

# Performance-Based Inflation Forecasts for CMB Stage 3 and Stage 4

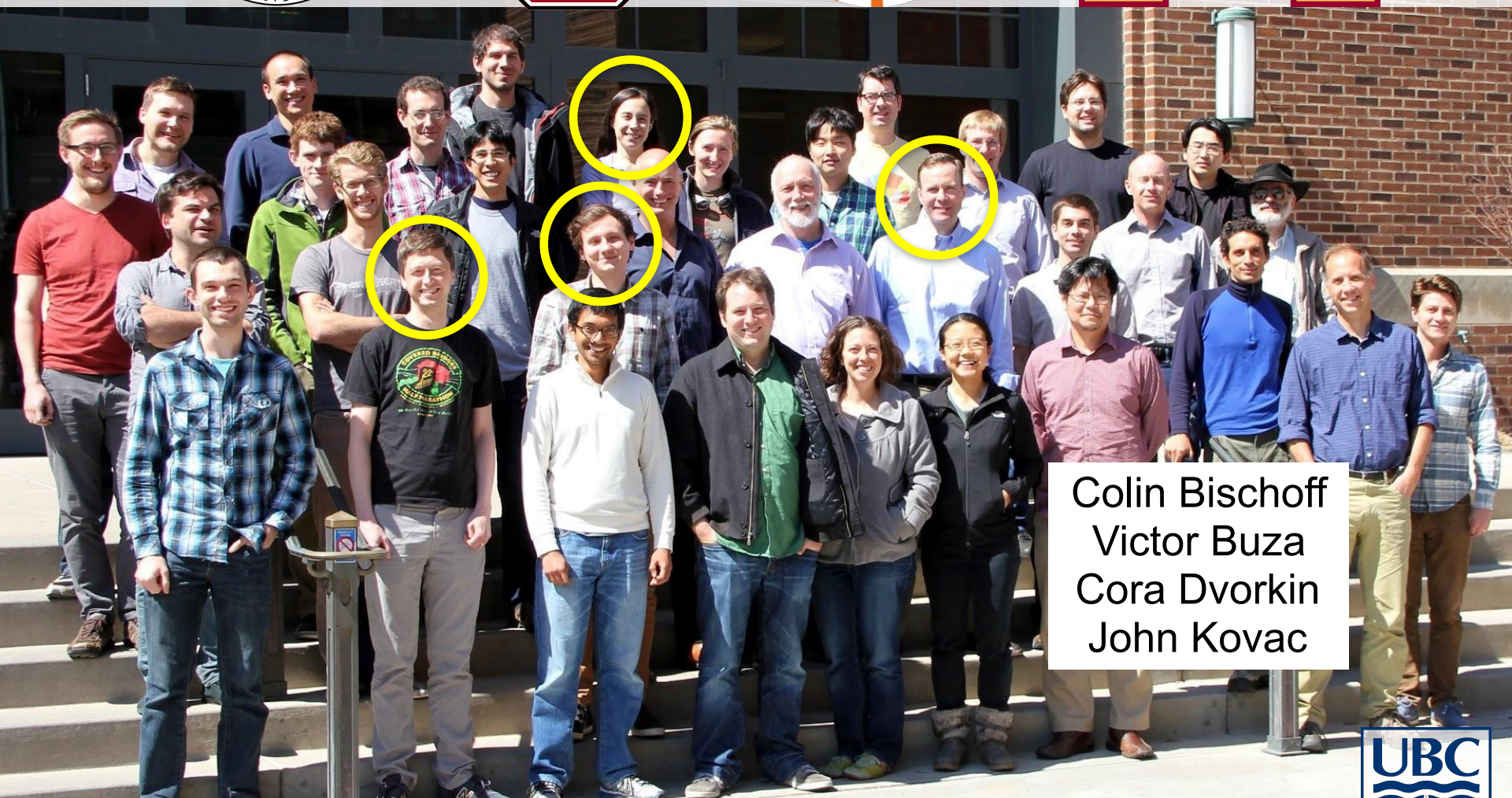
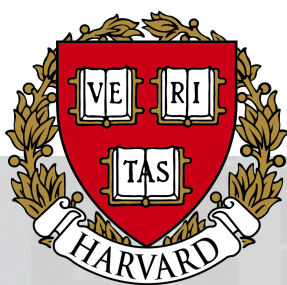
Nordita, July 17th, 2017  
Victor Buza  
Harvard University



# Outline

- Multicomponent analysis of CMB polarization maps using BICEP/Keck, Planck and WMAP data.
  - Framework Setup
  - Application to real data sets: BKP and BK14
- Morphing the Multicomponent Framework to a Fisher Framework
  - Comparing Fisher with BKP and BK14
- Performance-based Inflation Forecasts for BICEP Array & CMB-S4
  - What does the near future hold?
  - What does the CMB-polarization landscape look like in 2020+?





Colin Bischoff  
Victor Buza  
Cora Dvorkin  
John Kovac

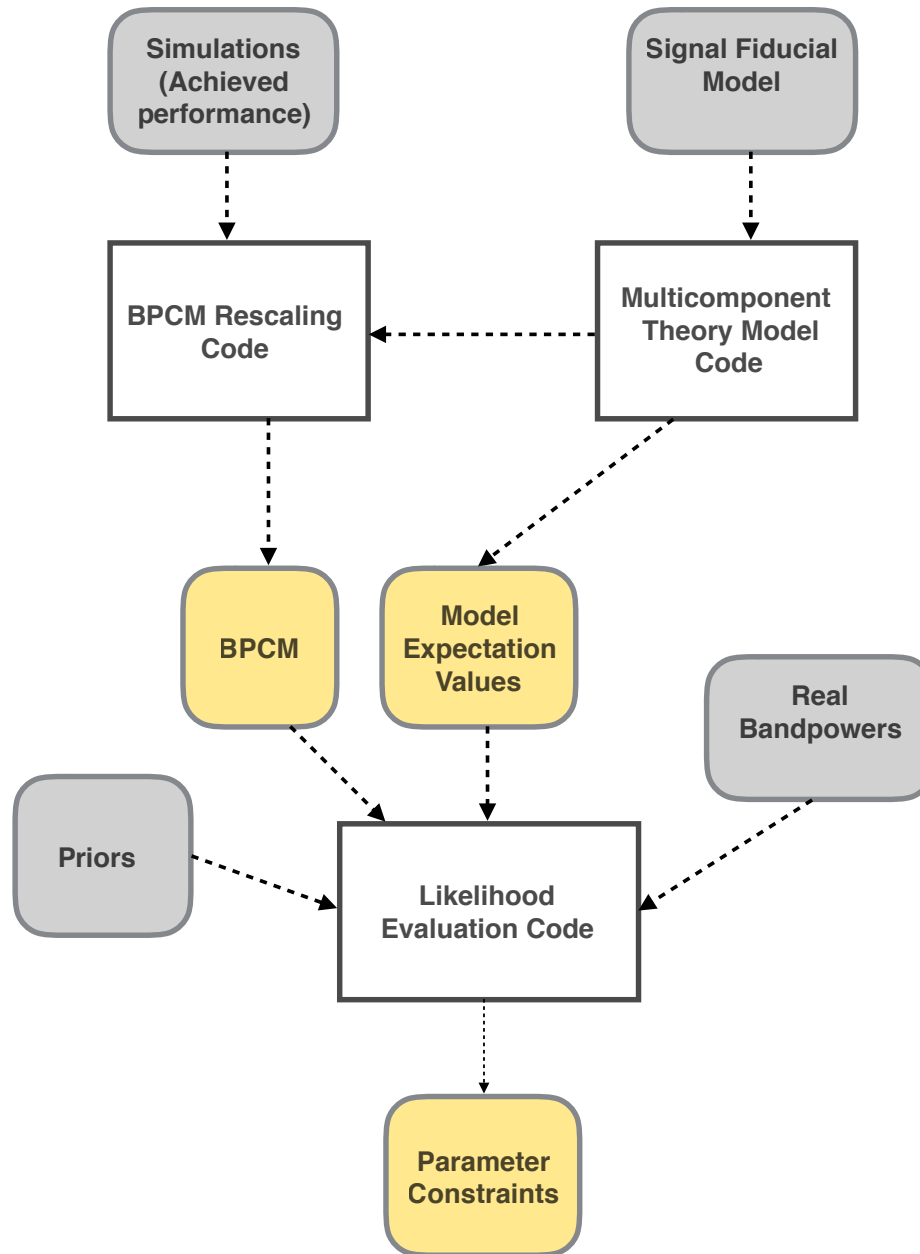


UNIVERSITY OF  
TORONTO



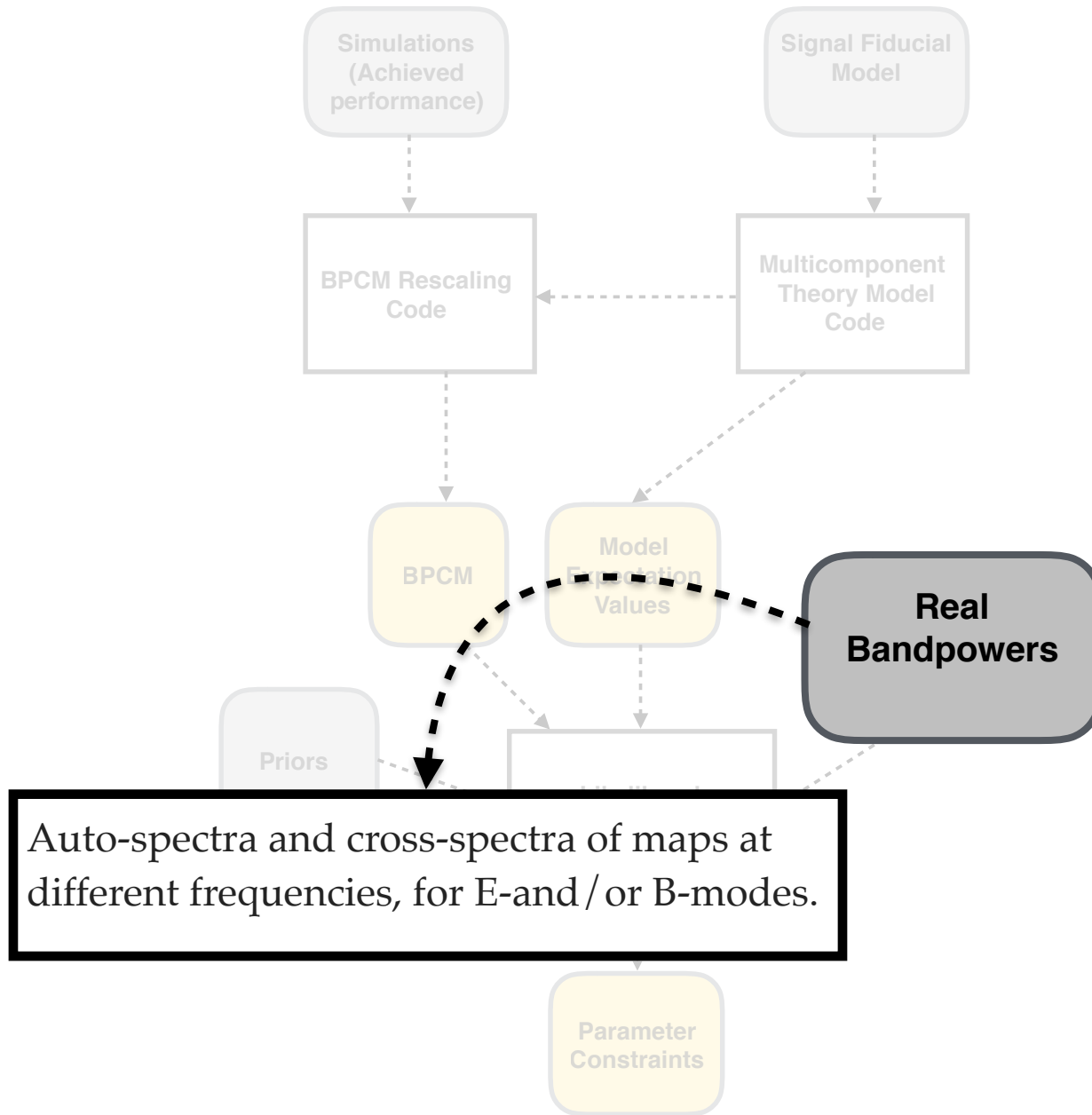


# Likelihood Framework Schematic





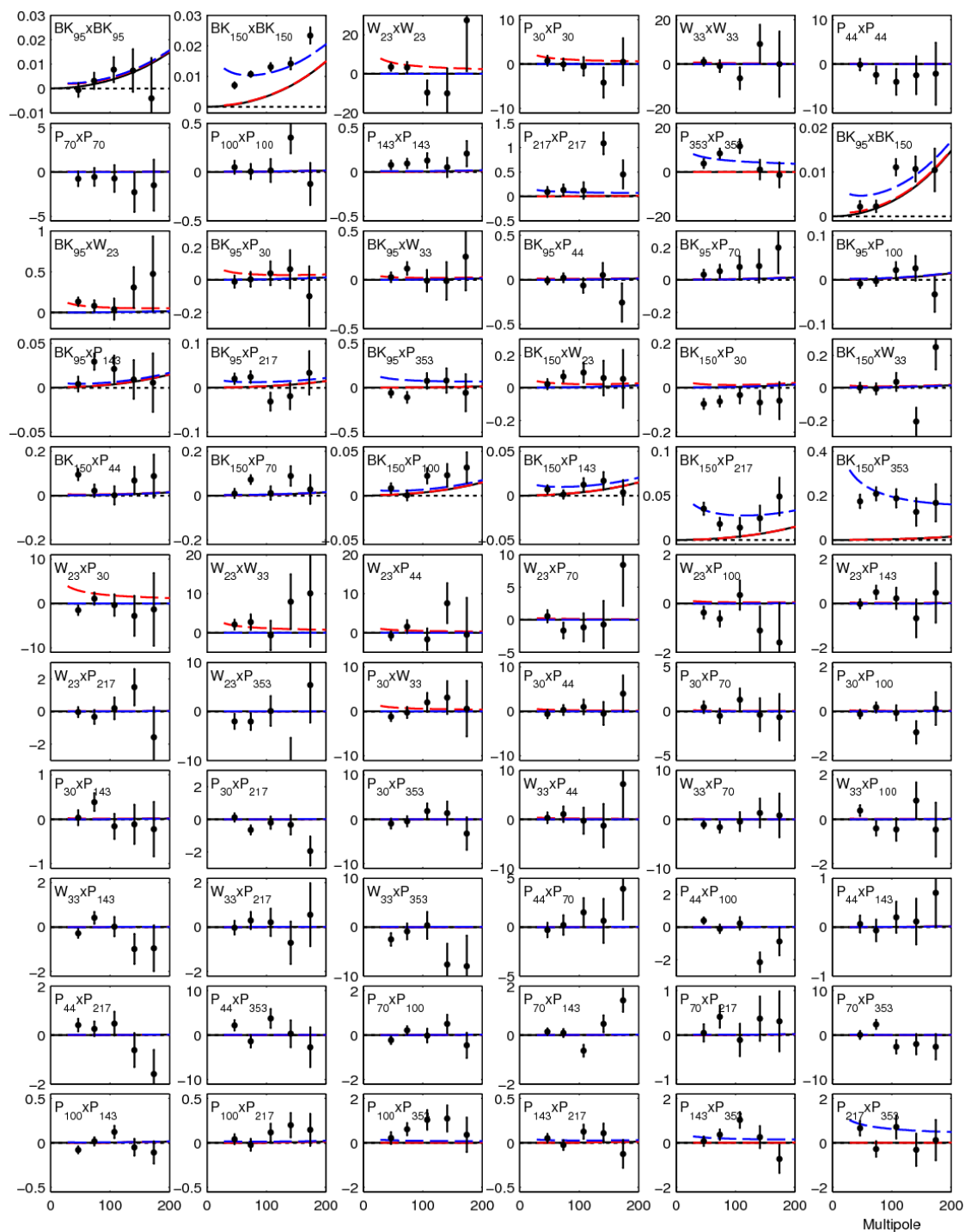
# Likelihood Framework Schematic



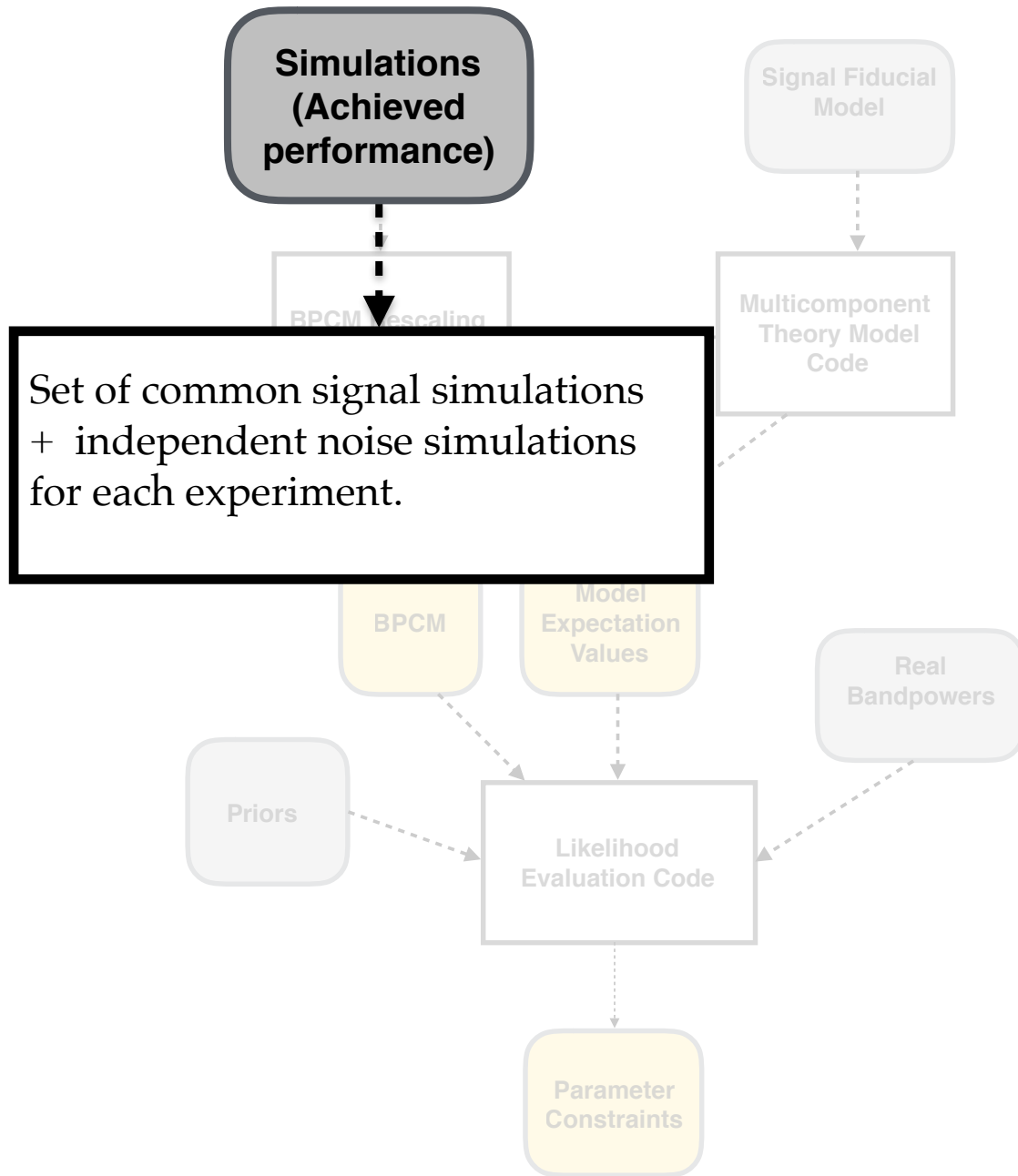


Take all possible  
auto- and cross  
spectra between  
BICEP/Keck,  
WMAP, and  
Planck bands

