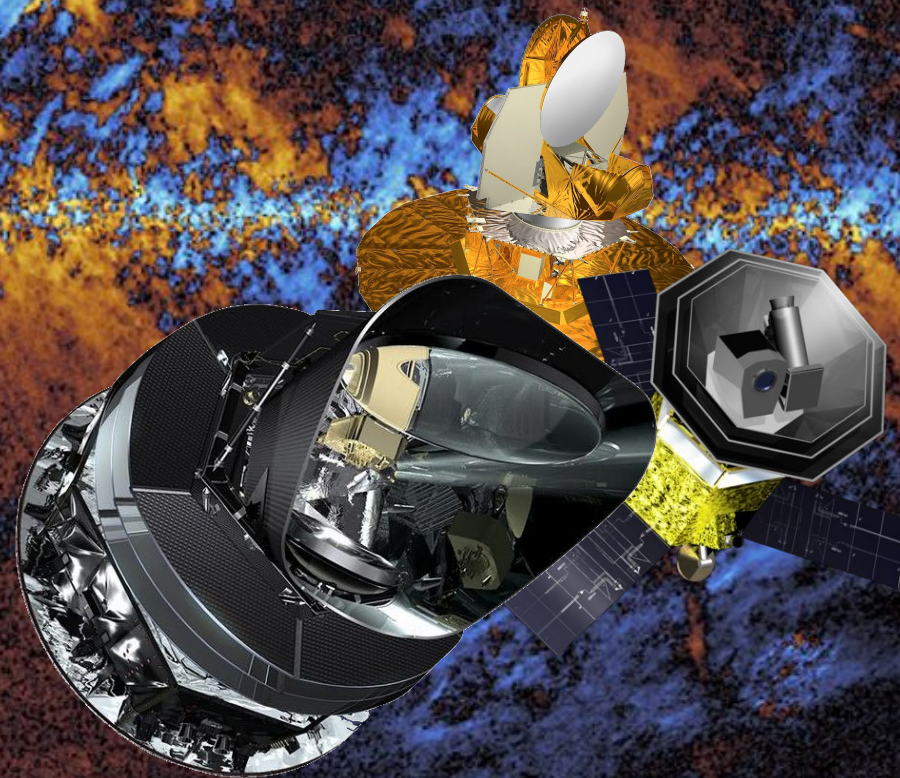


Motivation and main results

Hans Kristian Eriksen



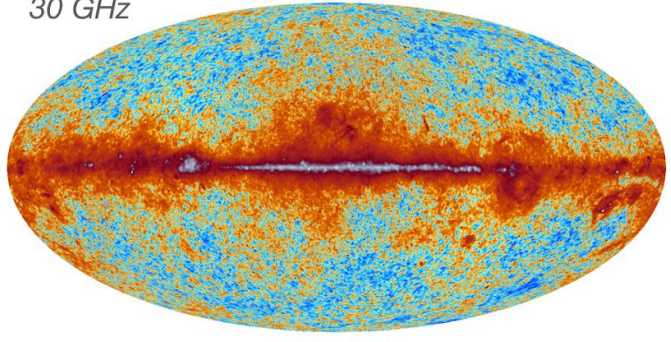
BeyondPlanck online release conference, November 18-20, 2020



