

# Sparse estimation of model-based diffuse thermal dust emission

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## Overview

Brief introduction to CMB component separation

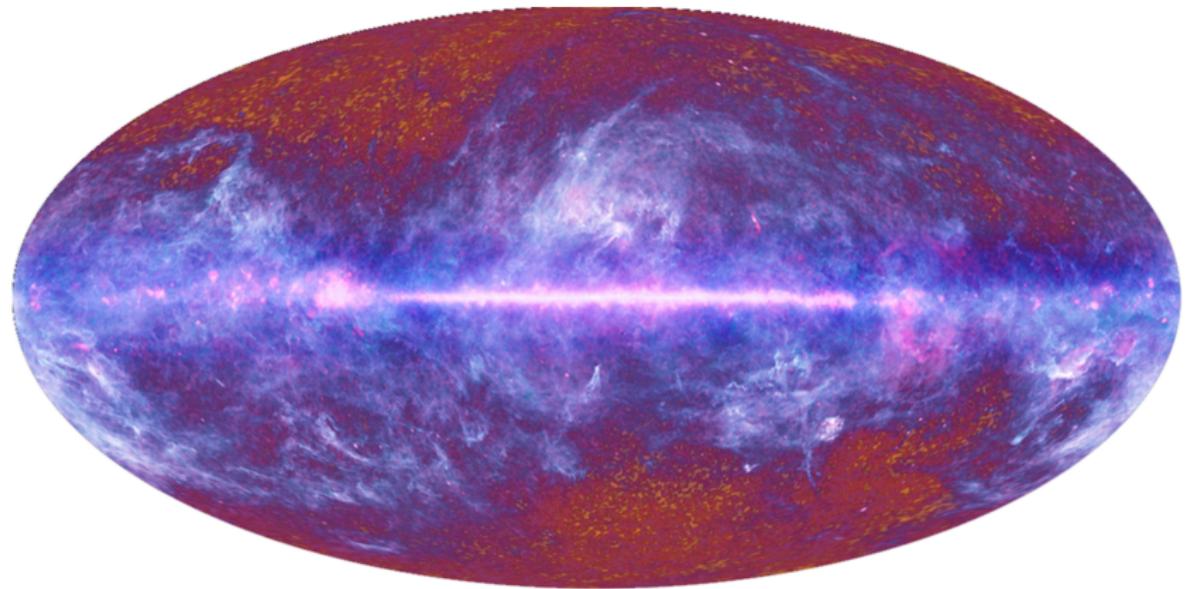
The problem at hand: thermal dust and the CIB

How sparsity can help

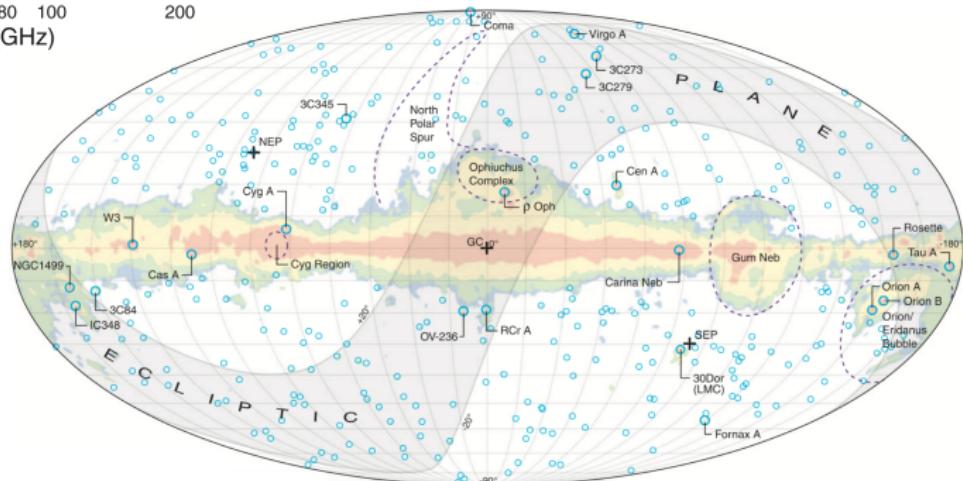
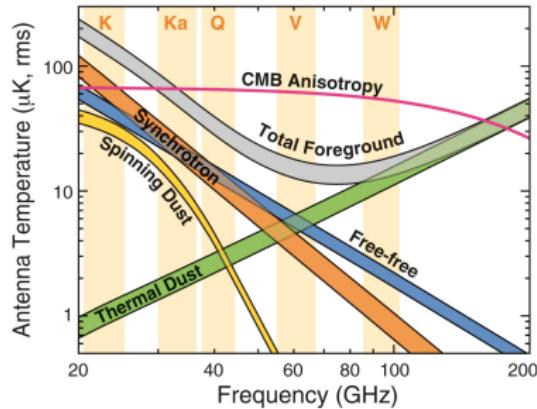
The new methodology