(ex. BK14)

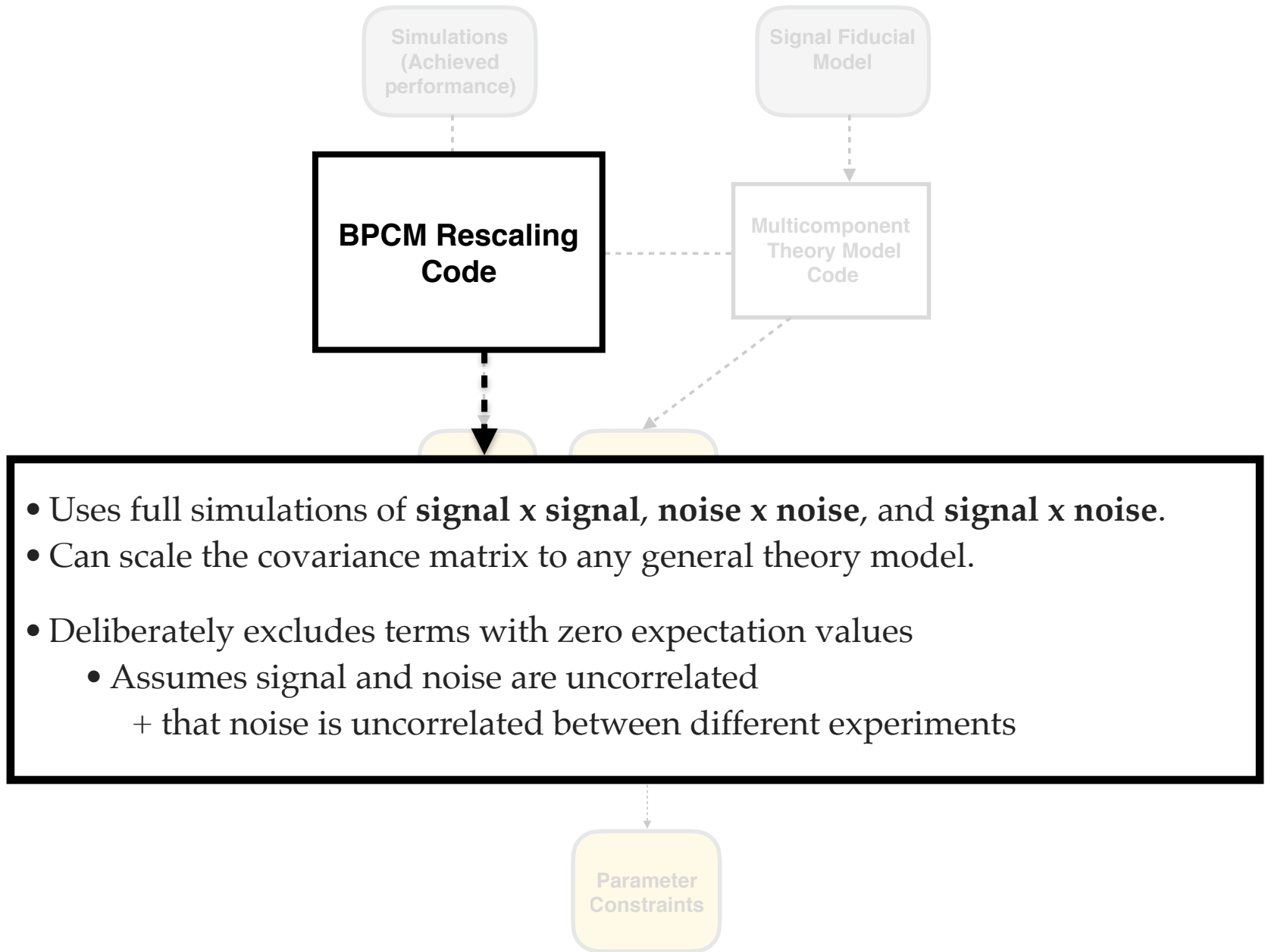


# Likelihood Framework Schematic

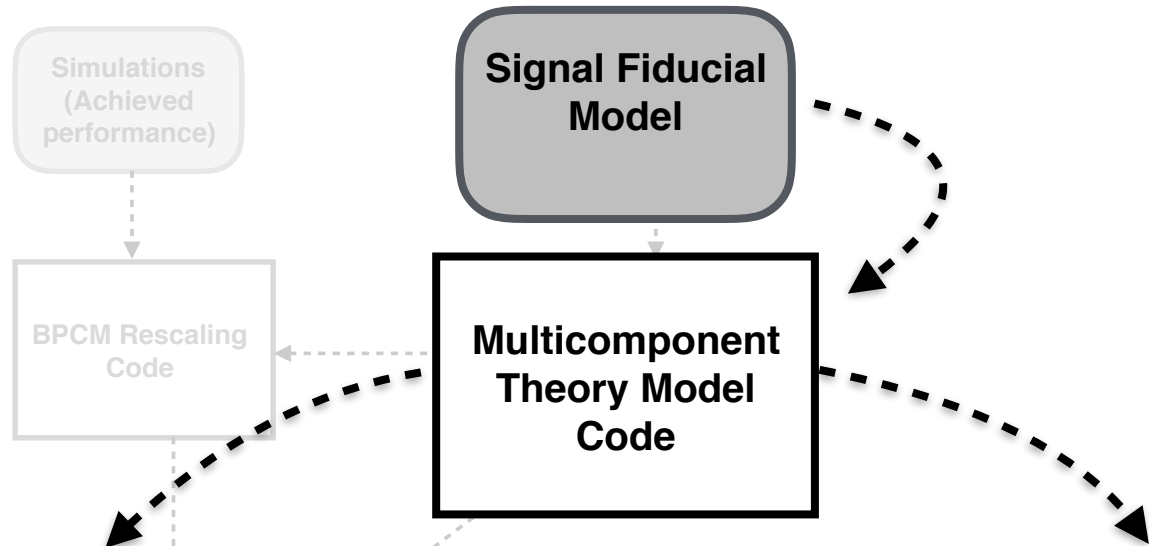




# Likelihood Framework Schematic



# Likelihood Framework Schematic



- Calculate likelihood for an arbitrary theory.
- We use a model containing the following four independent components:

1. CMB component: lensed- $\Lambda$ CDM + tensors
2. Dust component
3. Synchrotron component
4. Spatially correlated dust/synchrotron

## Model inputs:

- 13 parameters
- Information about the combination of experiments used

## Outputs:

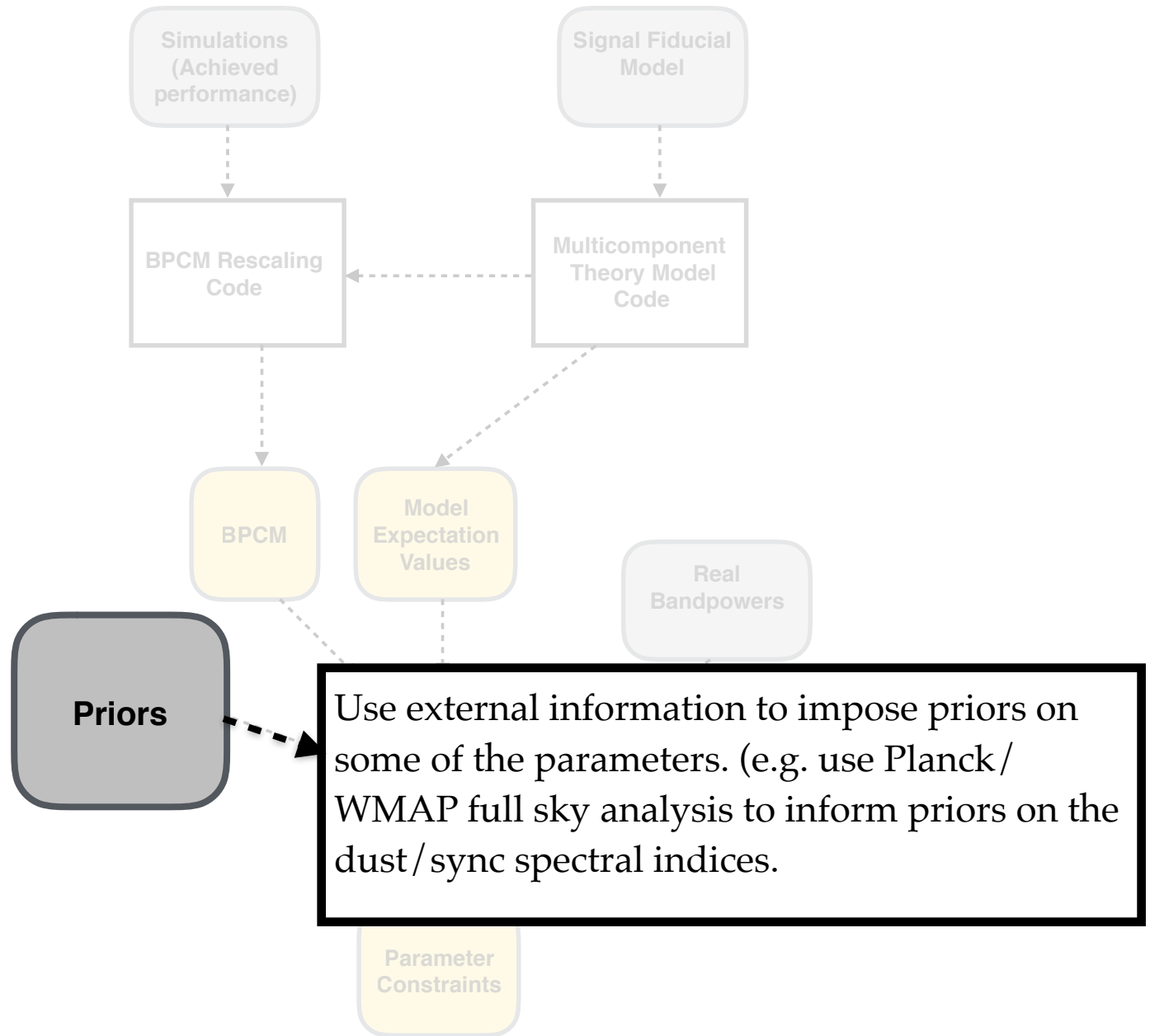
- Bandpower expectation values for all relevant spectra.

## The 13 model parameters are:

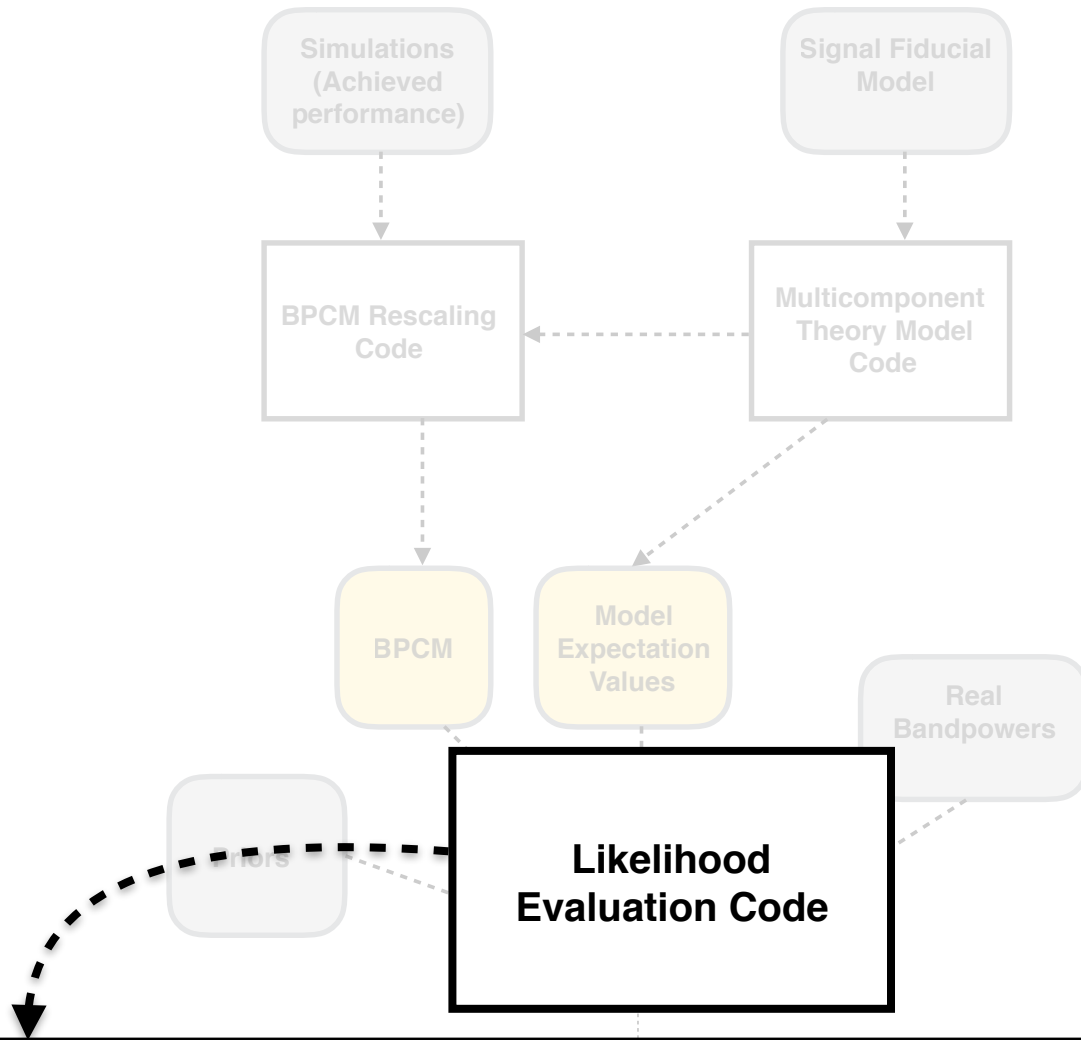
1.  $r$  — tensor-to-scalar ratio
2.  $A_L$  — lensing amplitude
3.  $A_d$  — dust amplitude, in  $\mu\text{K}_{\text{CMB}}^2$ , at 353 GHz and  $\ell = 80$ .
4.  $\beta_{\text{dust}}$  — dust spectral index
5.  $T_{\text{dust}}$  — dust greybody temperature
6.  $\alpha_{\text{dust}}$  — dust spatial spectral index
7.  $\delta_{\text{dust}}$  — dust frequency decorrelation
8. EE / BB ratio for dust
9.  $A_s$  — sync amplitude, in  $\mu\text{K}_{\text{CMB}}^2$ , at 23 GHz and  $\ell = 80$ .
10.  $\beta_{\text{sync}}$  — sync spectral index
11.  $\alpha_{\text{sync}}$  — sync spatial spectral index
12. EE / BB ratio for synchrotron
13.  $\epsilon$  — synchrotron–dust spatial correlation



# Likelihood Framework Schematic



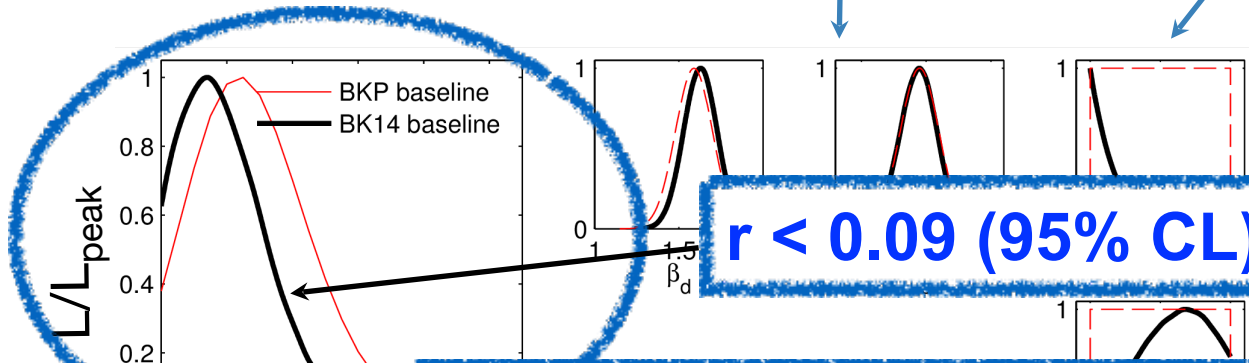
# Likelihood Framework Schematic



Use the Hamimeche-Lewis likelihood approximation (PRD 77, 103013; arXiv:0801.0554) to describe non-gaussian bandpower statistics.

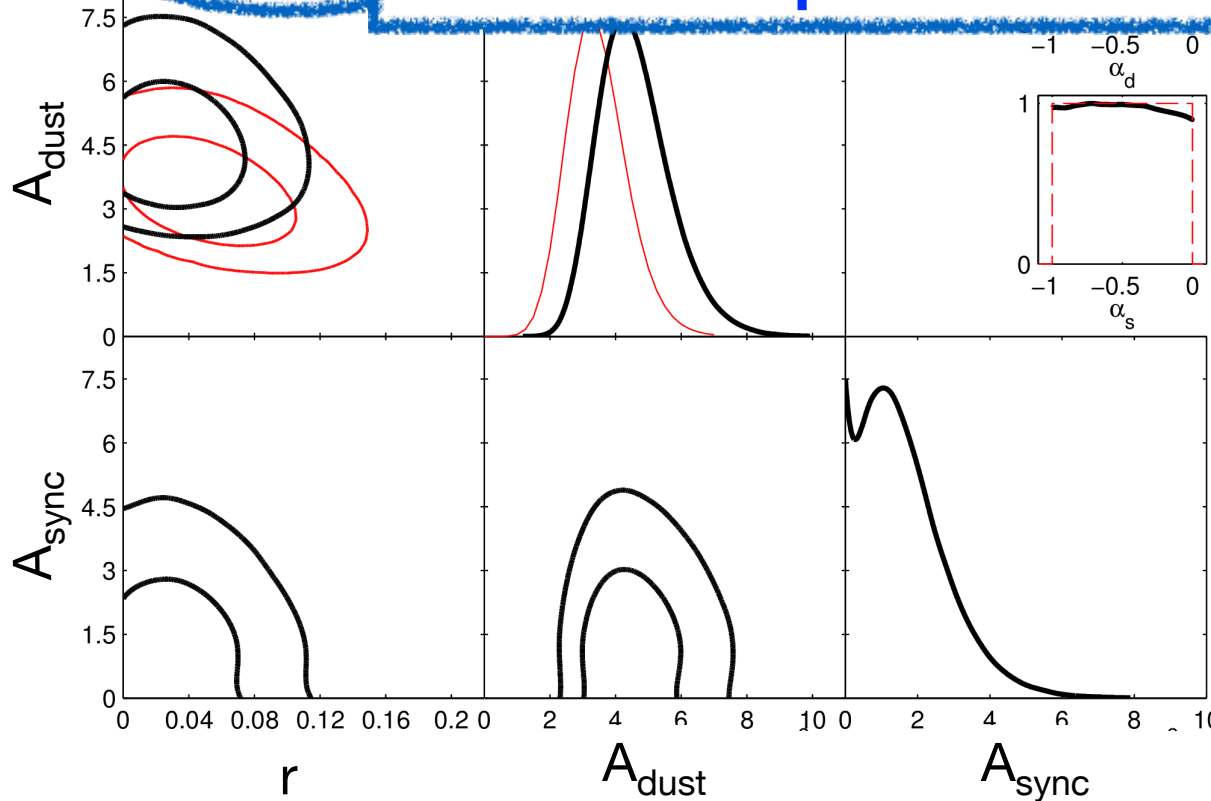
Put priors on the frequency spectral indices of dust & sync

Allow dust/sync correlation



**Beats temperature constraints**

dust vs.  $r$  →  
degeneracy lifted



Marginalize over  
generous ranges in  
spatial spectral  
indices



# Fisher Formalism

Given a Likelihood function of the form

$$\mathcal{L}(\theta; d) \sim \frac{1}{\sqrt{\det(\Sigma(\theta))}} \exp\left\{-\frac{1}{2}(d - \mu(\theta))^T \Sigma(\theta)^{-1} (d - \mu(\theta))\right\}$$

$d$  is the data,

$\theta$  are the theory parameters,

$\mu(\theta)$  are the expectation values given the parameters

$\Sigma(\theta)$  is the band-power covariance matrix.