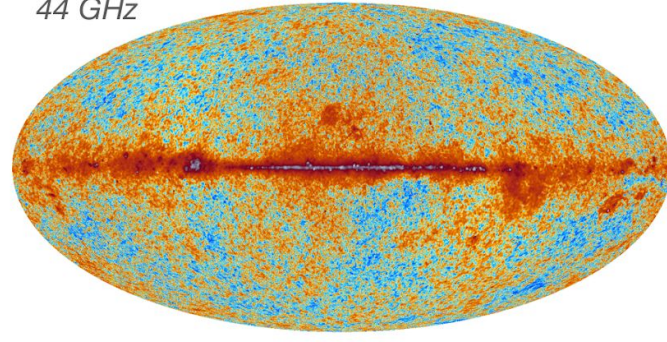
Planck 2018 frequency maps



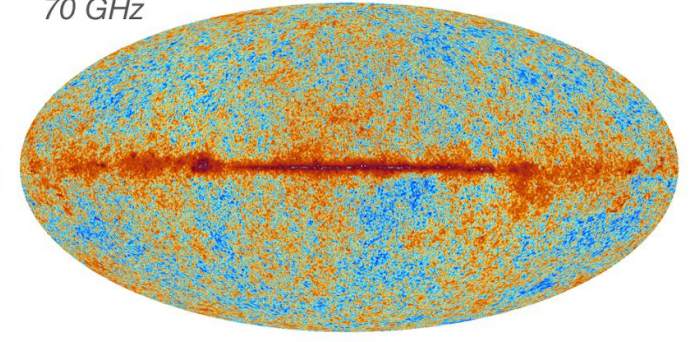
30 GHz



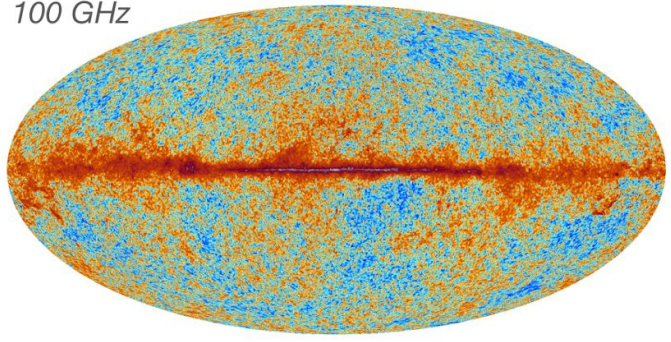
44 GHz



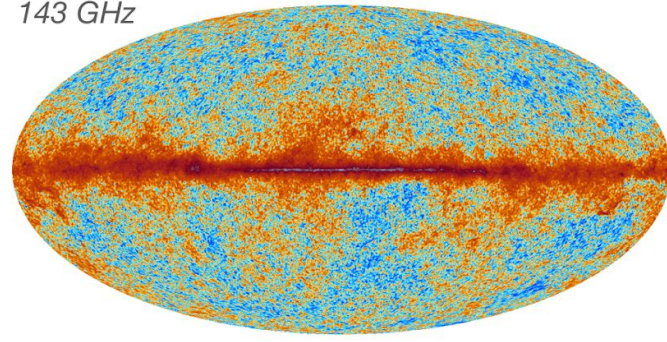
70 GHz



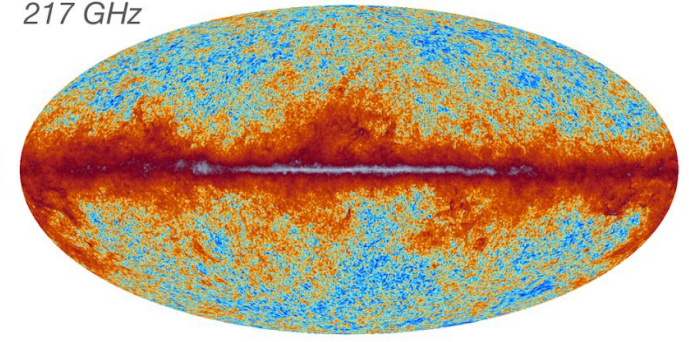
100 GHz