Validation on Planck simulation data

## CMB component separation



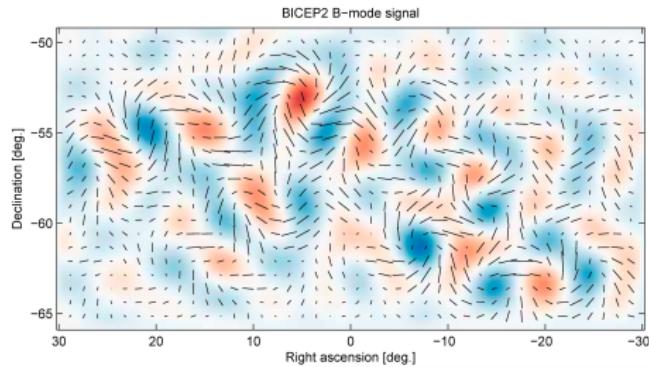
# CMB component separation



Bennett et al. 2013

# Thermal dust and the CIB

BICEP2 2014



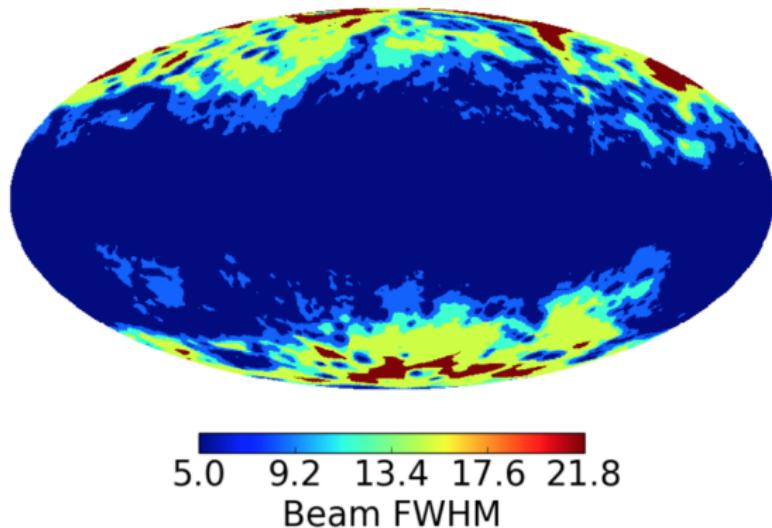
- Focus of this work:

$$x_{\nu_i}^{\text{dust}} = \tau_{353} \times B(T, \nu) \times \left(\frac{\nu}{353 \text{ GHz}}\right)^{\beta}$$

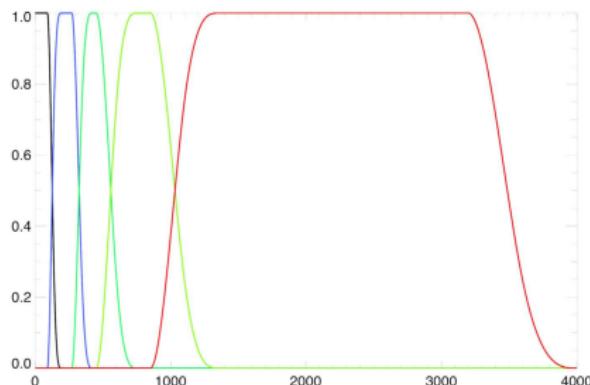
- CIB - unresolved galaxies
- Smoothing conundrum!

$$\text{total flux} = \text{dust} + \underbrace{\text{CIB} + \text{CMB} + \text{noise}}_{\text{nuisance, Gaussian approx}}$$

- Clever smoothing using nuisance estimates.