We can calculate the average of the log-likelihood curvature (the Fisher Matrix):

$$F_{ij} = - \left\langle \frac{\partial^2 \log(\mathcal{L}(\theta; d))}{\partial \theta_i \partial \theta_j} \right\rangle$$

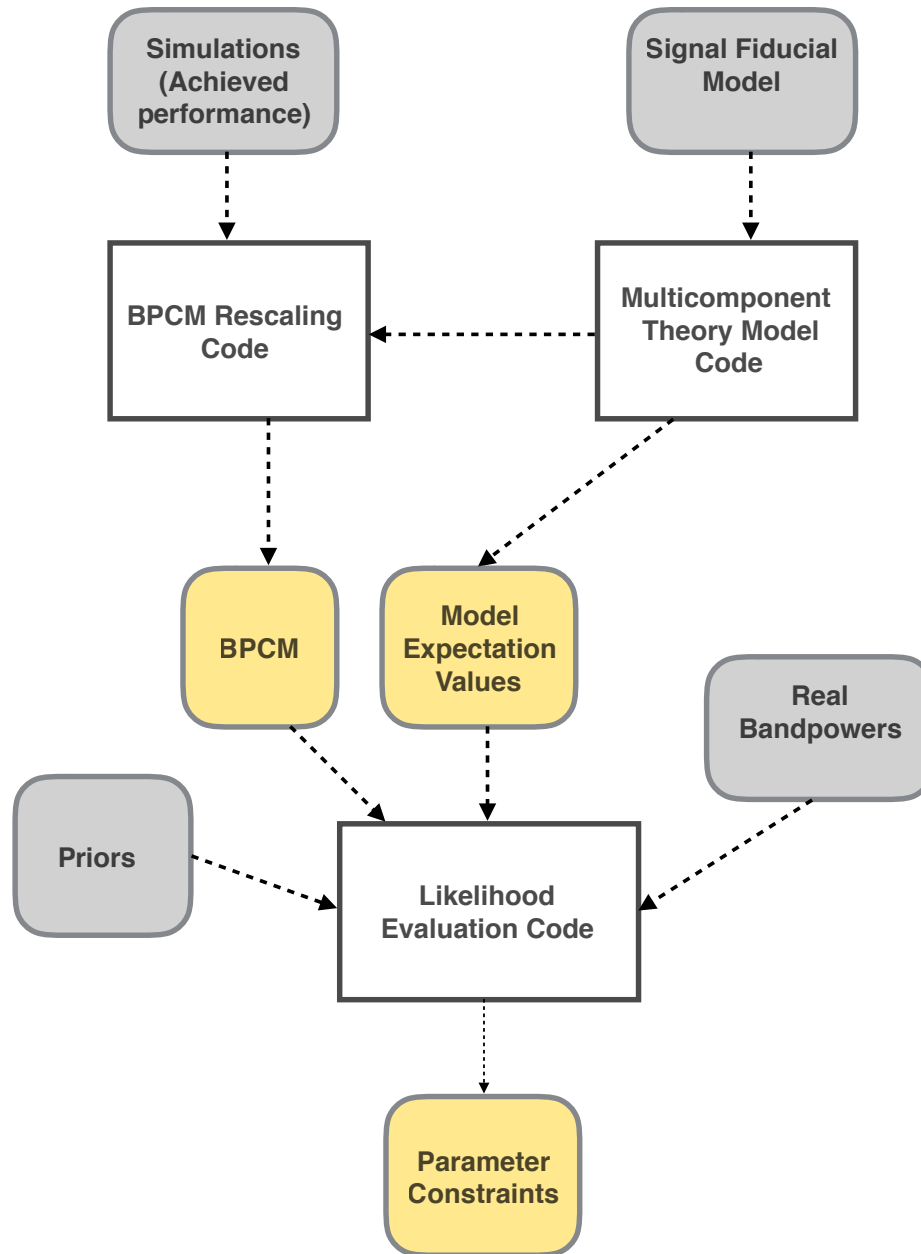
Which can be rewritten as:

$$F_{ij} = \frac{\partial \mu^T}{\partial \theta_i} \Sigma^{-1} \frac{\partial \mu}{\partial \theta_j} + \frac{1}{2} \text{Tr}(\Sigma^{-1} \frac{\partial \Sigma}{\partial \theta_i} \Sigma^{-1} \frac{\partial \Sigma}{\partial \theta_j})$$

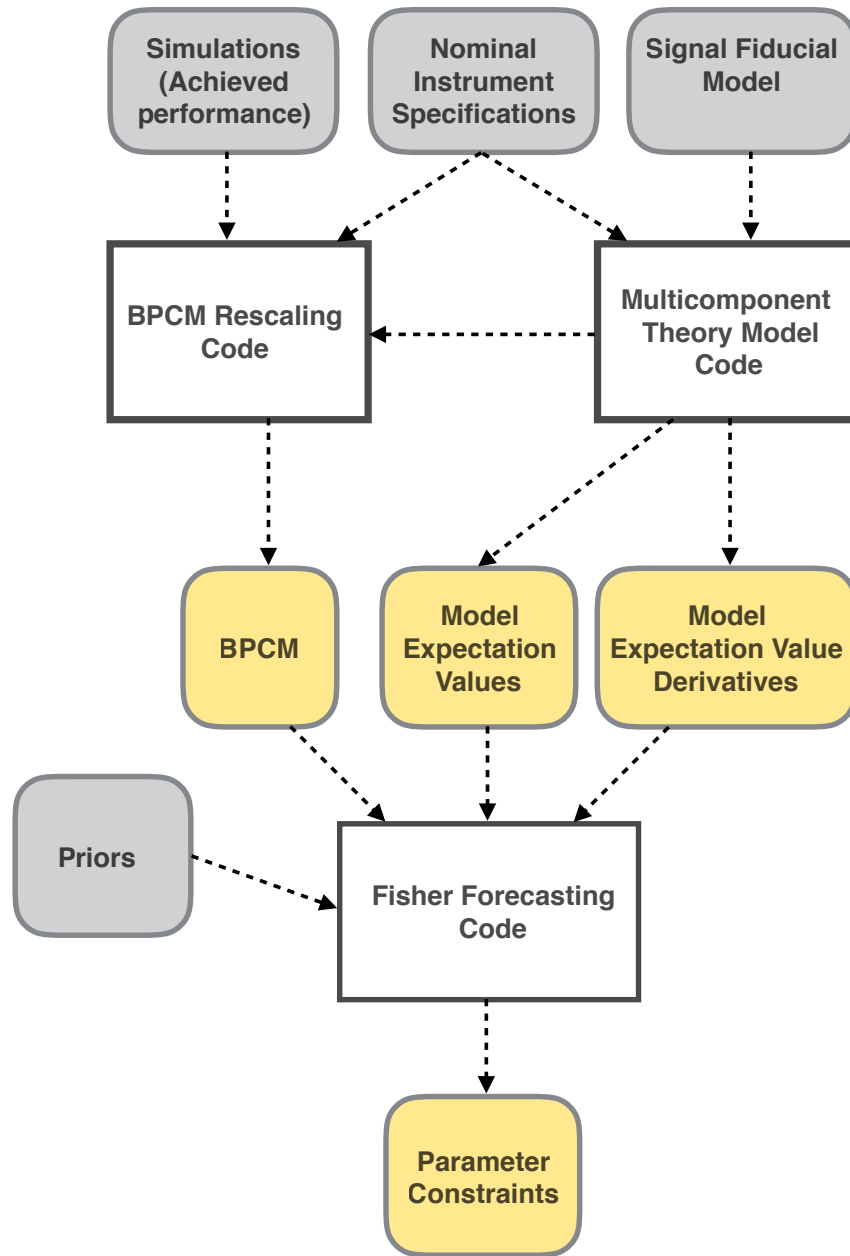
Very useful for projecting into the future!

Offers parameter constraints simply with the knowledge of  $\mu$  and  $\Sigma$ .

# Likelihood Framework Schematic

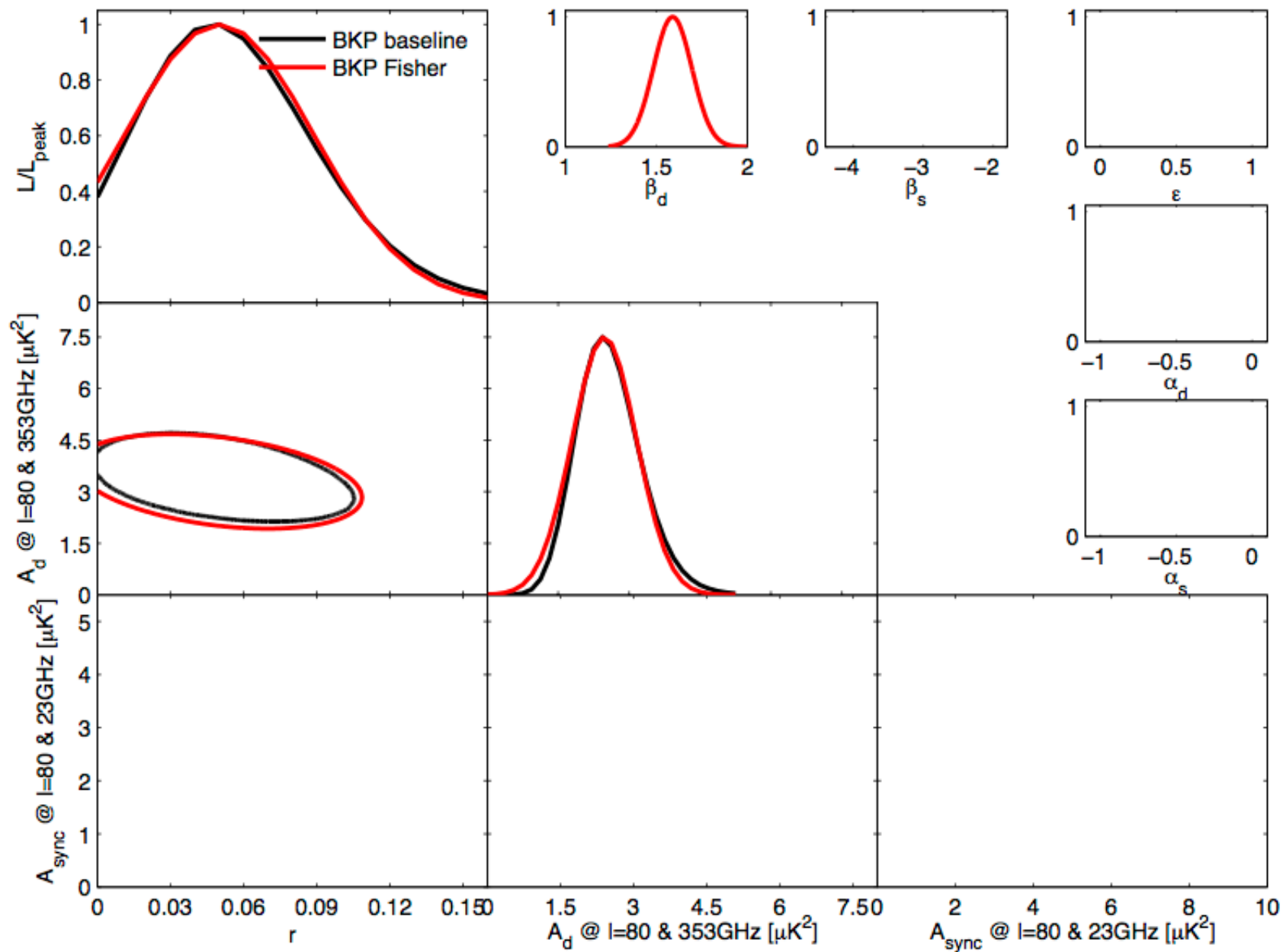


# Fisher Framework Schematic

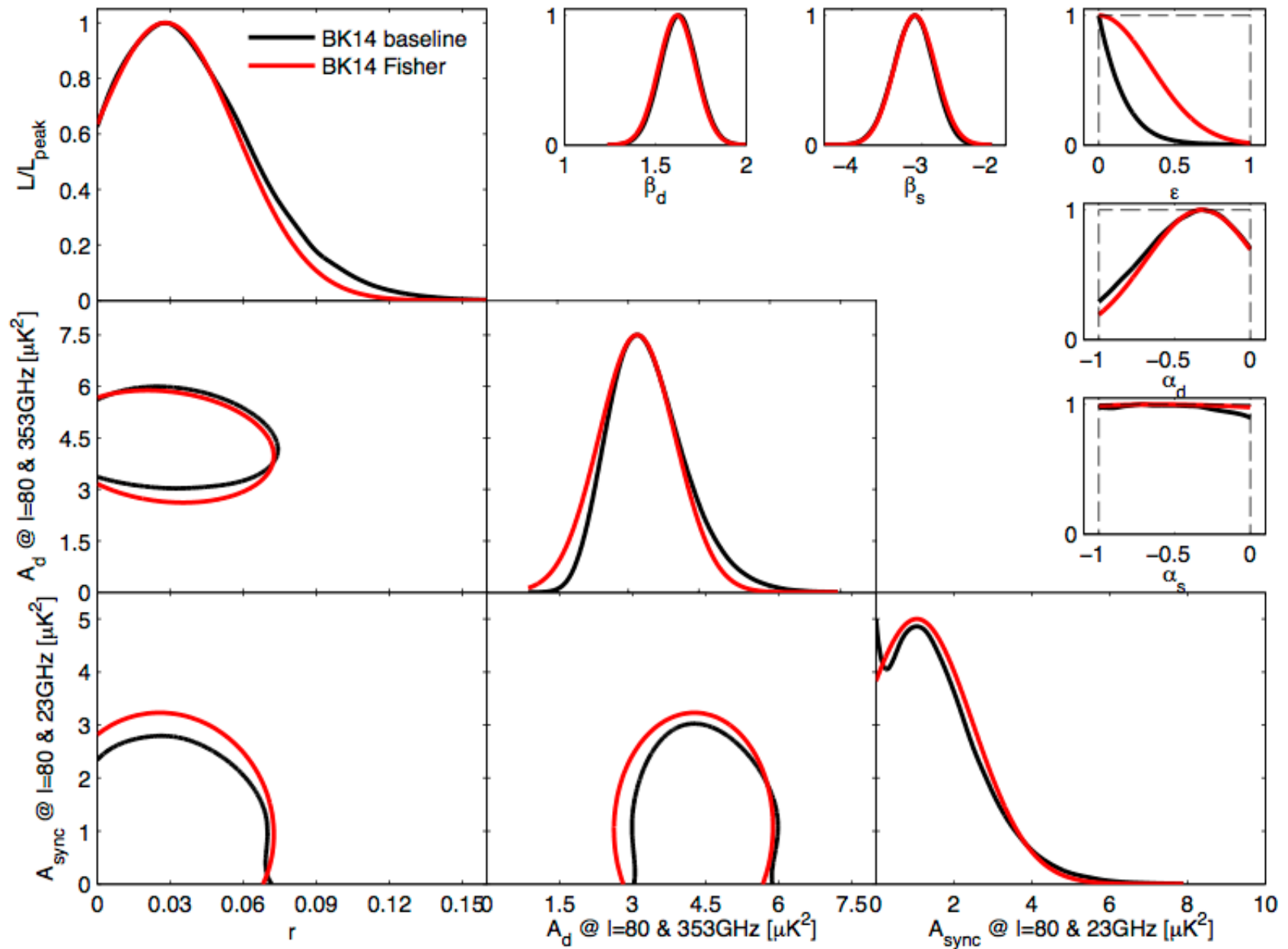




# Comparing Fisher vs real BKP constraints

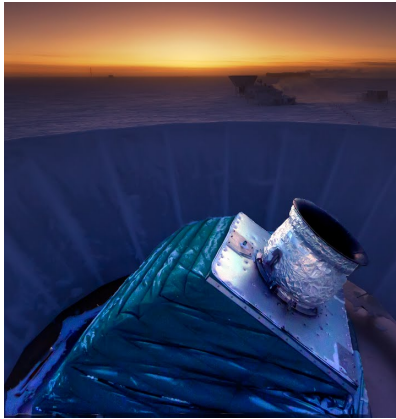


# Comparing Fisher vs real BK14 constraints

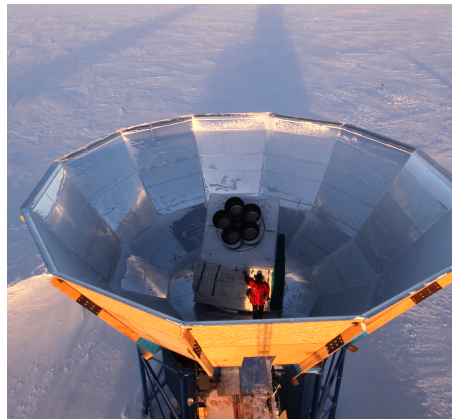


## Stage 2

**BICEP2**  
(2010-2012)

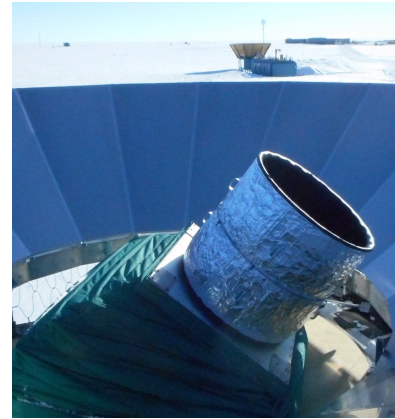


**Keck Array**  
(2012-2017)

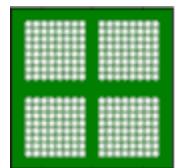
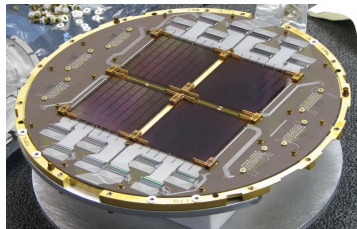
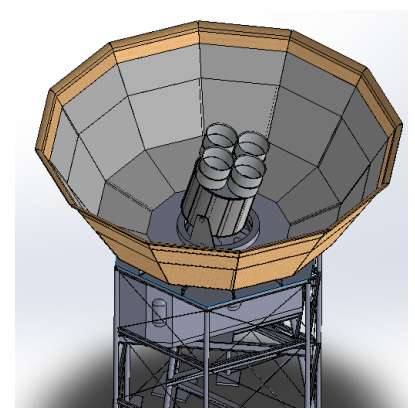