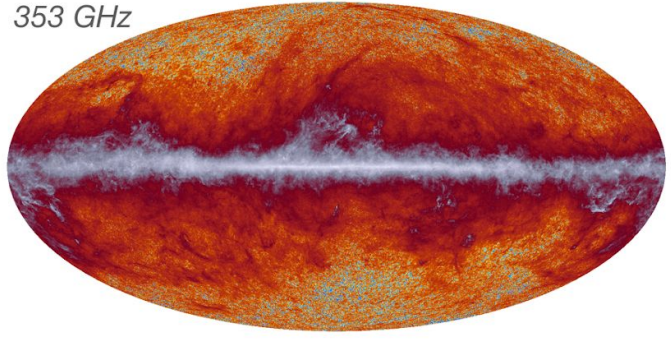
143 GHz



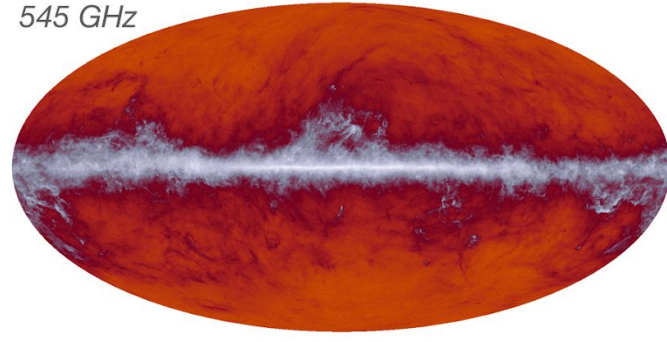
217 GHz



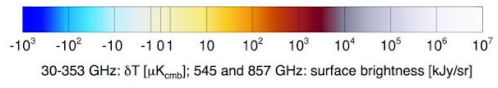
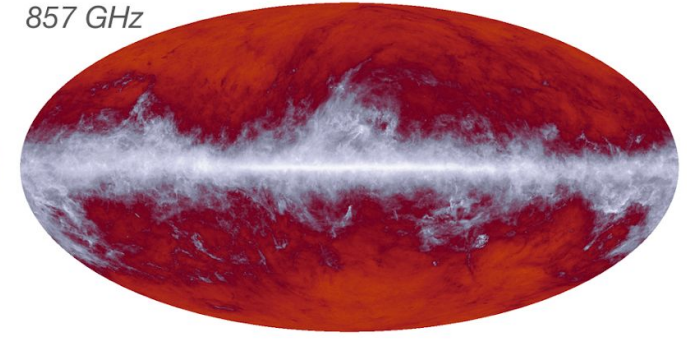
353 GHz



545 GHz



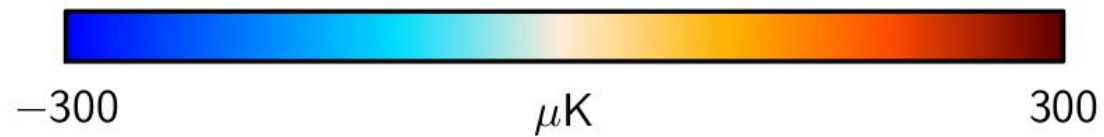
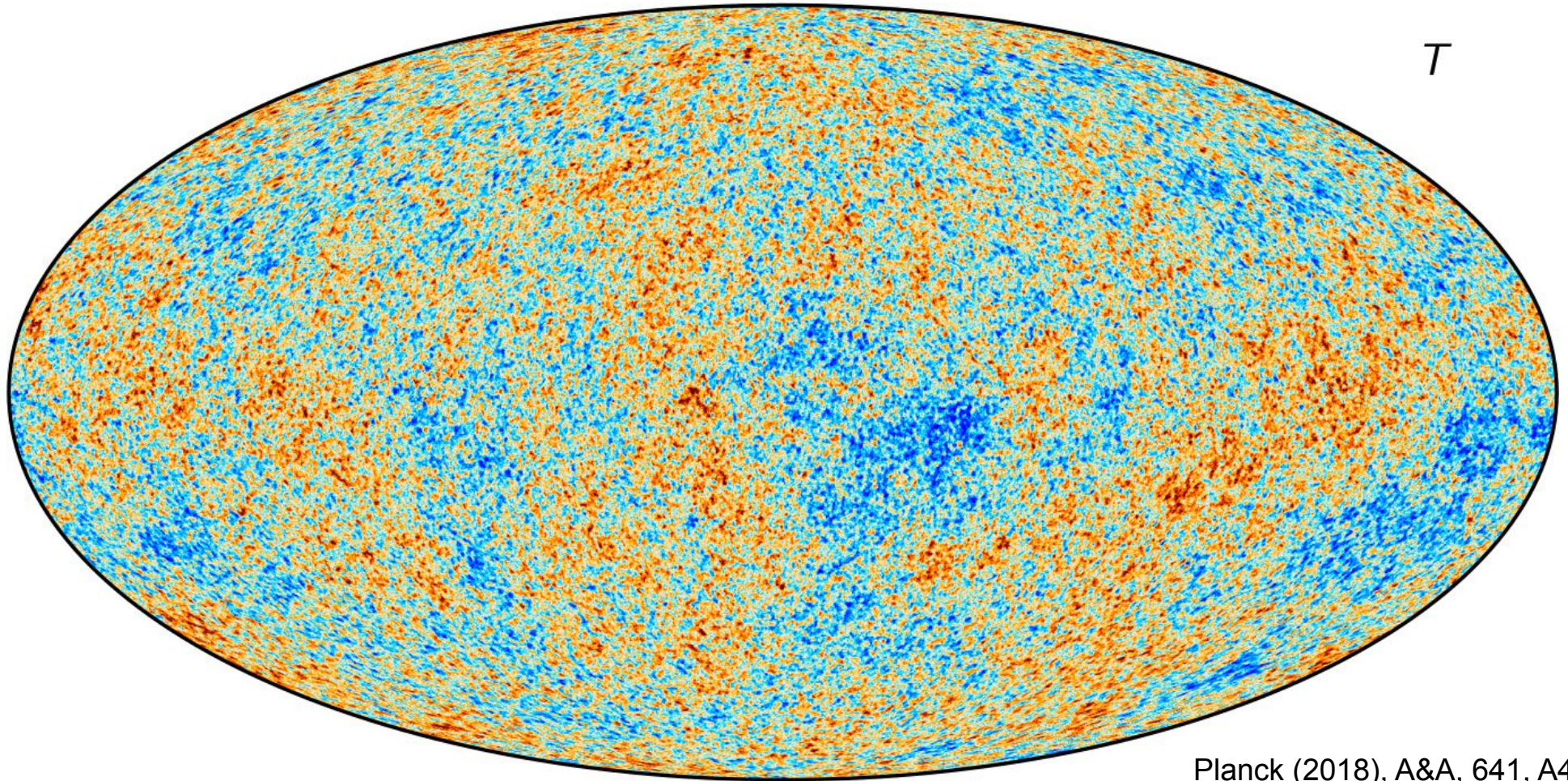
857 GHz