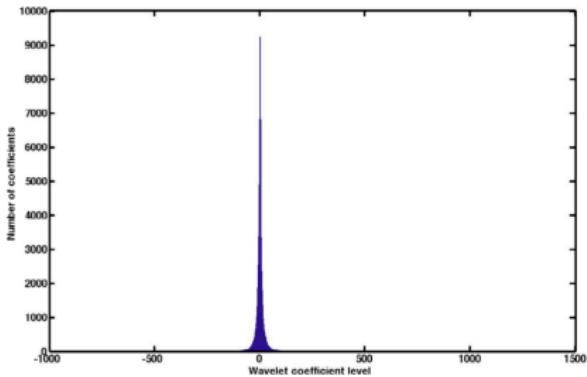
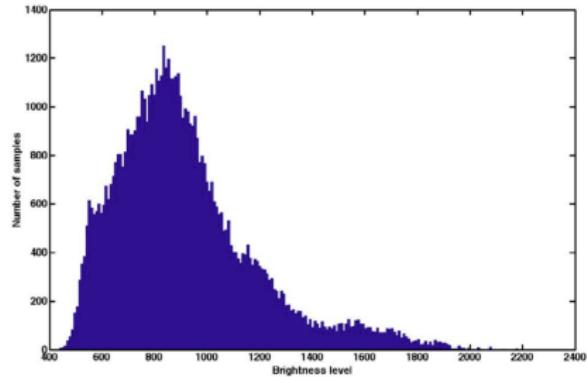


# Sparsity and the wavelet domain



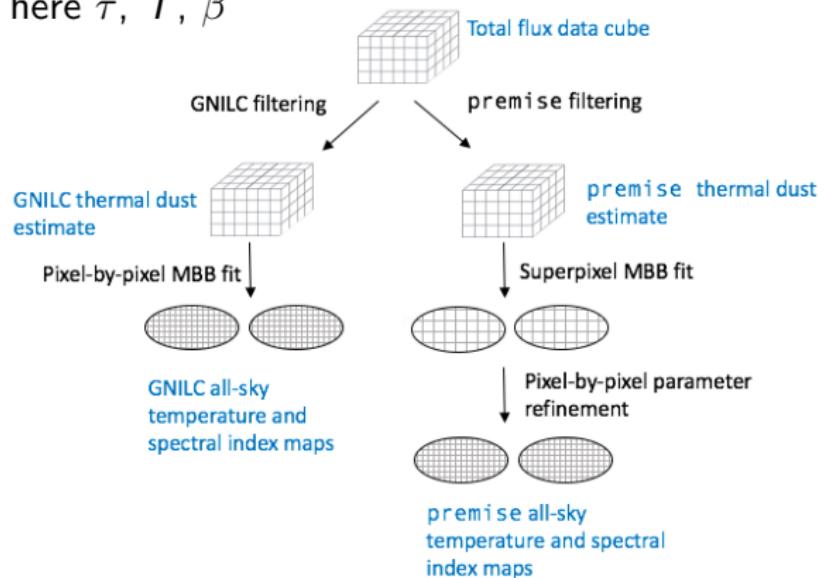
*Bobin et al. 2013*

- Sparse: majority of signal is zero
- Spatially correlated source
- Wavelets filter in spherical harmonic domain (x-axis:  $\ell$ )



