely recover the input
- We quantify that decorrelation as noise in our reconstruction
- Standard reconstruction
   has noise terms due to
   spurious correlations of
   random gaussian fields,
   etc.
- Reaching similar levels of noise as maximum likelihood methods!

### Tests

2)

Is the network really sensitive to lensing??

- Null test
- Sensitive to differences in input  $\Omega_M$

How sensitive are the outputs to the initialization randomness?

 Randomness due to initialization

Toy fit for cosmology

Fit Ω<sub>M</sub>



Key takeaways:

- The network κ recovery's S/N is similar to maximumlikelihood methods;
- 2) Network is sensitive to changes in cosmology;
- 3) Lots to explore in understanding how the network extracts information.

## Summary

- Lensing variance will dominate σ(r) in the next few years; σ(r) is currently dominated by foreground and instrumental noise uncertainties.
  - \* First demonstration of  $\sigma(r)$  reduction underway.
- Lensing potential reconstructed from CMB maps will be competitive φ tracers for delensing B-modes by SPT-3G era.
  - SPTpol lensing potential is amongst the highest S/N per mode measurement to date.
- \* Future low-noise experiments will benefit from beyond quadratic estimator lensing/delensing. Neural network is a viable technique.

## Thank you for listening!