Planck (2018), A&A, 641, A1



Planck 2018 CMB temperature map

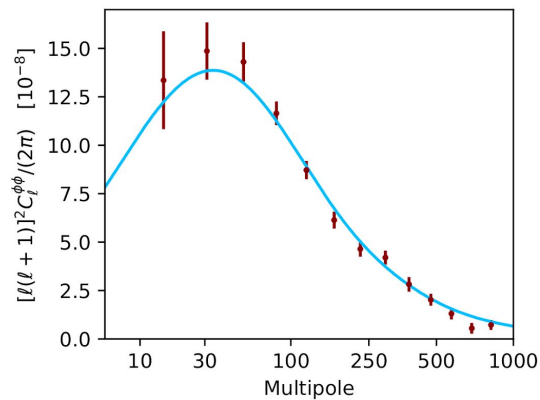
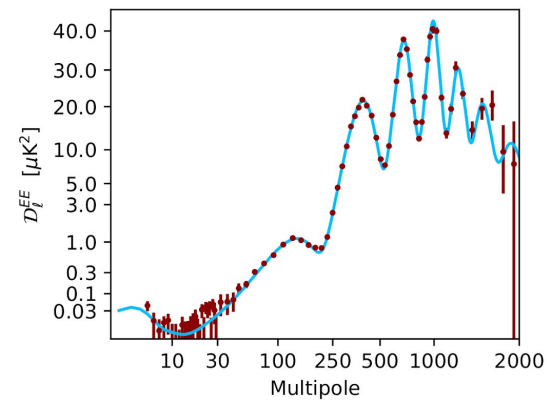
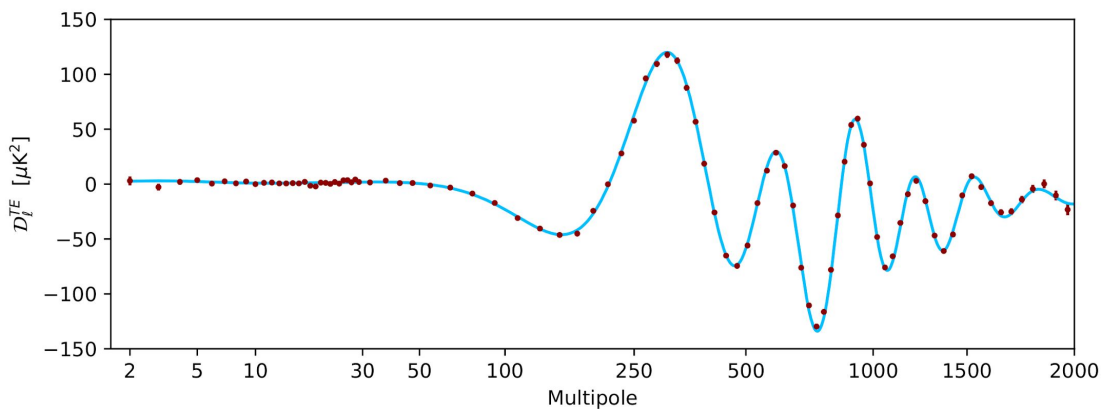
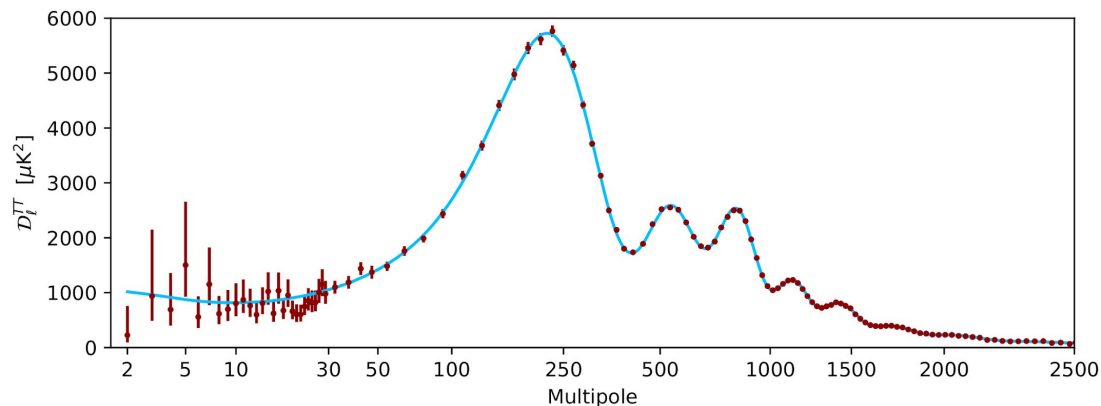