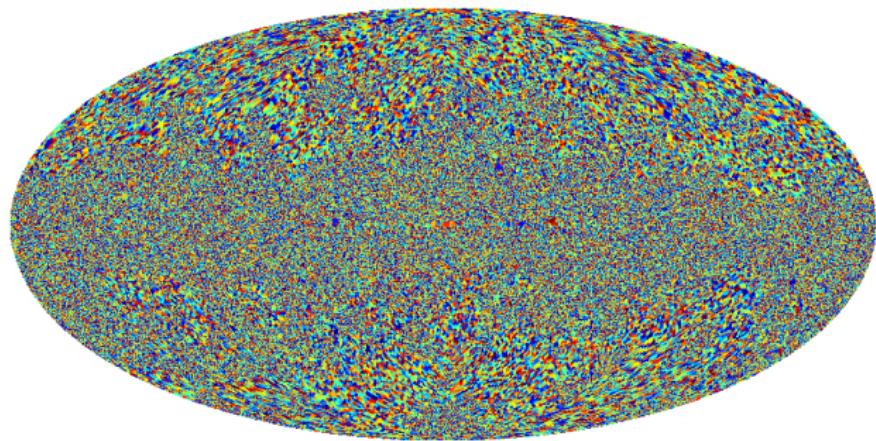
# PREMISE

- Parameter Recovery Exploiting Model Informed Sparse Estimates
- Requires model - here  $\tau$ ,  $T$ ,  $\beta$



## Filtering and Super-pixels

- Essentially GNILC filtering BUT penalise in favour of sparsity
- Accurate and fast parameter estimates from fit -  $\tau$ ,  $T$ ,  $\beta$



## Refinement

- Low resolution, fast informed initial guesses
- $T$  and  $\beta$  refinement - normalisation factor subject to degeneracies

$$x_{\nu_i}^{\text{dust}} = \tau_{353} \times B(T, \nu) \times \left( \frac{\nu}{353 \text{ GHz}} \right)^{\beta}$$

- Gradient descent at each pixel (until convergence)

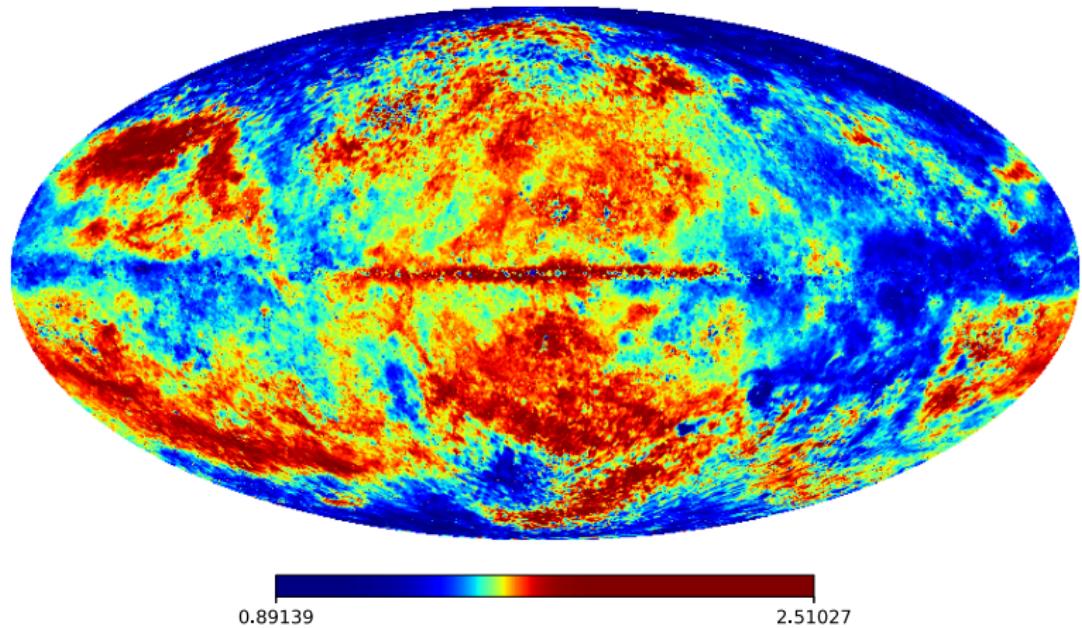
$$\beta_n / T_n = \beta_0 / T_0 + \rho \times \Delta((\text{Data} - \text{model}) \text{ w.r.t } \beta \text{ and } T)$$

$$\tau_{353} = \frac{X_{857}}{B(T, 857 \text{ GHz}) \times \left( \frac{\nu}{353 \text{ GHz}} \right)^{\beta}}$$