## Stage 3

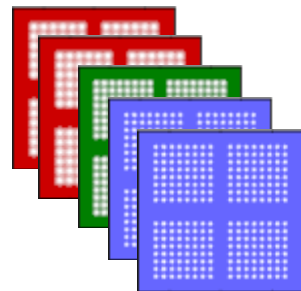
**BICEP3**  
(2015-)



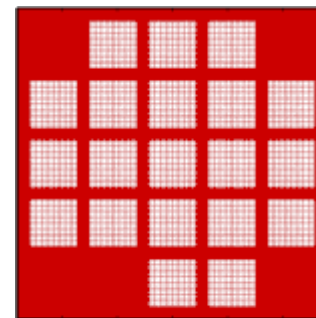
**BICEP Array**  
(2018-)



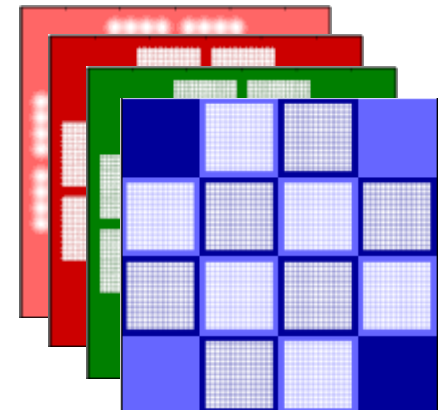
-5 0 5  
Degrees on sky



-5 0 5  
Degrees on sky



-10 -5 0 5 10  
Degrees on sky



-10 -5 0 5 10  
Degrees on sky

Telescope and Mount

Focal Plane

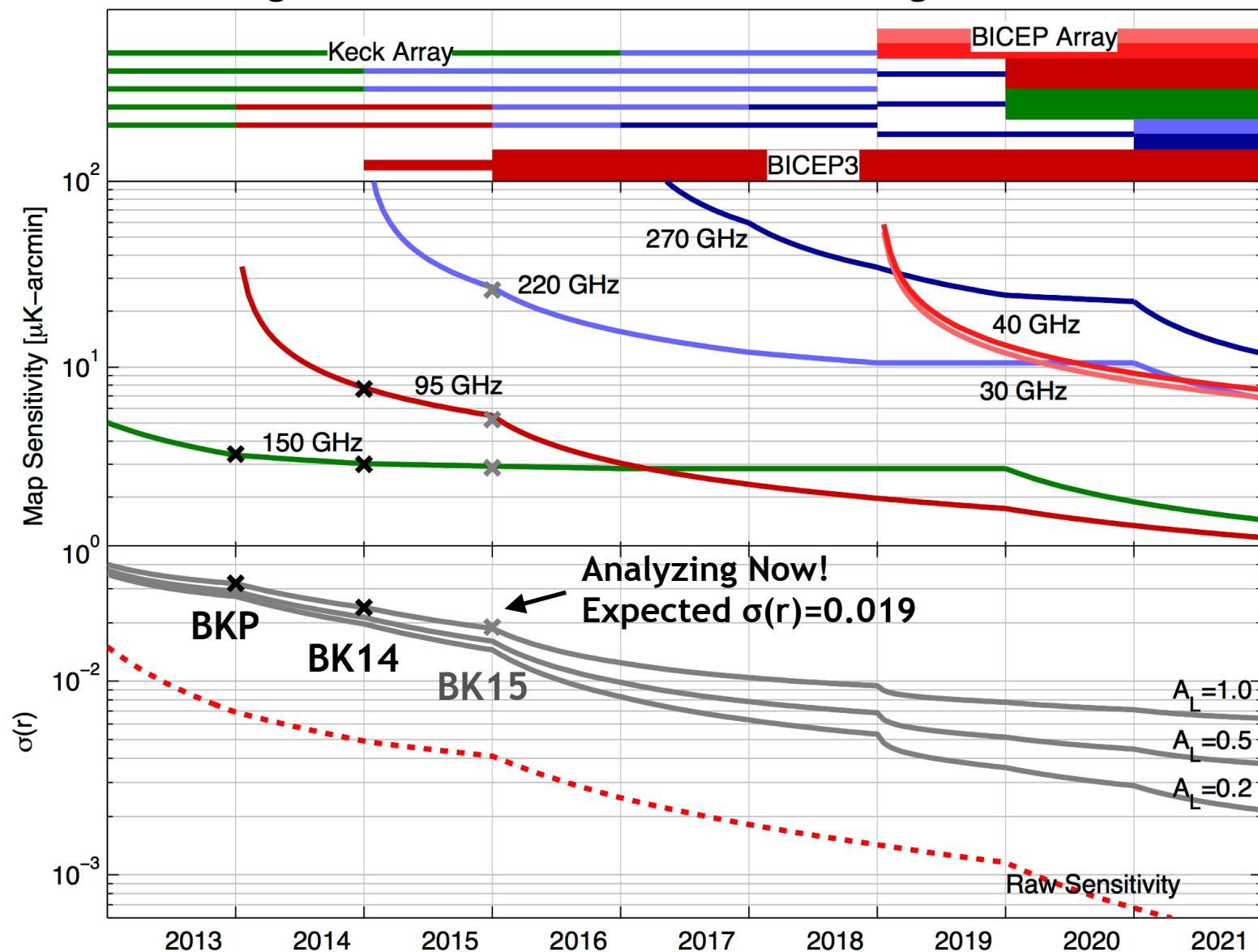
Beams on Sky



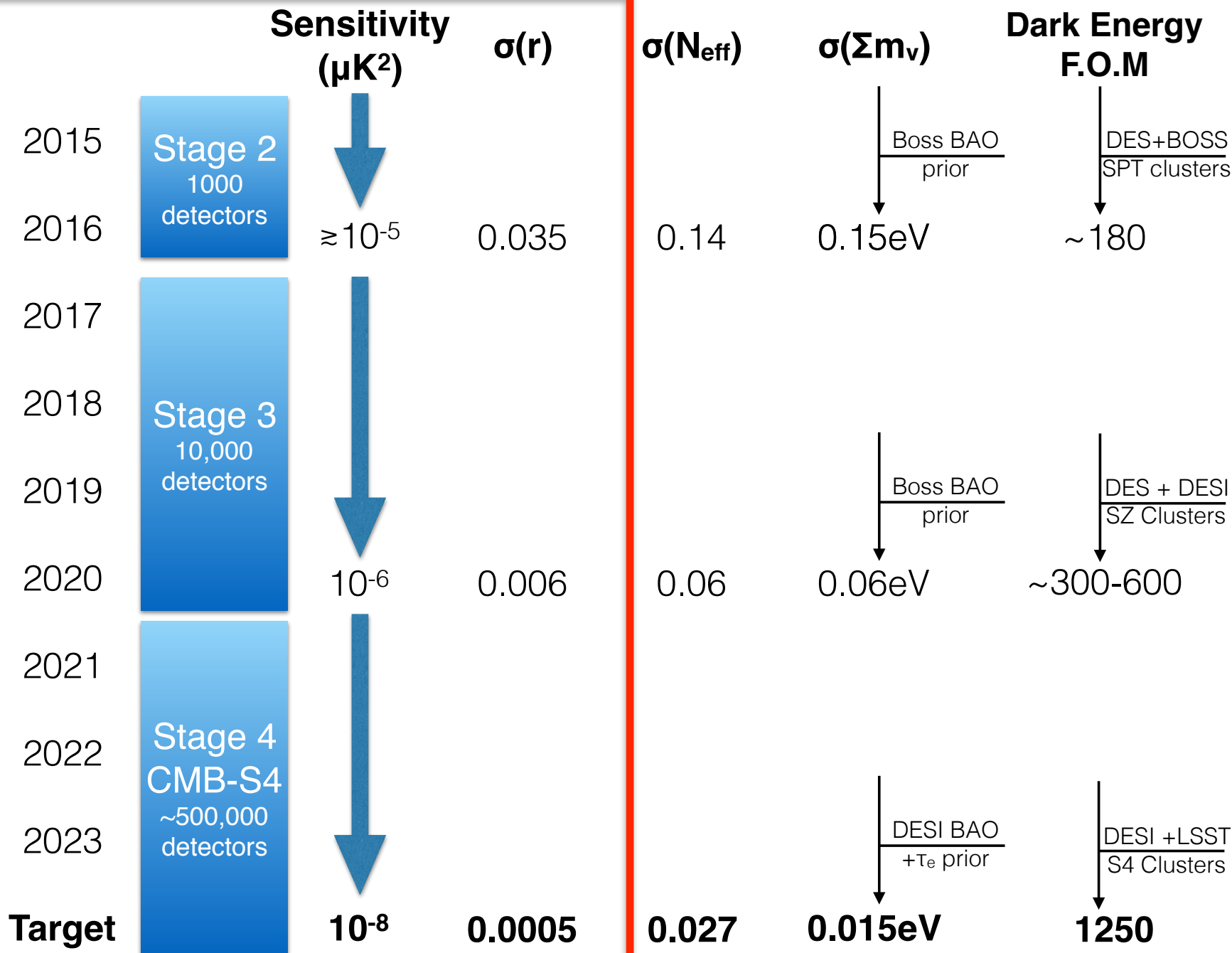
# BICEP Array + SPT3G as a Pathfinder for the CMB-S4 Deep Survey

Stage 2

Stage 3



- 6 band foreground control and  $\sigma(r)<0.003$  by 2021
- Scaling from achieved published analyses (i.e. all real-world performance hits included)



# CMB-S4 Science Book Baseline

- Experiment Selection:
  - CMB-S4: {30, 40, 85, 95, 145, 155, 215, 270} GHz
  - + Planck: {30 - 353} GHz
  - + WMAP: {23, 33} GHz



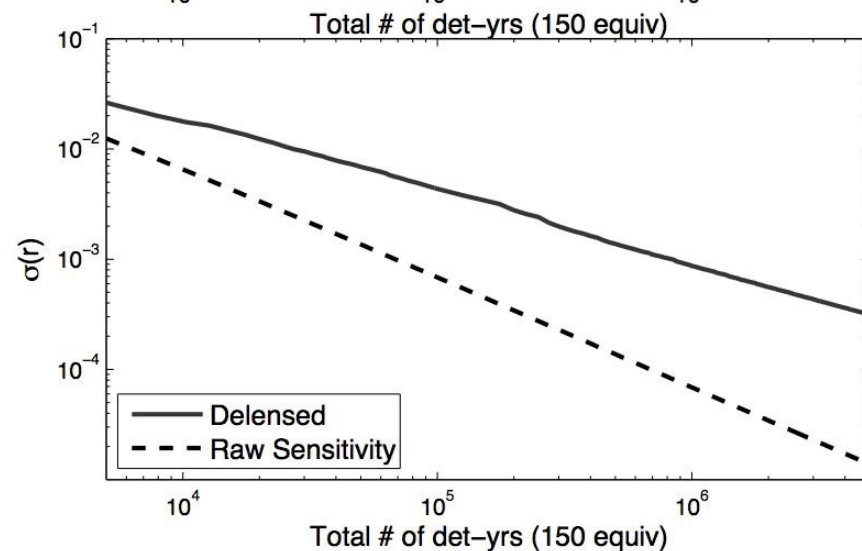
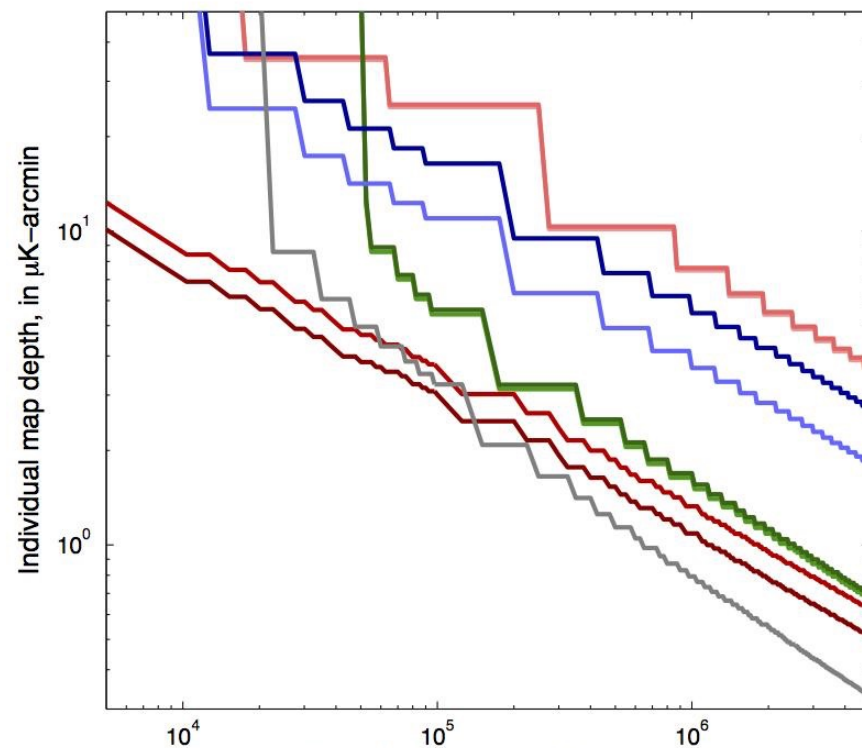
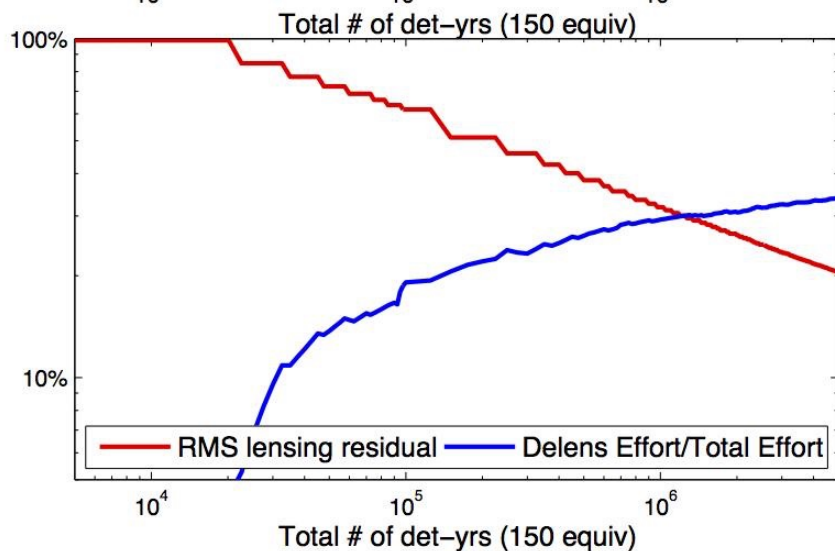
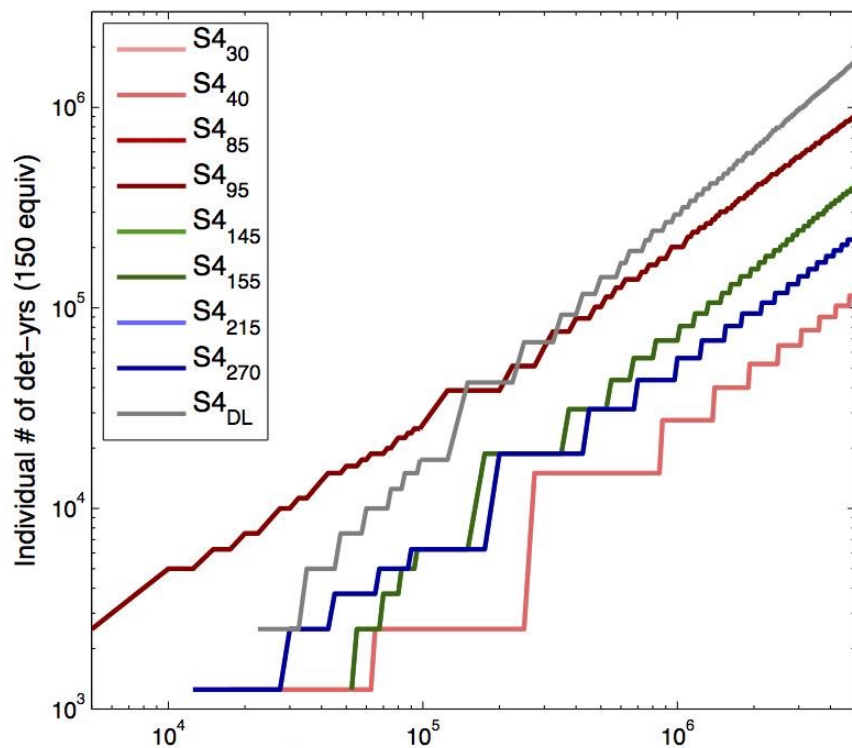
# CMB-S4 Science Book Baseline

- Experiment Selection:
  - CMB-S4: {30, 40, 85, 95, 145, 155, 215, 270} GHz
  - + Planck: {30 - 353} GHz
  - + WMAP: {23, 33} GHz
- Heterogeneous survey with (500k detectors x 4 yrs = 2M det-yrs):
  - 1/2 effort (1M det-yrs) for primordial B-modes
    - small-aperture ( $\sim 1$ m)
    - on 3% of the sky
    - fraction of effort targeting arcminute-scale information (used to remove lensing B-modes)
  - 1/2 effort (1M det-yrs) for CMB-lensing, Neutrino Science, Dark Energy, etc.
    - medium-to-large-aperture telescopes
    - on 40% of the sky