Planck (2018), A&A, 641, A4



European Commission

CMB power spectra and cosmological parameters



Parameter	Planck best fit
$\Omega_b h^2$	0.022383
$\Omega_c h^2$	0.12011
$100\theta_{MC}$	1.040909
τ	0.0543
$\ln(10^{10} A_s)$	3.0448
n_s	0.96605
$\Omega_m h^2$	0.14314
H_0 [km s ⁻¹ Mpc ⁻¹] . . .	67.32
Ω_m	0.3158
Age [Gyr]	13.7971
σ_8	0.8120
$S_8 \equiv \sigma_8 (\Omega_m / 0.3)^{0.5}$. .	0.8331
z_{re}	7.68
$100\theta_*$	1.041085
r_{drag} [Mpc]	147.049

Planck (2018), A&A, 641, A5

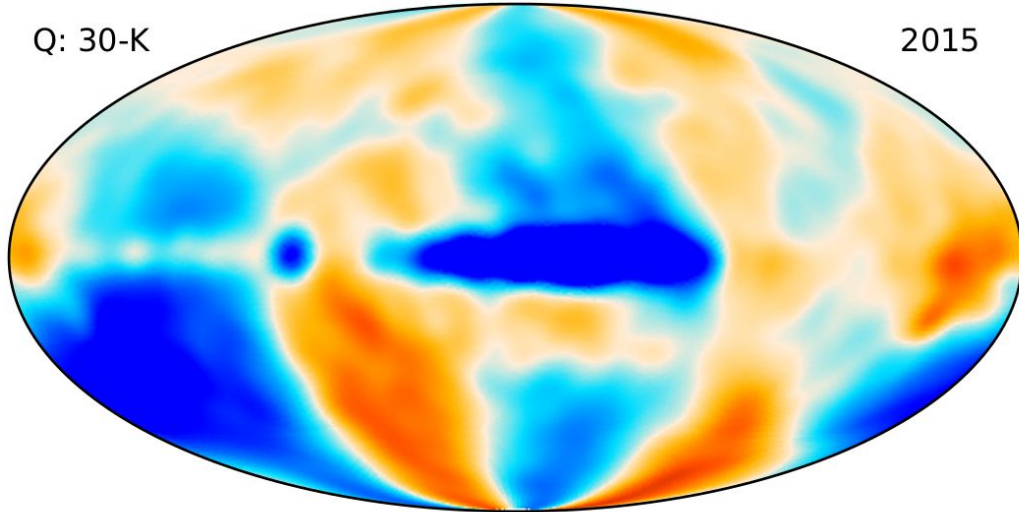


What about Planck - WMAP?