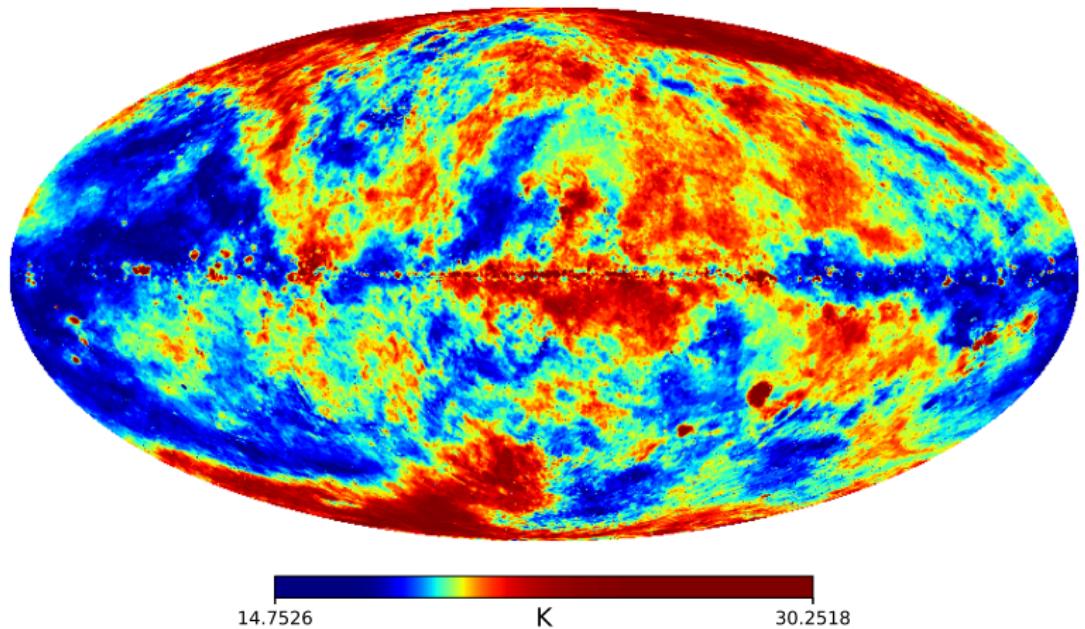
## Validation

**simulation data only**

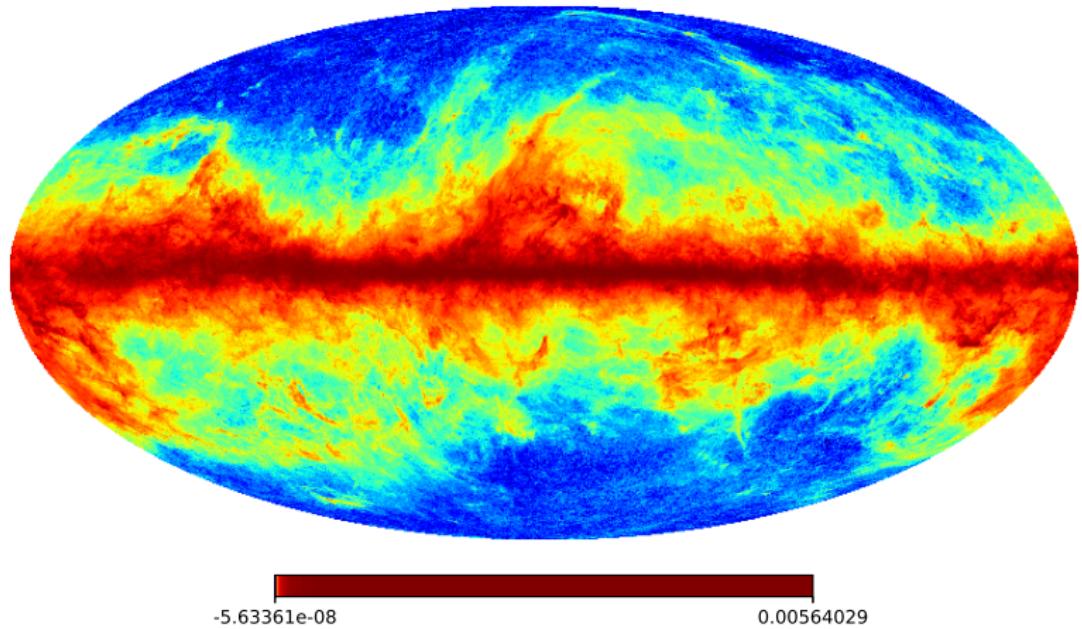