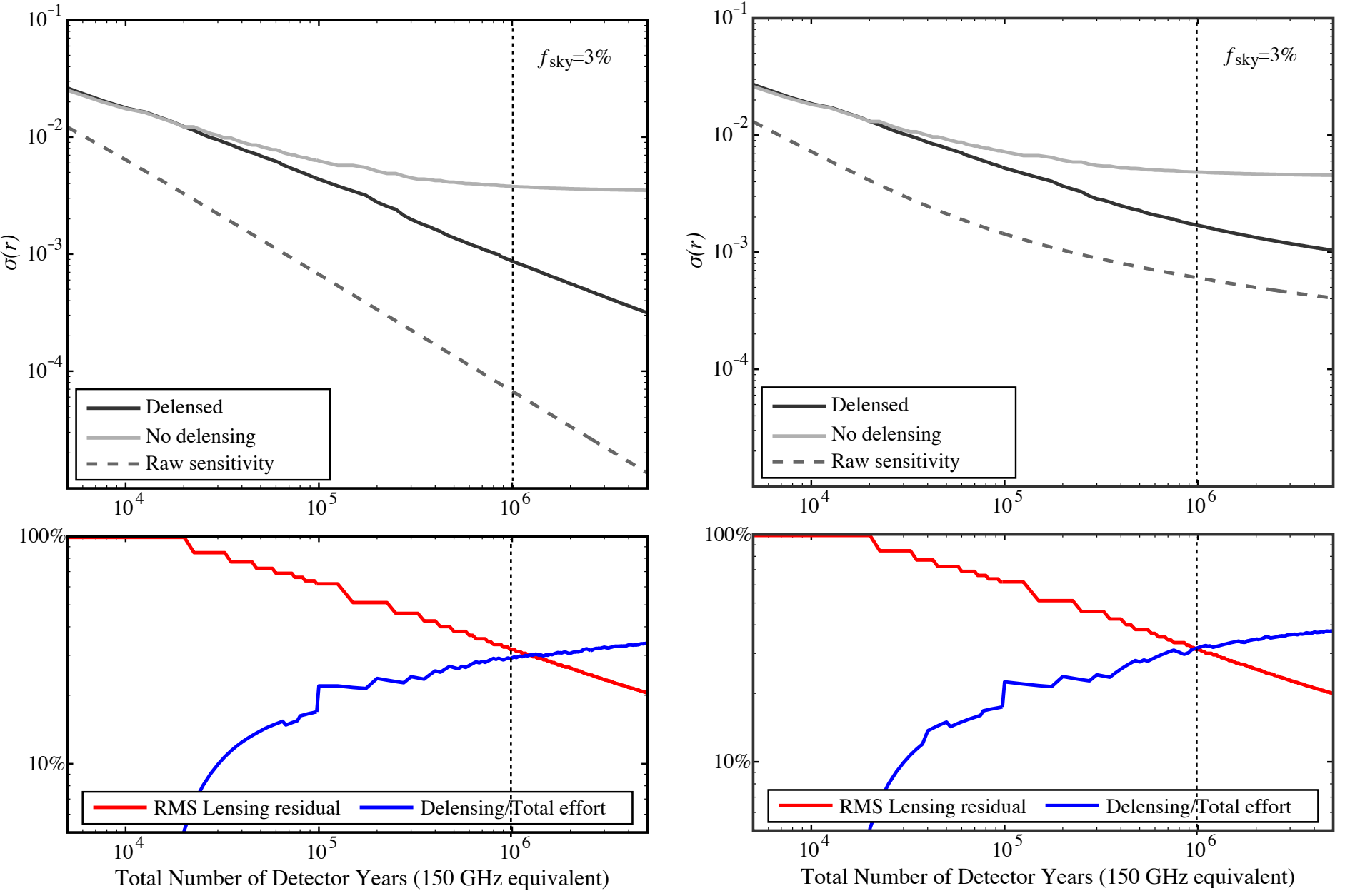
# CMB-S4 Science Book Baseline

- Experiment Selection:
  - CMB-S4: {30, 40, 85, 95, 145, 155, 215, 270} GHz
  - + Planck: {30 - 353} GHz
  - + WMAP: {23, 33} GHz
- Heterogeneous survey with (500k detectors x 4 yrs = 2M det-yrs):
  - 1/2 effort (1M det-yrs) for primordial B-modes
    - small-aperture ( $\sim 1$ m)
    - on 3% of the sky
    - fraction of effort targeting arcminute-scale information (used to remove lensing B-modes)
  - 1/2 effort (1M det-yrs) for CMB-lensing, Neutrino Science, Dark Energy, etc.
    - medium-to-large-aperture telescopes
    - on 40% of the sky
- Optimize effort, for various sky fractions, with degree-scale component separation and arcminute-scale delensing

# Optimization example, assuming $r = 0$ and fsky=3%.

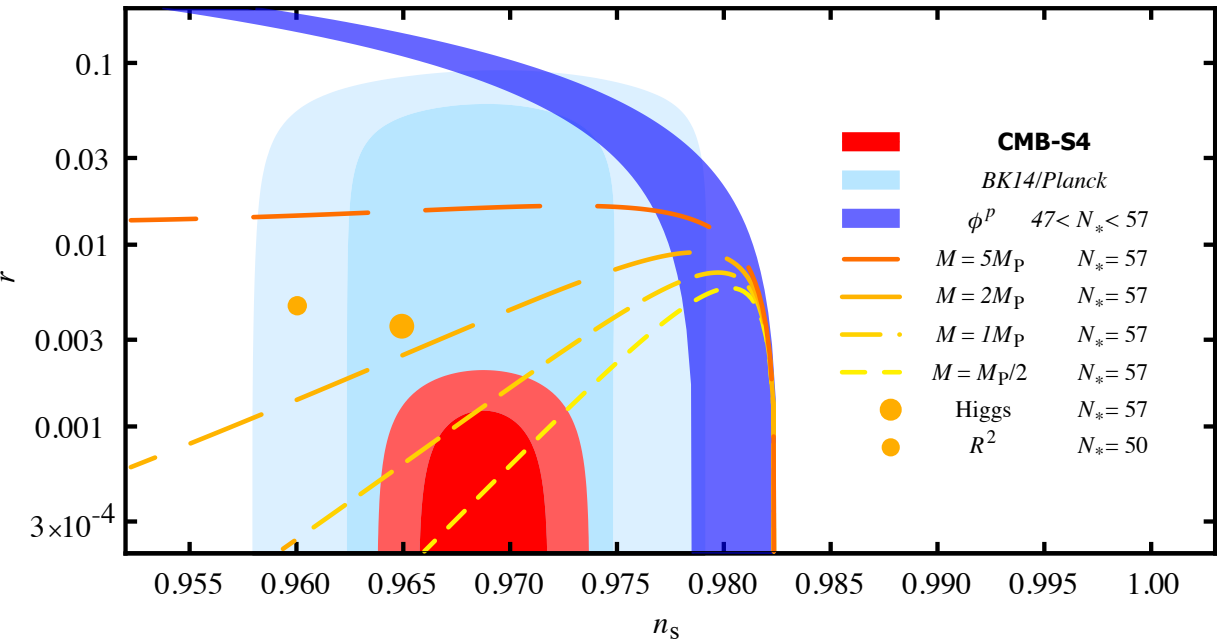


Uncertainty forecasts on  $r$ , assuming  $r = 0$  (left panel) and  $r = 0.01$  (right panel).

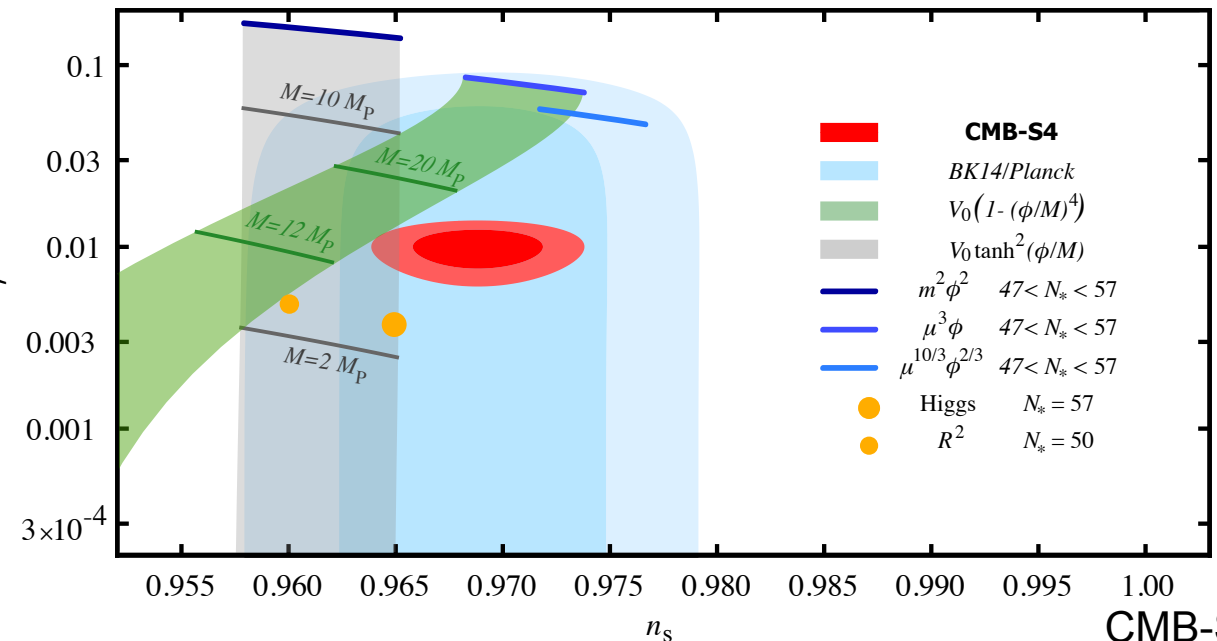




Forecasted CMB-S4  $r$ - $n_s$  constraints, assuming  $r = 0$  (top panel) and  $r = 0.01$  (bottom panel).



- In the absence of a detection,** CMB-S4 would rule out or disfavor all models that naturally explain the observed value of the scalar spectral index (in the sense that  $ns(N)-1 \sim 1/N$ ) and in which the characteristic scale in field space exceeds the Planck scale.



- A detection of primordial B modes** with CMB-S4 would provide evidence that the theory of quantum gravity must accommodate a Planckian field range for the inflaton.

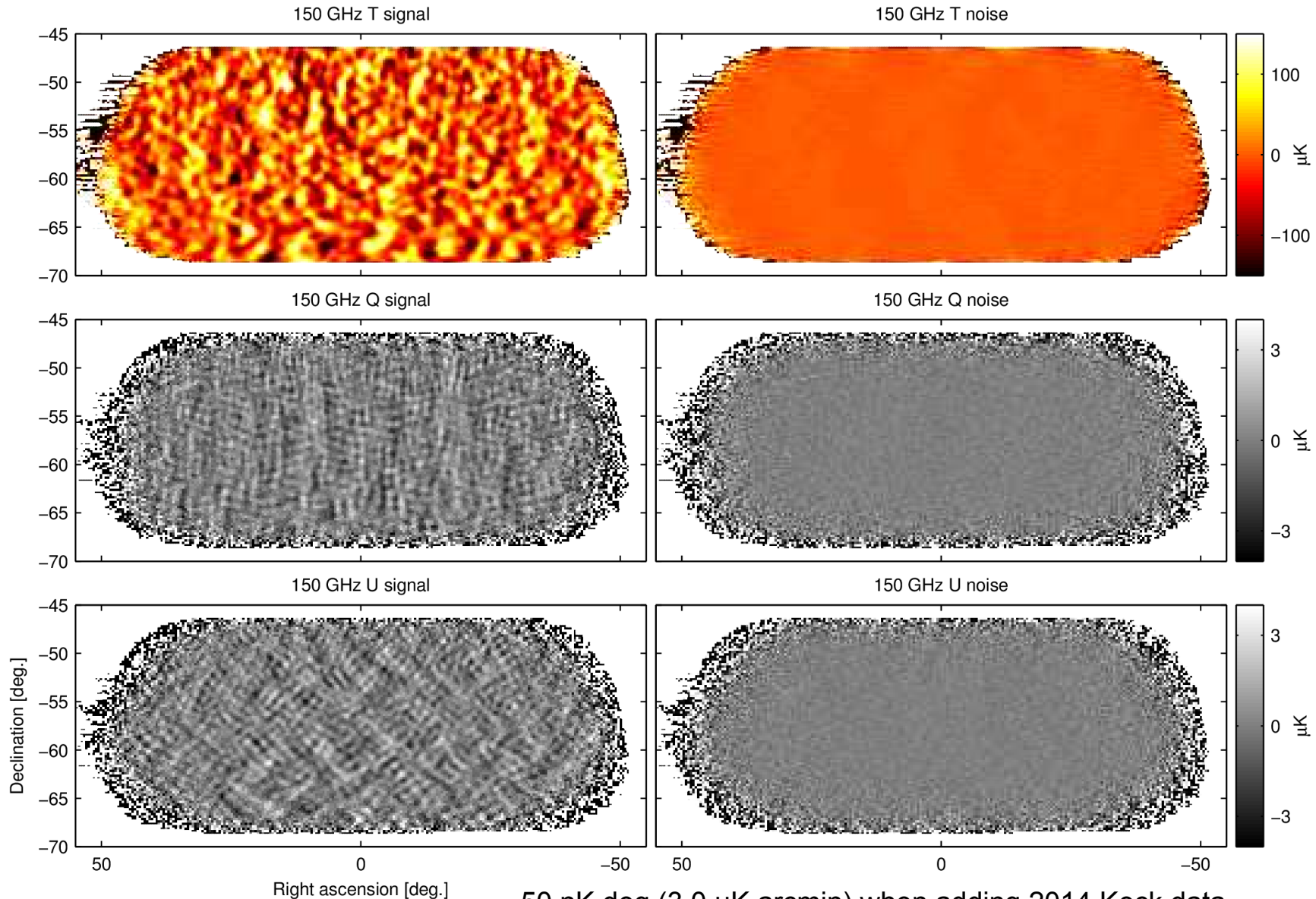
# Conclusions

- We use on-sky current achieved performances from various CMB experiments to make robust forecasts for future CMB-polarization endeavors.
  - Used and validated in the context of optimizing various survey configurations.
  - Used to calculate inflation projections for the next generation experiment in the BICEP/Keck series, and for CMB-S4.
- 
- Projecting six band foreground control and  $\sigma(r) < 0.003$  by 2021 with BICEP-Array
- 
- CMB-S4 brings us to  $\sigma(r) < 0.0007$  allowing us to significantly narrow down the space of possible inflationary models.

**Thank You!**

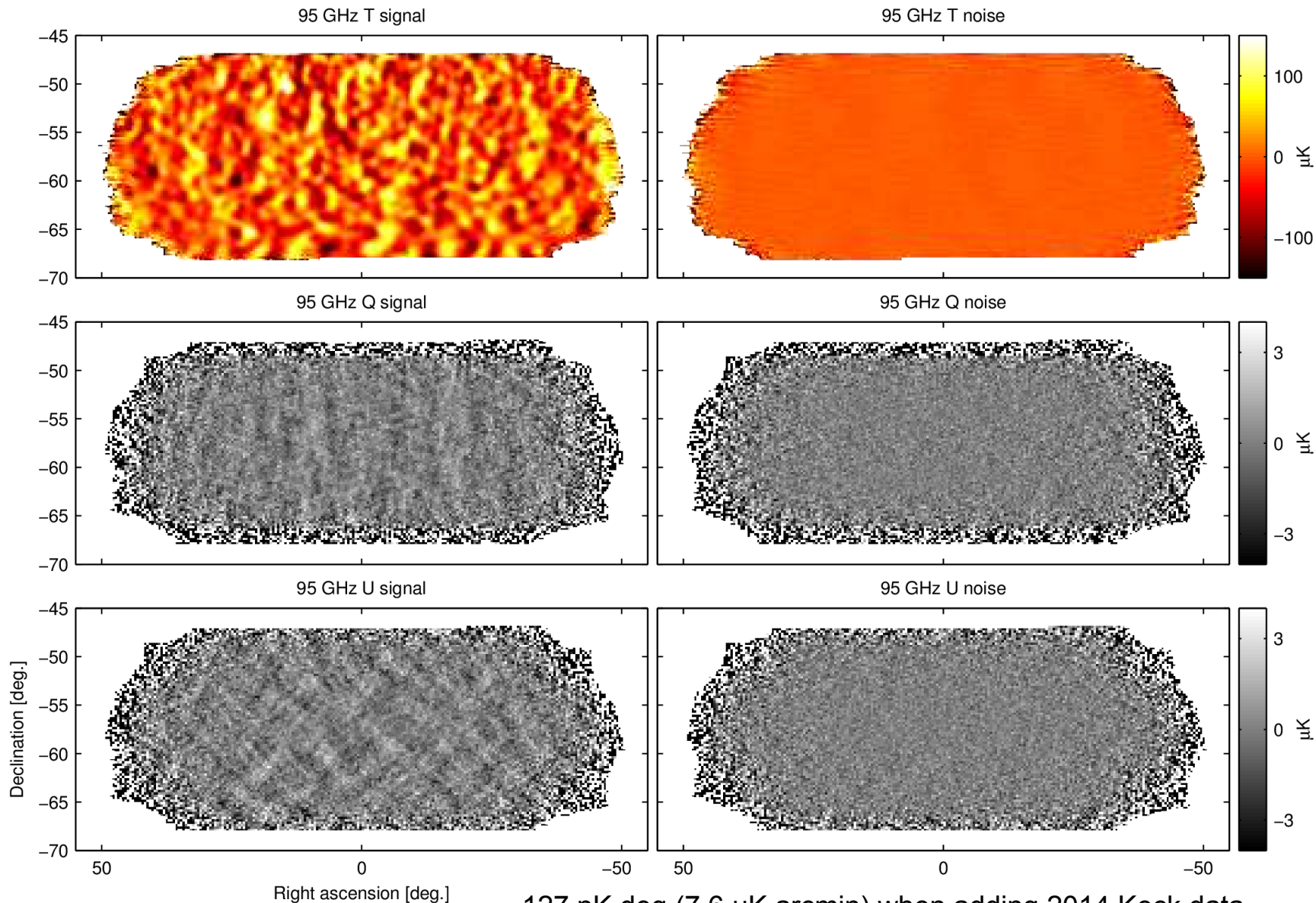
Extra Slides

# T/Q/U 150 GHz maps including Keck 2014 data

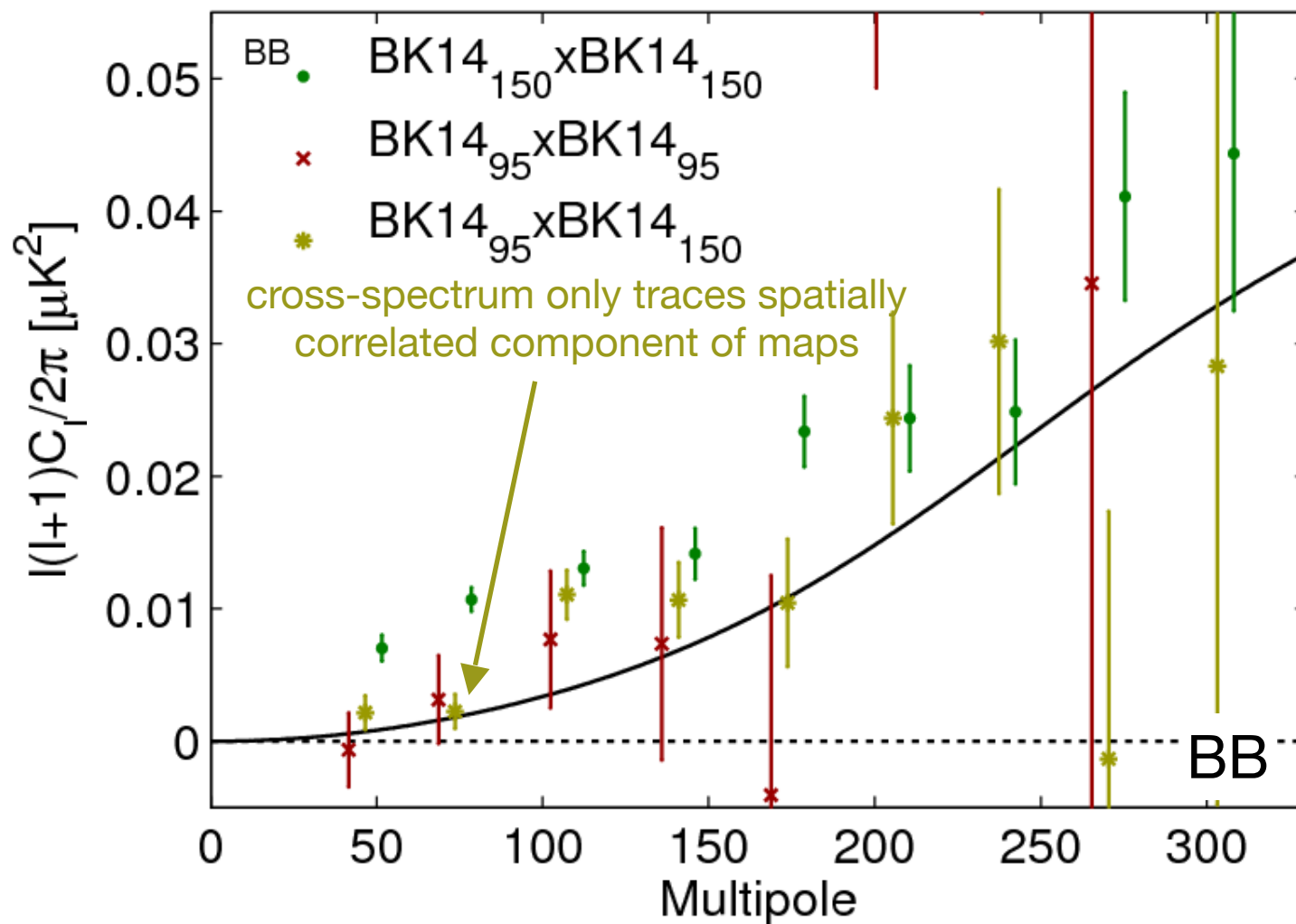




# Keck 2014 T/Q/U 95 GHz maps



# BICEP2 + Keck BB auto and cross-spectra



Polarized galactic  
**synchrotron** dominates  
at low frequencies

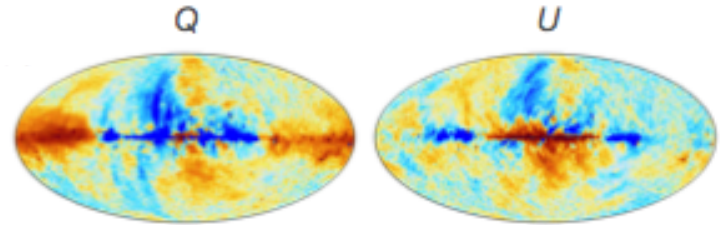


Planck then provided  
polarized maps at 7  
frequencies  
(two more from WMAP  
at low frequencies  
already existed)

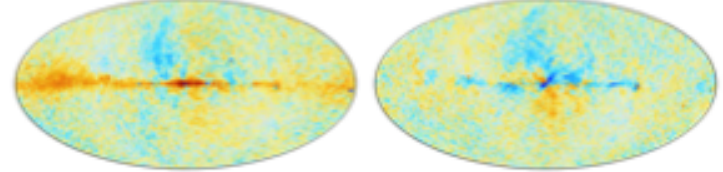
Polarized thermal emission  
(~20K) from galactic **dust**  
aligned in magnetic fields  
dominates at high frequencies



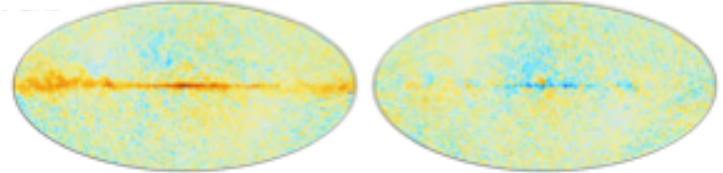
30  
GHz



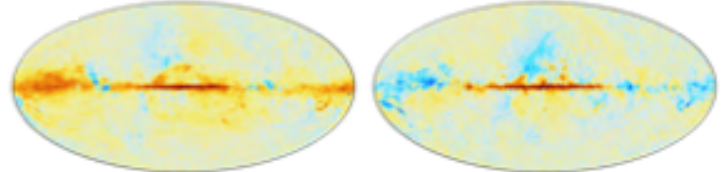
44  
GHz



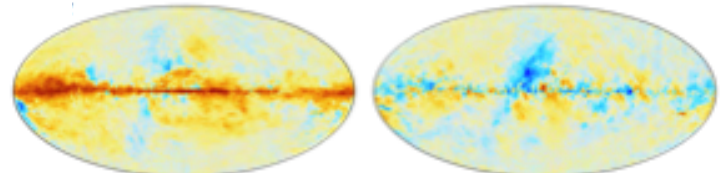
70  
GHz



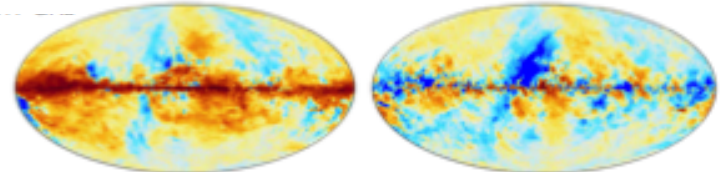
100  
GHz



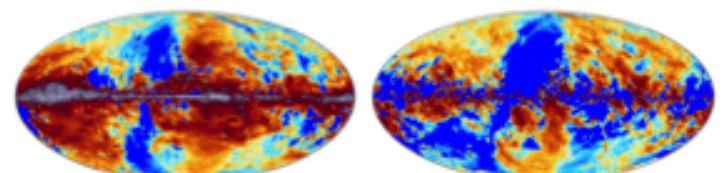
143  
GHz



217  
GHz



353  
GHz

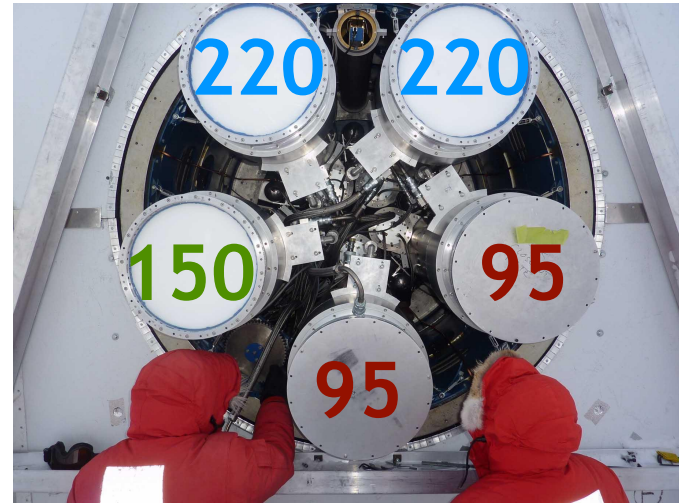




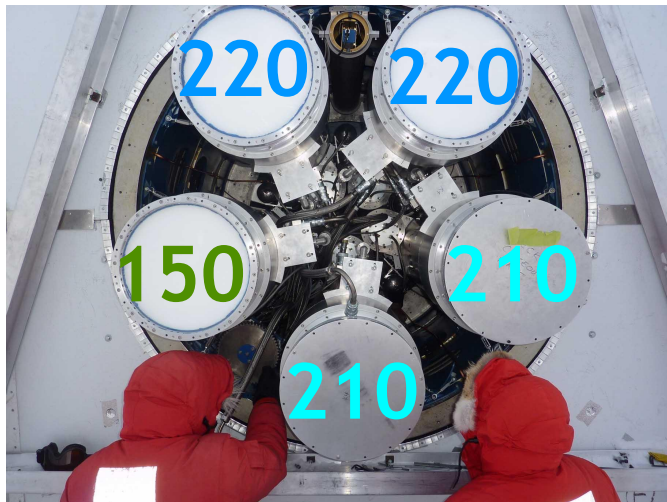
# Keck Array Frequency Coverage



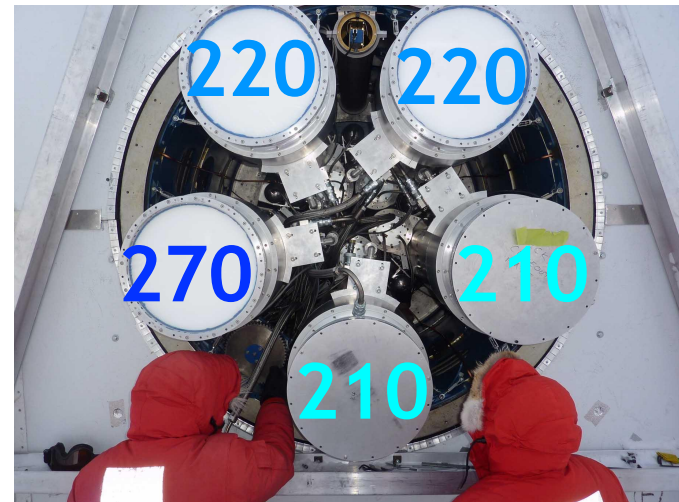
2014



2015

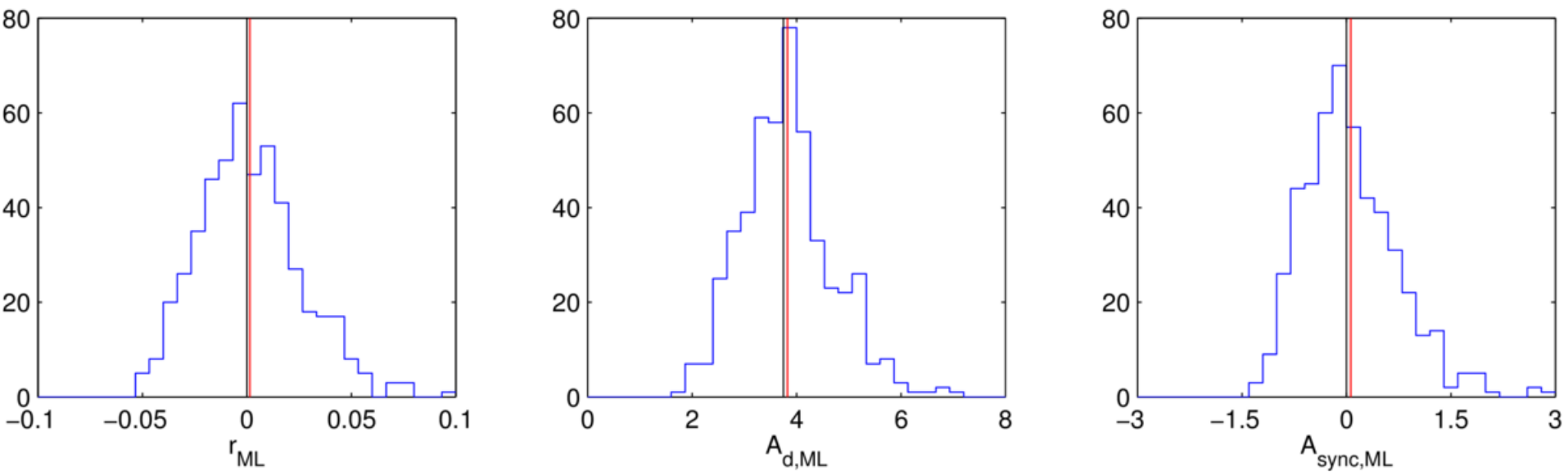


2016



2017

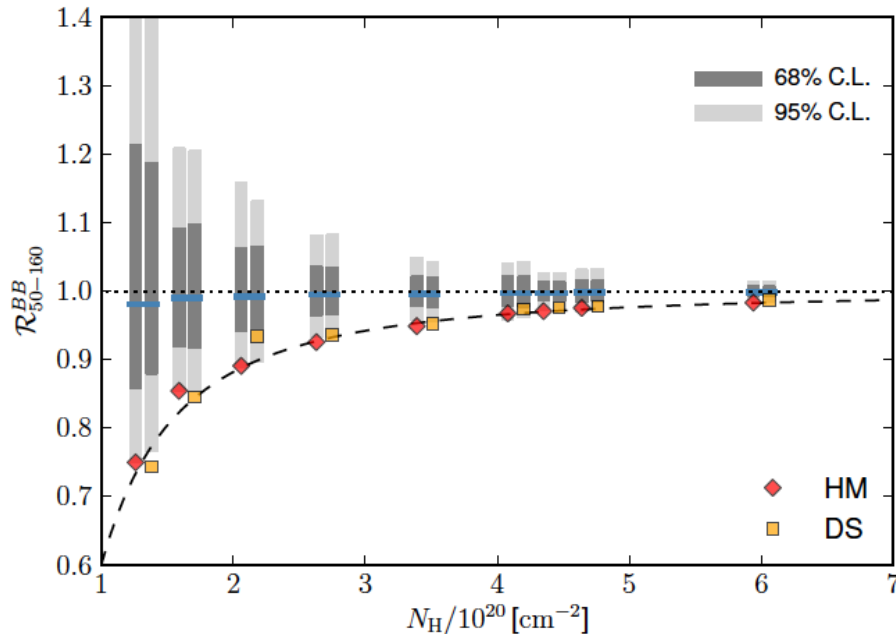
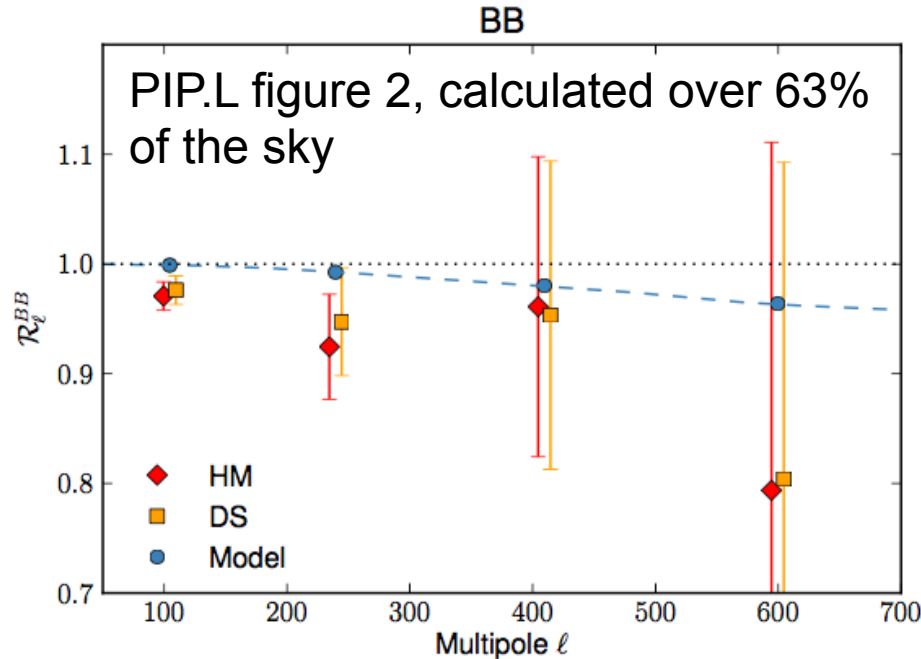
# Validation Tests



Maximum likelihoods from 500 simulations of a lensed-LCDM + dust model.

- The **means** match **the input values**.
- In the left panel the standard deviation of the histogram yields  **$\sigma(r)=0.024$**

# Evidence for dust decorrelation



Planck Intermediate Results L, A&A 599, A51 (2017), [arXiv:1606.07335](https://arxiv.org/abs/1606.07335)

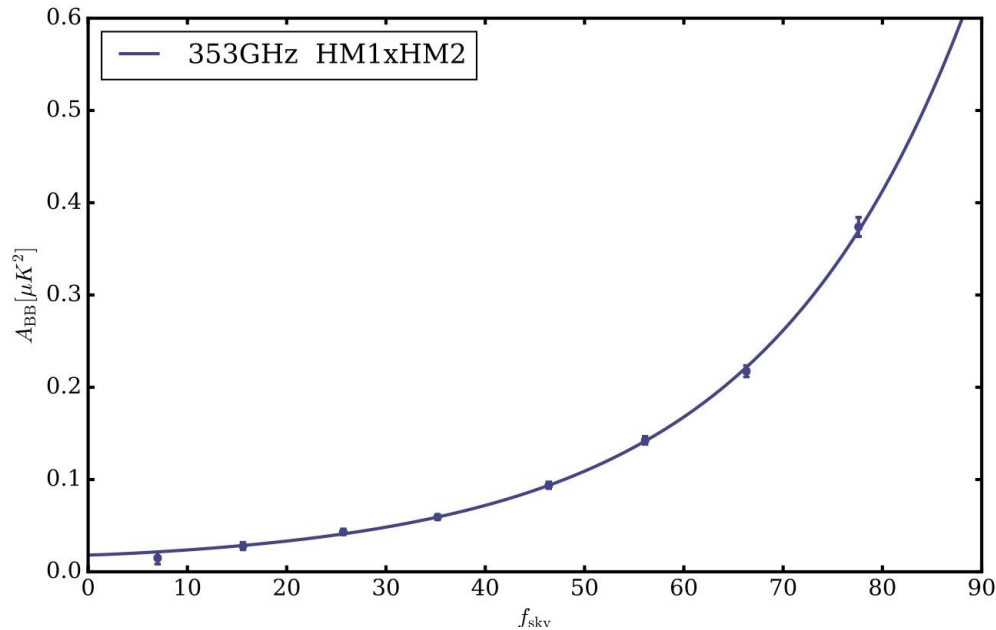
Correlation ratio calculated between 217 and 353 GHz, mostly dust-dominated for BB.

$$\mathcal{R}_\ell^{XX} \equiv \frac{C_\ell^{XX}(353 \times 217)}{\sqrt{C_\ell^{XX}(353 \times 353) C_\ell^{XX}(217 \times 217)}}, \quad (1)$$

- Extrapolation to regions of low dust column density seem to suggest improbably large decorrelation for BICEP field.
- However, in Section 4, PIP.L models dust decorrelation effect on BKP analysis assuming  $R=0.95$  between 217 and 353 GHz.
- Best constrained by our own data in our field!