## Full-sky $\beta$ estimate - 5 arcmin



## Full-sky $T$ estimate - 5 arcmin

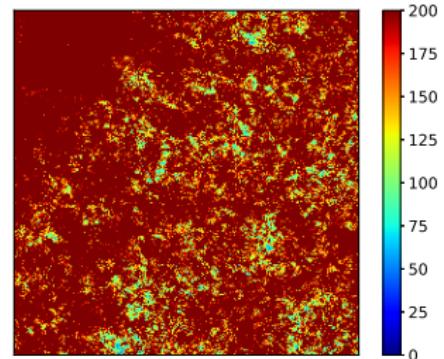
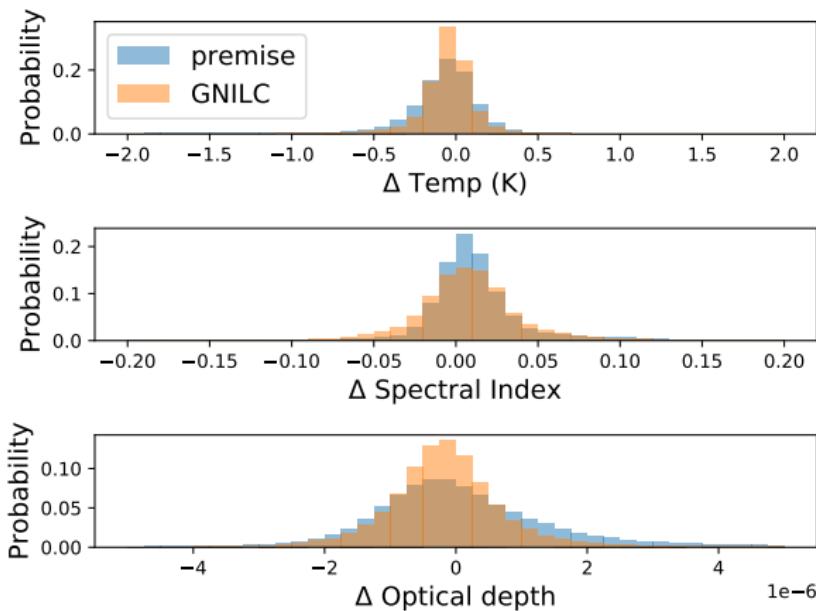