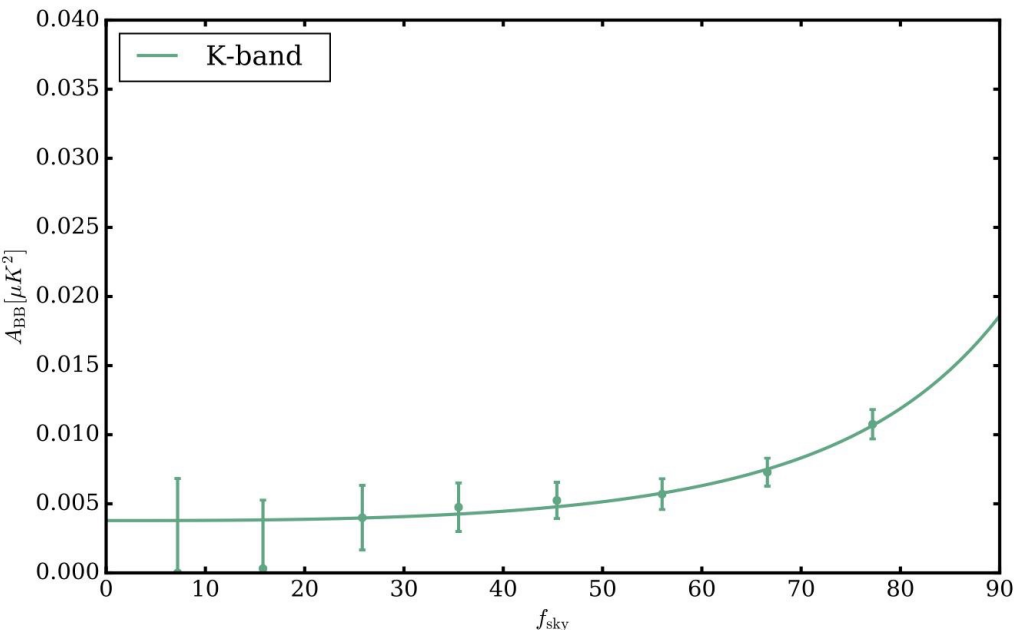


# Large fsky features (in development)



**Inclusion of foreground variation with fsky.**

Currently we assume equal foreground amplitudes even as we increase the sky area, whereas we know that above fsky of 0.1 or so the foregrounds will get brighter.

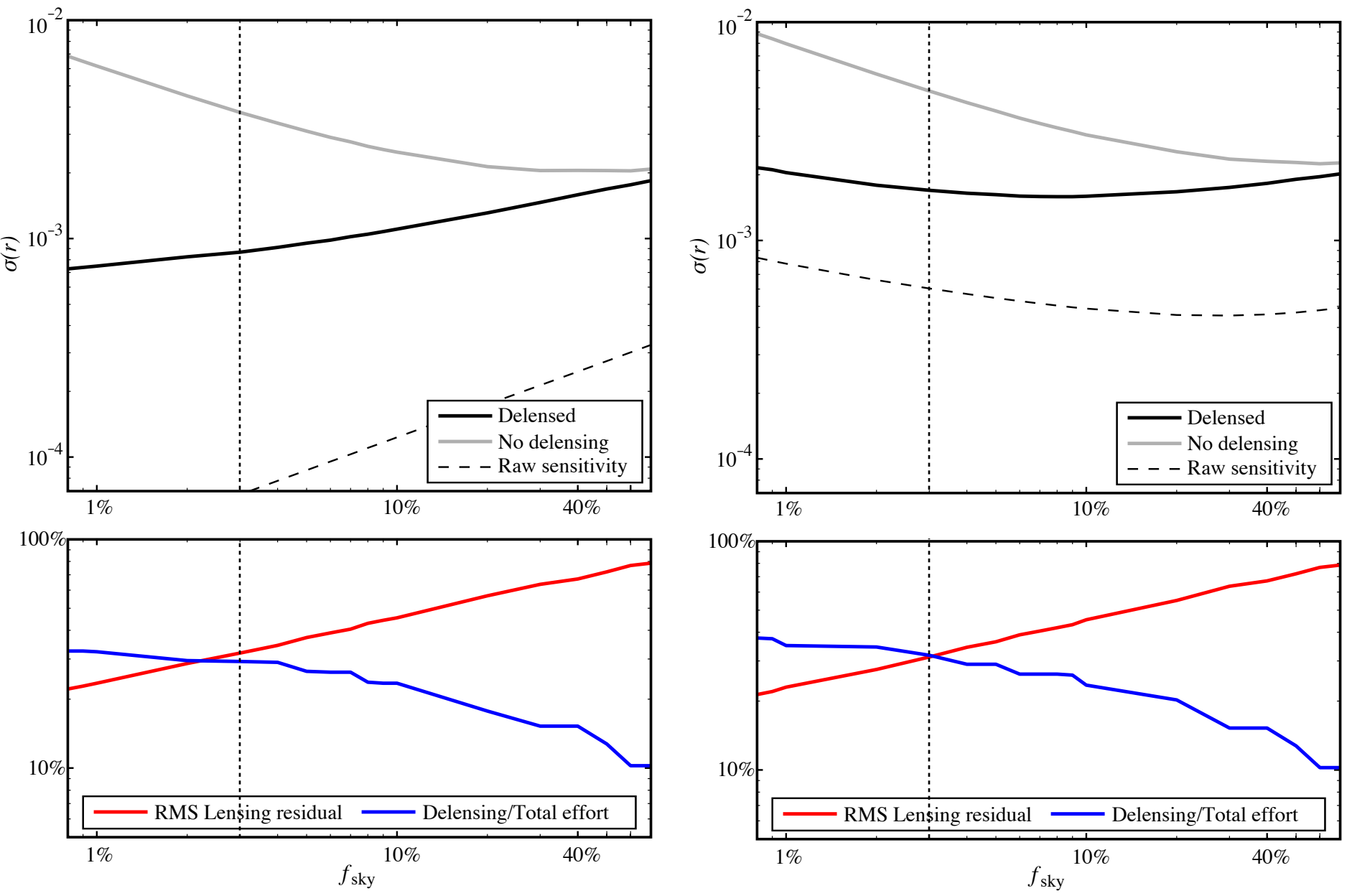


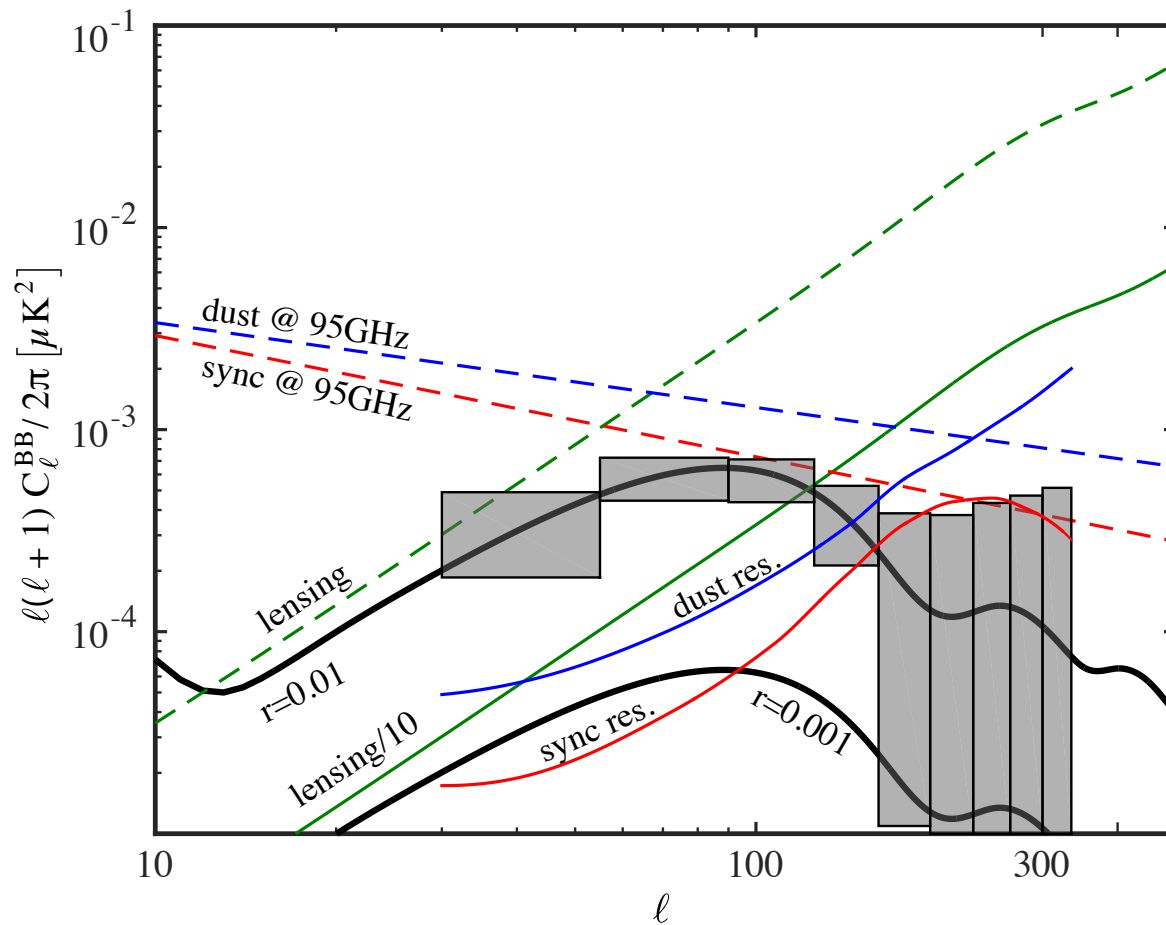
**Imposing a systematic penalty and a FG floor**

Mitigating systematics and cleaning FG's to high precision will benefit more from high S/N per mode vs raw number of modes measured. This will penalize larger fsky.

Fig. by R.Flauger

Uncertainty forecasts on  $r$ , assuming  $r = 0$  (left panel) and  $r = 0.01$  (right panel).





### Bin-by-bin forecasted tensor constraints for $r=0.01$ , $f_{\text{sky}}=3\%$ .

The contribution of dust and synchrotron to the vertical error bars are shown in solid blue and red lines. The “effective frequency” at which these foreground residuals are defined varies with each bin, allowing the residual lines to go above the input foreground model lines which are defined at a fixed frequency of 95GHz.

# Current and Future Directions

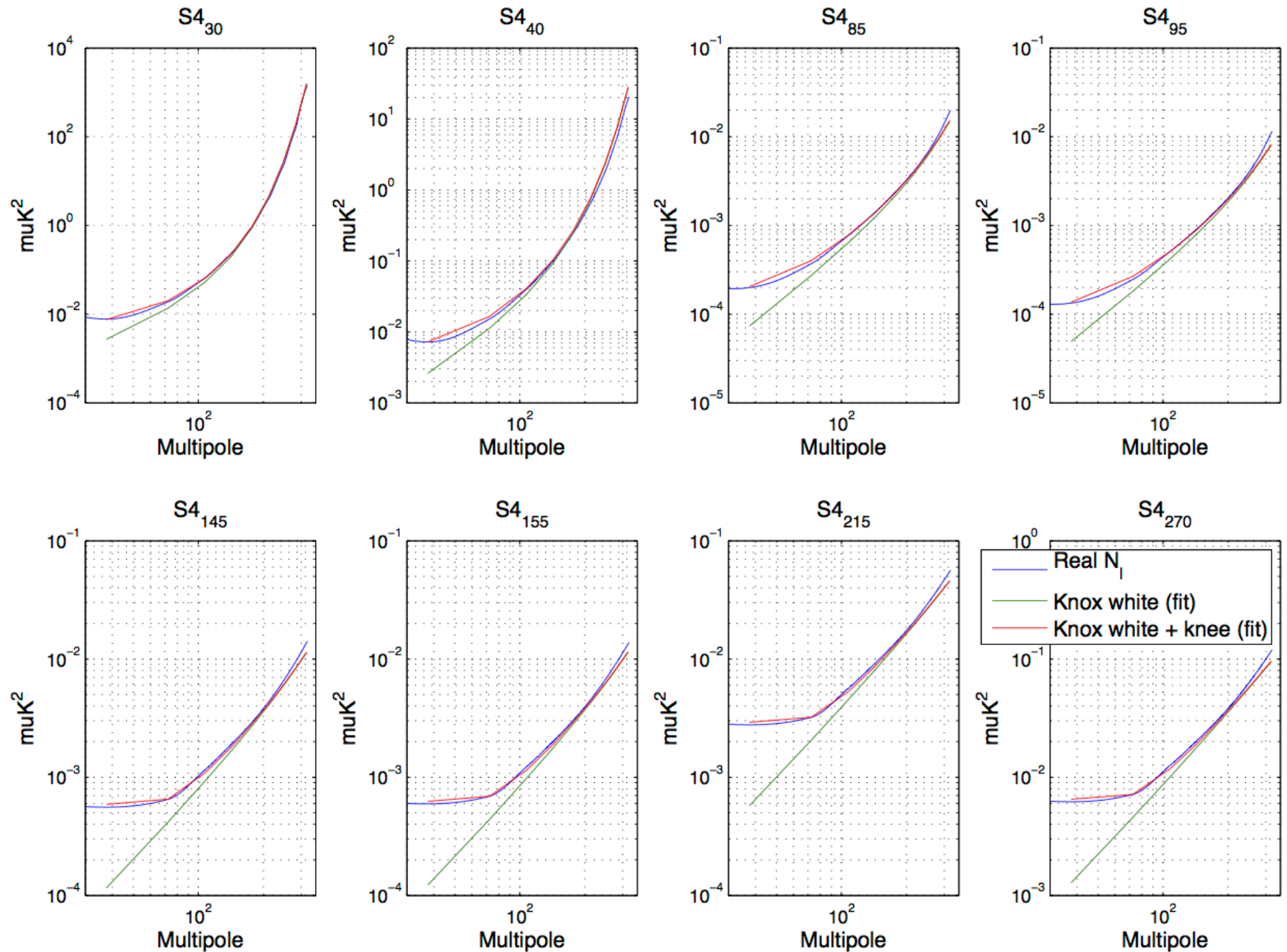
- Testing the robustness of the Likelihood/Fisher frameworks on maps from sky models with extra foreground complexity (S4 Data Challenges).
- Framework extension to take in arcmin-scale information about the lensing B-modes for “delensing” — point of contact with other collaborations.
- Including extra nuisance parameters to describe more complex foregrounds and systematics.

# Setting up the Validation

## Notes on the method:

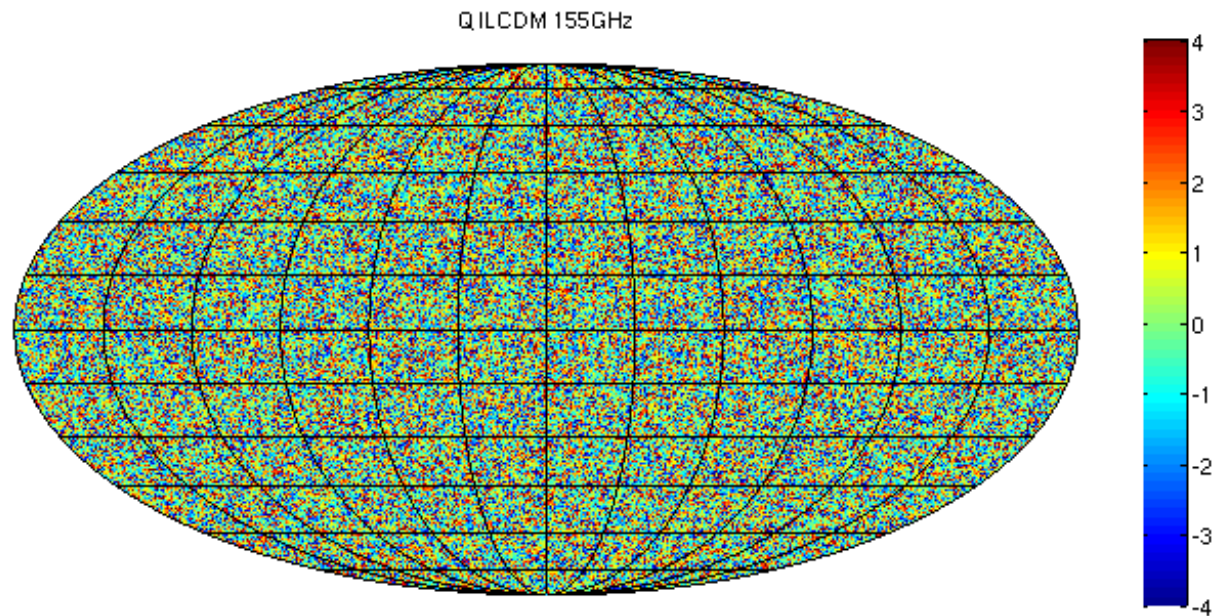
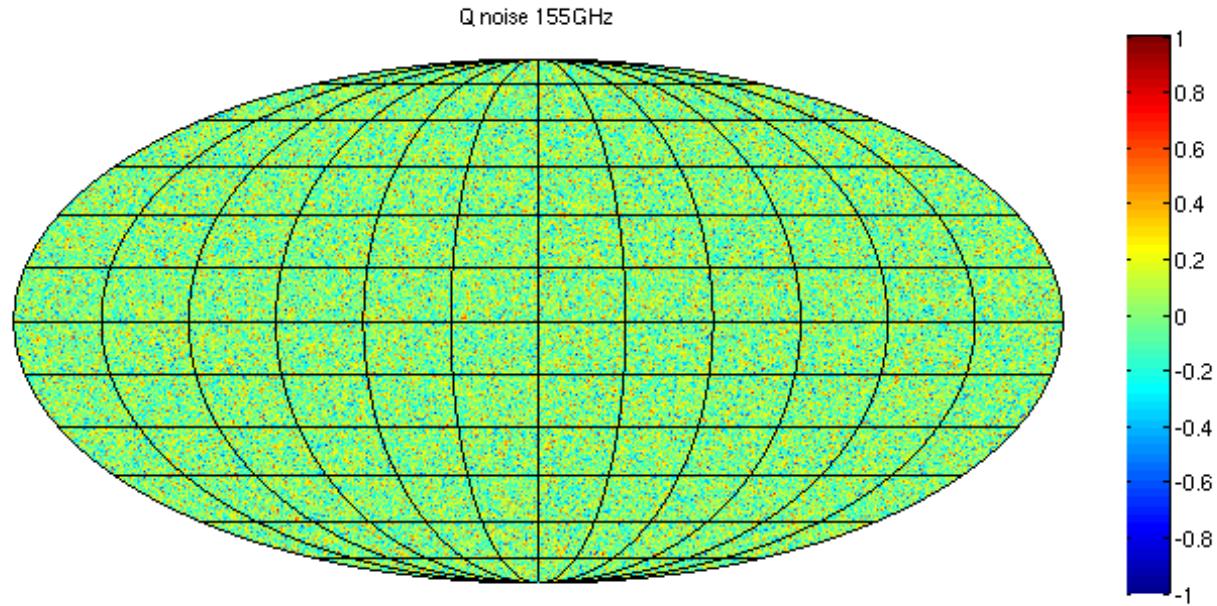
- Form Data Challenge 1.0 simulations:
  - Get S4 noise curves scaled from achieved performance
  - Form full sky Noise + LCDM + Dust + Sync maps
  - Apply  $f_{\text{sky}}$  and apodization mask
- Use a pure-B estimator (J. Grain et al, Phys. Rev. D. 79,12315) to calculate all the auto and cross spectra.
- Use the BICEP/Keck multi-component spectral-based likelihood framework.
- Use a global Maximum-Likelihood peak search (of similar dimensionality as in the Science Book Fisher forecasts, bar a dust decorrelation parameter) to obtain recovered ML histograms. The standard deviations of histograms offer a measure of the constraining power in our dataset. The means of histograms offer a measure of bias.
- Compare to Fisher results used in the CMB-S4 Science Book

# — Fit noise curves scaled from achieved performance



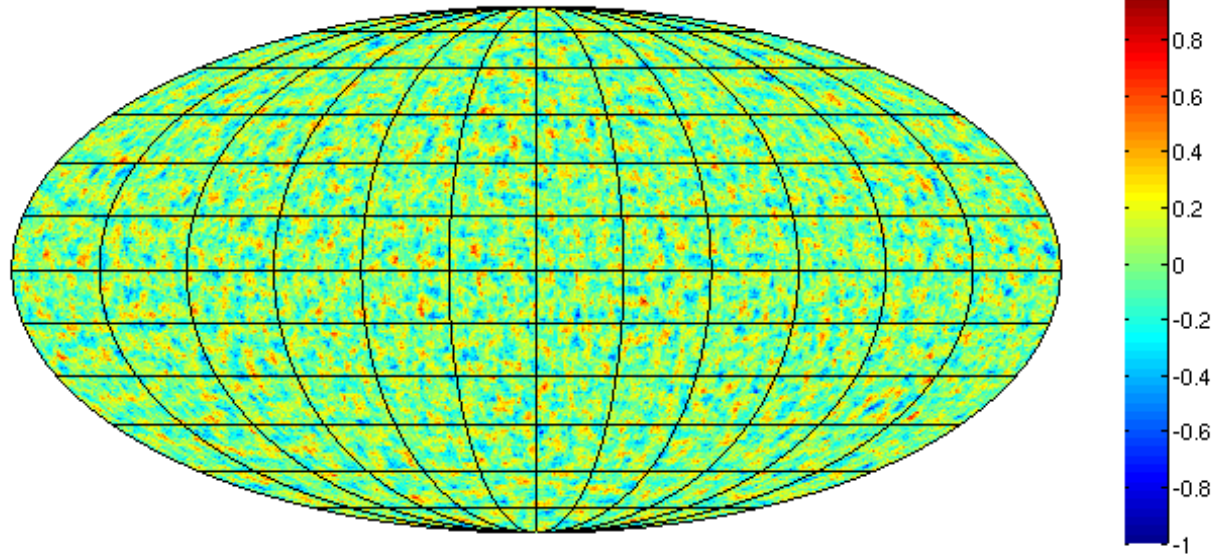


# Form noise and LCDM maps

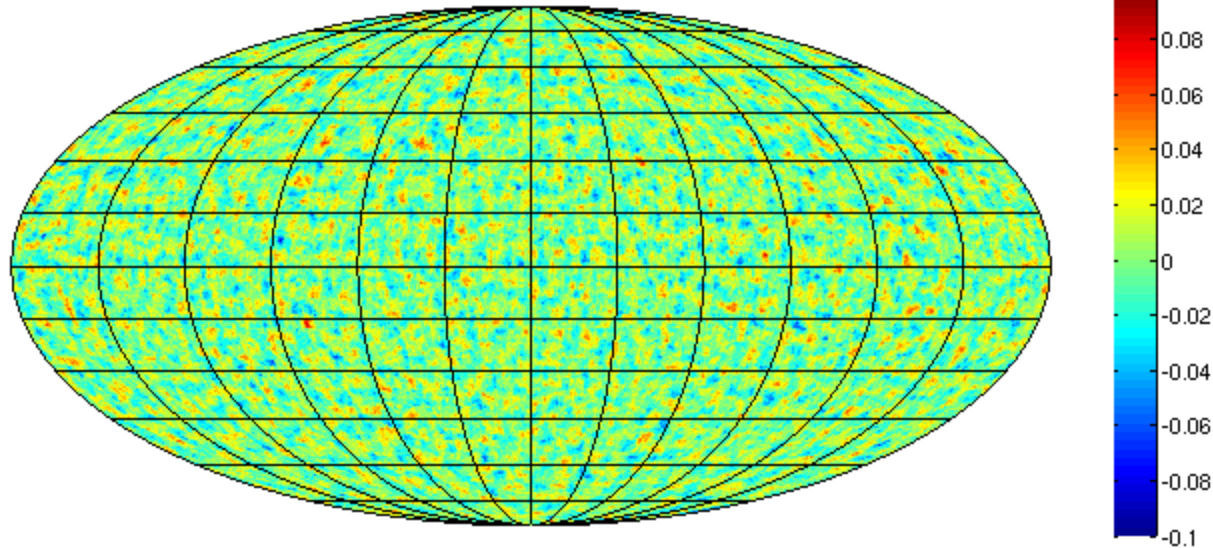


# Add Gaussian Dust and Synchrotron

Q Gaussian Dust 155GHz

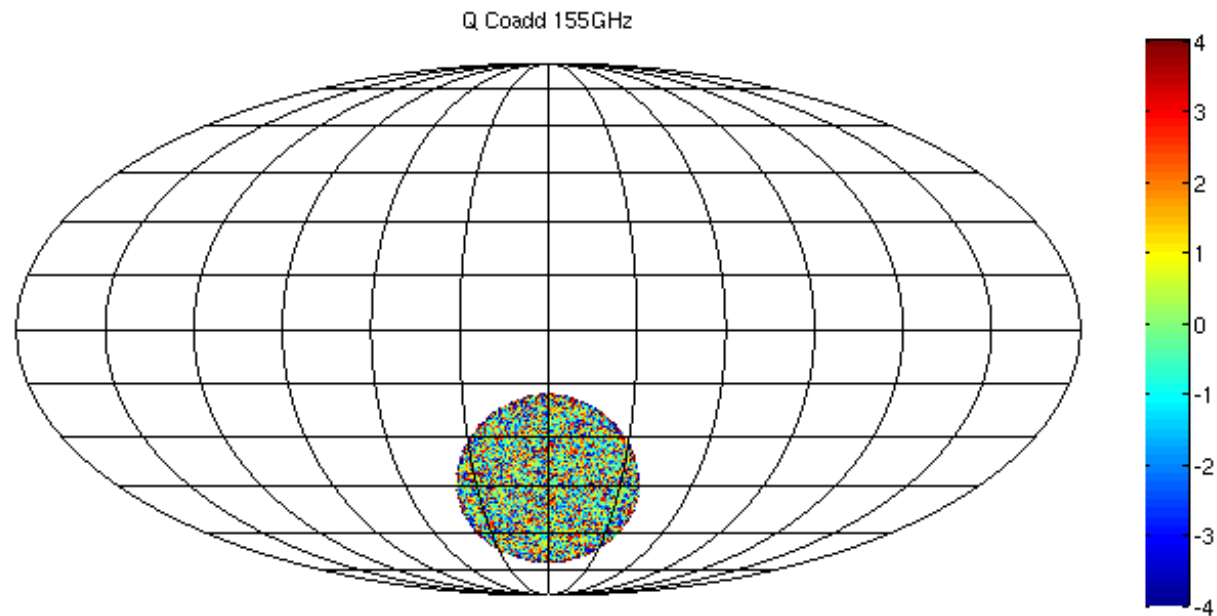
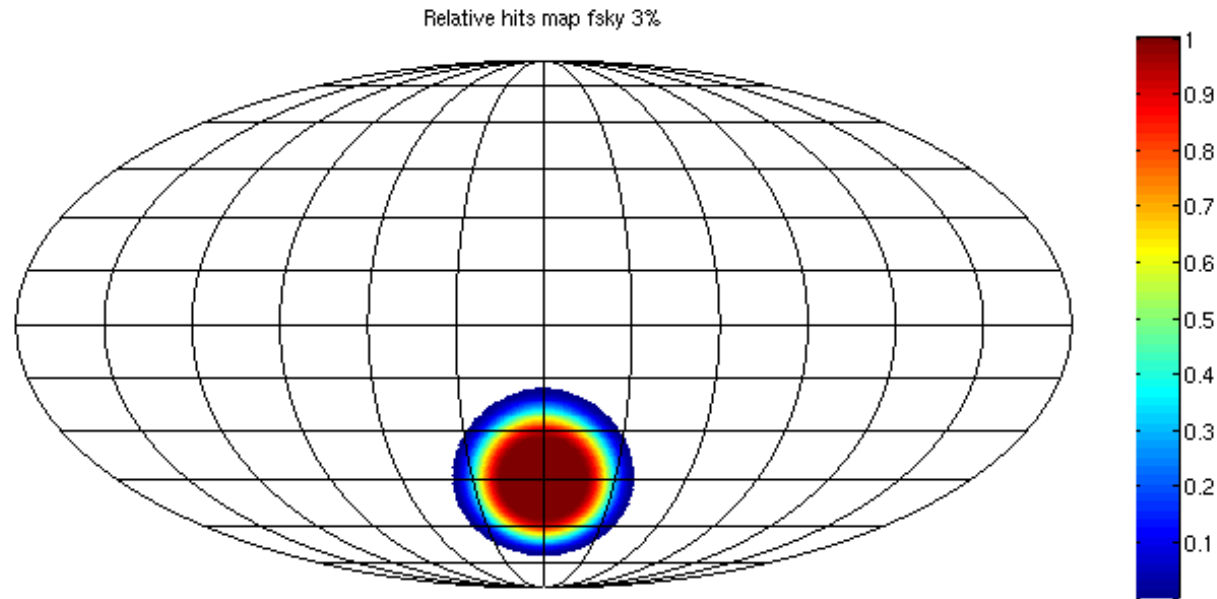


Q Gaussian Sync 155GHz

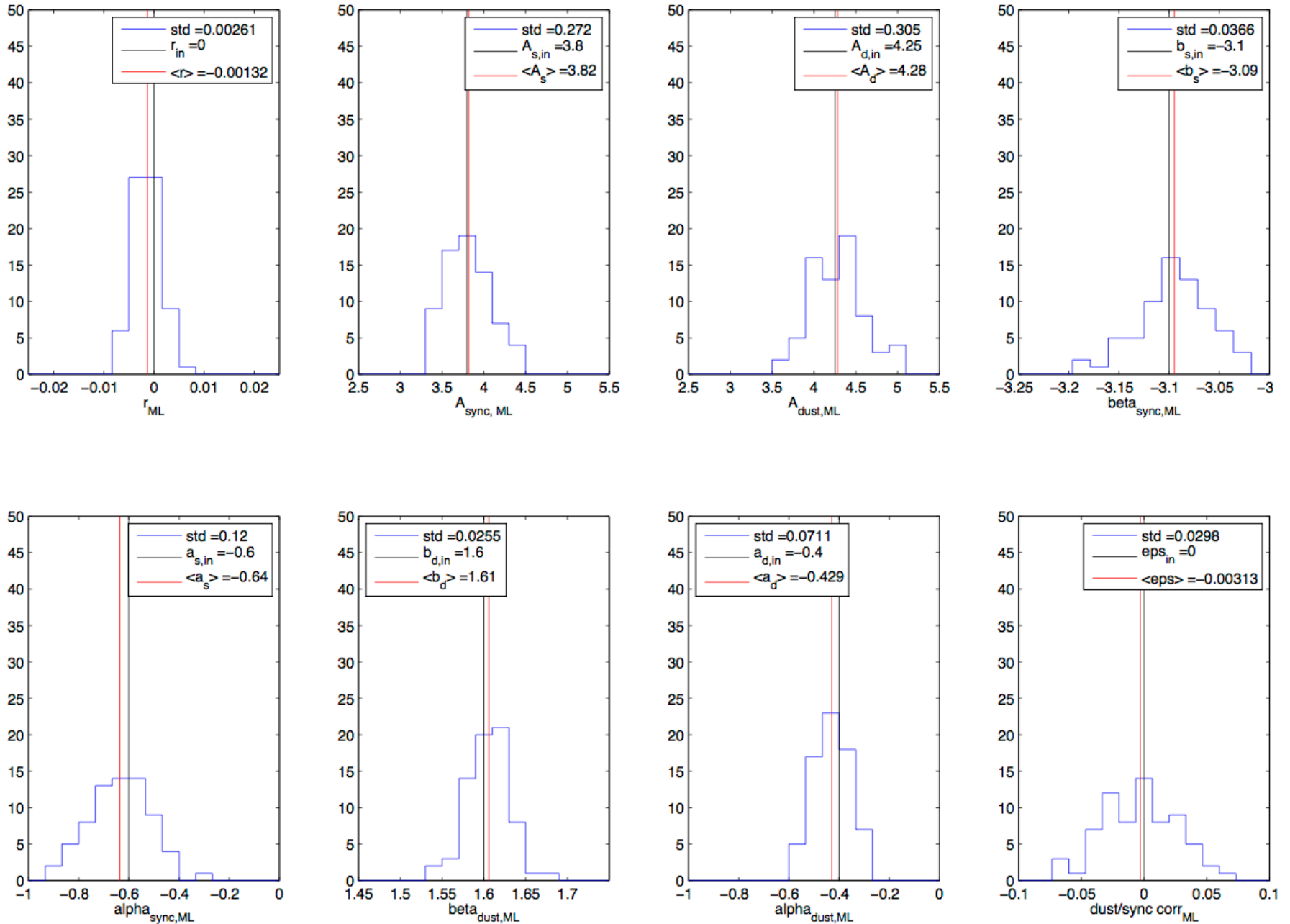


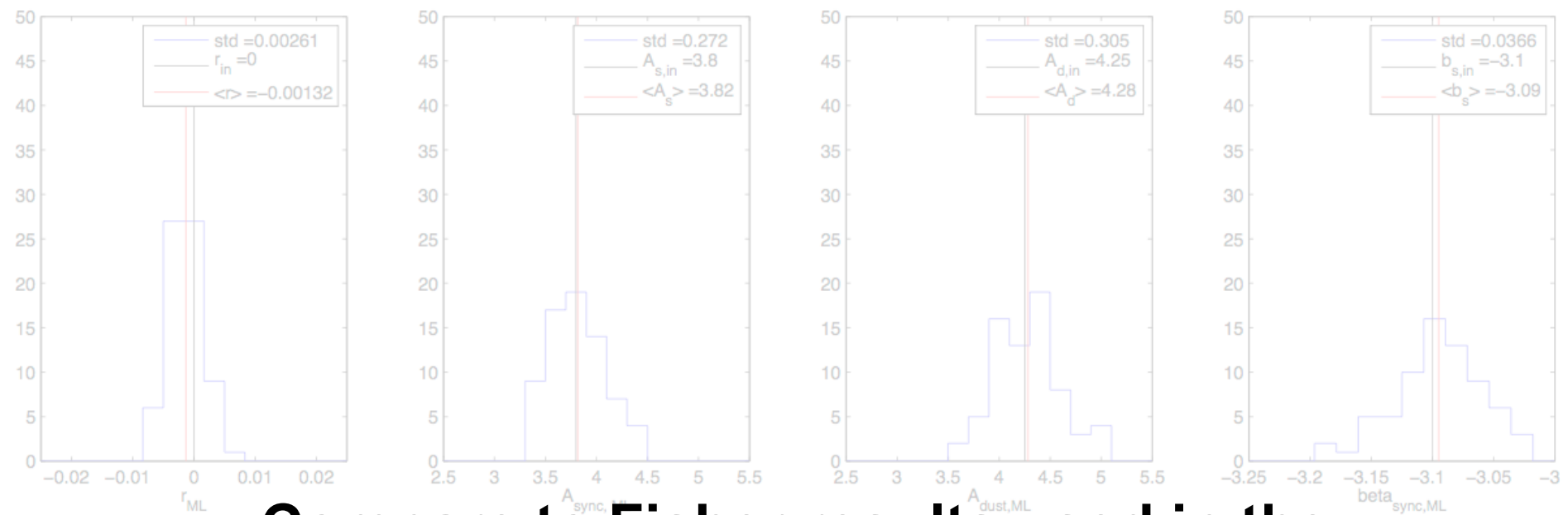


# Apply Mask and Form Full Map

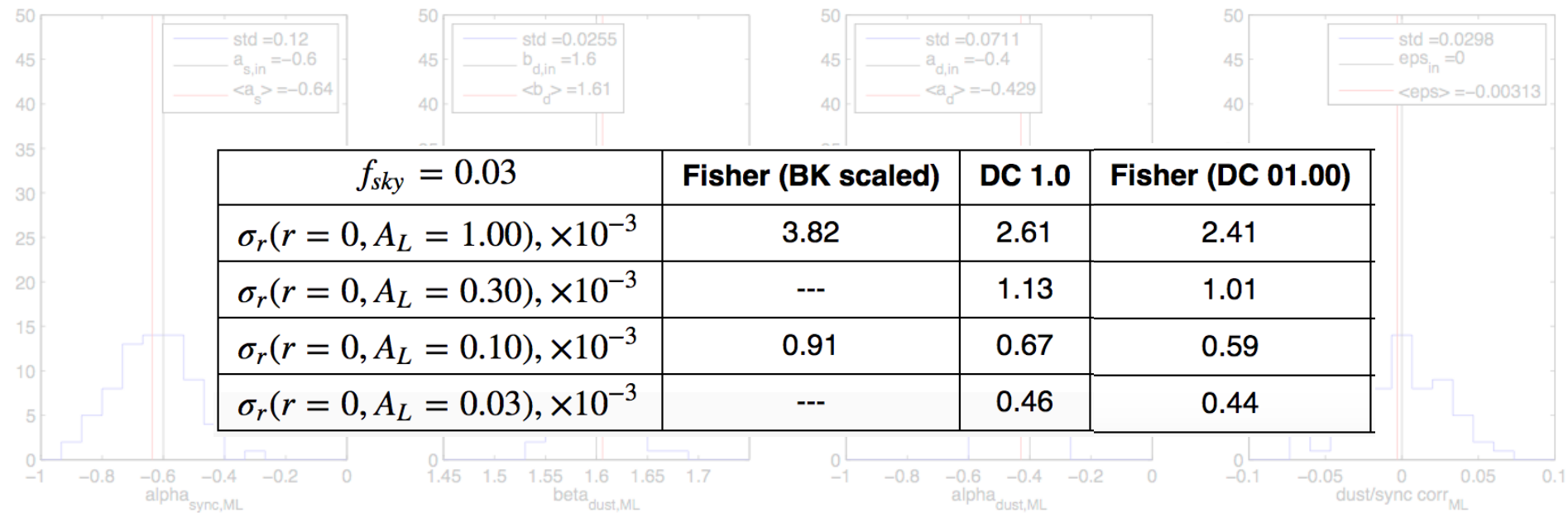


# Run Global Maximum-Likelihood Search (100 rlz)





## Compare to Fisher results used in the CMB-S4 Science Book



$f_{sky} = 0.03$	Fisher (BK scaled)	DC 1.0	Fisher (DC 01.00)
$\sigma_r(r = 0, A_L = 1.00), \times 10^{-3}$	3.82	2.61	2.41
$\sigma_r(r = 0, A_L = 0.30), \times 10^{-3}$	---	1.13	1.01
$\sigma_r(r = 0, A_L = 0.10), \times 10^{-3}$	0.91	0.67	0.59
$\sigma_r(r = 0, A_L = 0.03), \times 10^{-3}$	---	0.46	0.44