## Full-sky $\tau_{353}$ estimate - 5 arcmin

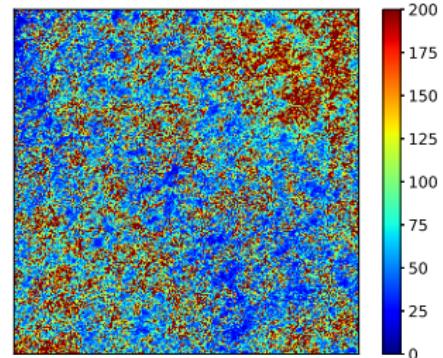
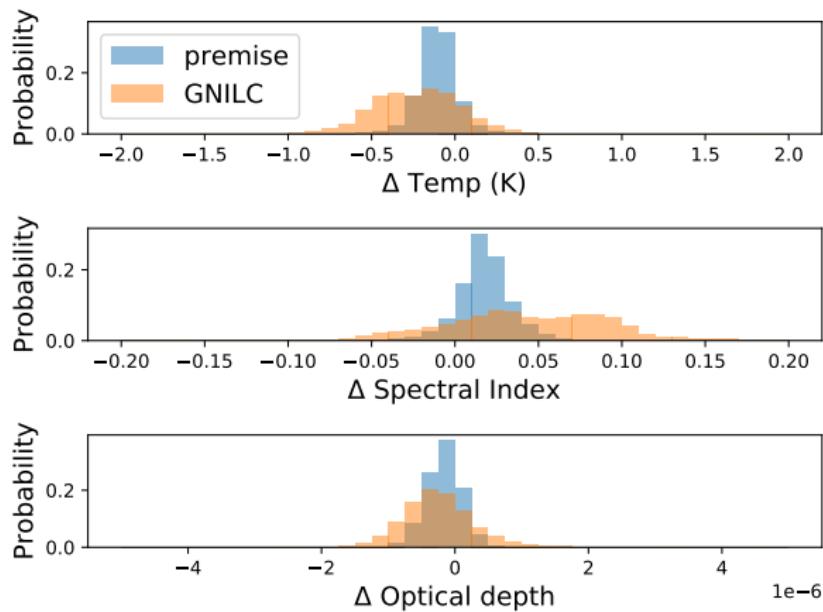


Value	Medium	$\% \Delta$	$1\sigma$	$\% \Delta$	$2\sigma$	$\% \Delta$	$3\sigma$	$\% \Delta$
Temperature		1.7	2.8		8.0		16.5	
Spectral index		3.4	5.7		15.4		25.6	
Optical depth (353 GHz)		3.7	7.2		31.2		77.0	

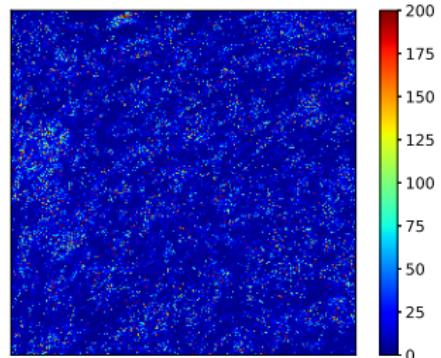
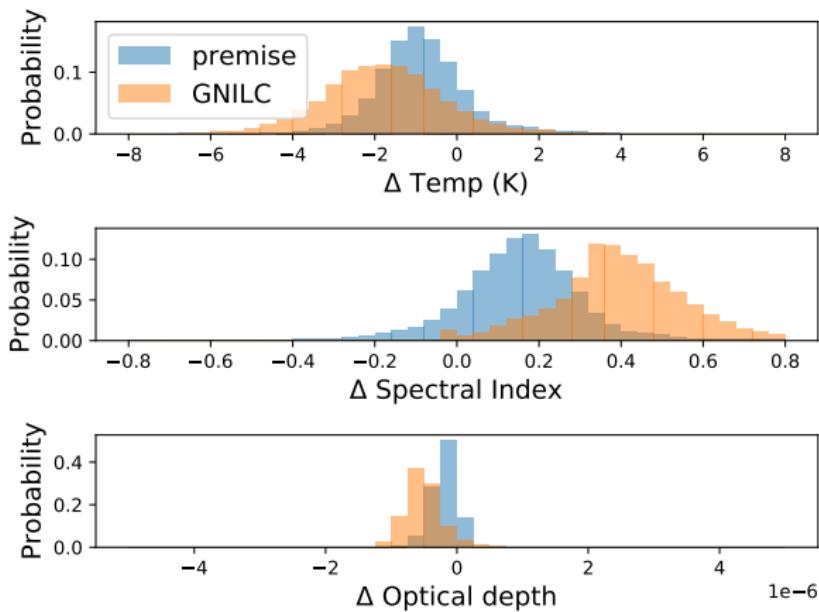
# Region 1 - High SNR



## Region 2 - Medium SNR



# Region 3 - Low SNR



## Conclusion

- Fast recovery of model parameters: full sky (varying signal to noise) at full resolution
- Sparsity in place of smoothing
- Improvement for all but the largest signal to noise regions