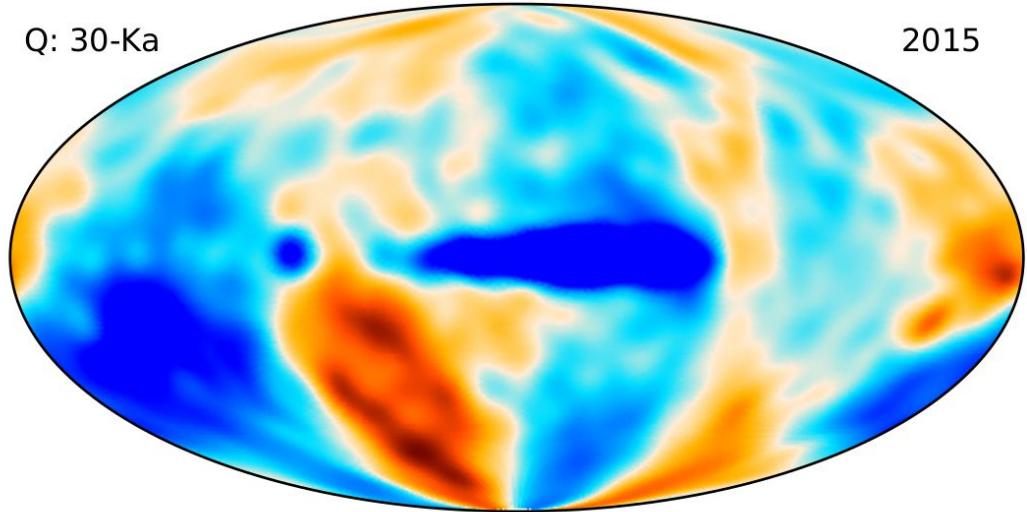
Q: 30-K

2015



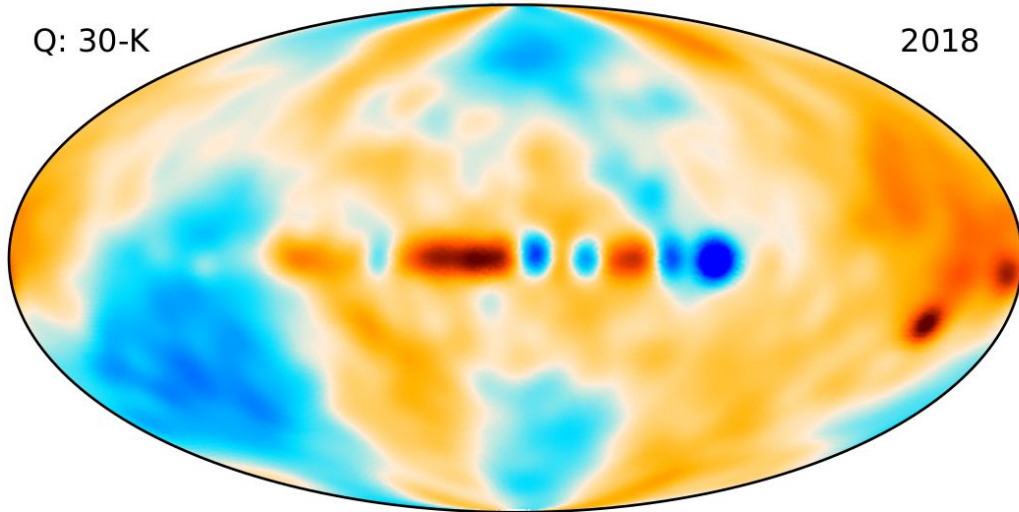
Q: 30-Ka

2015



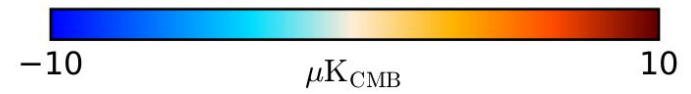
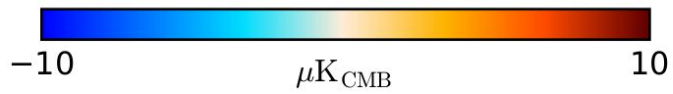
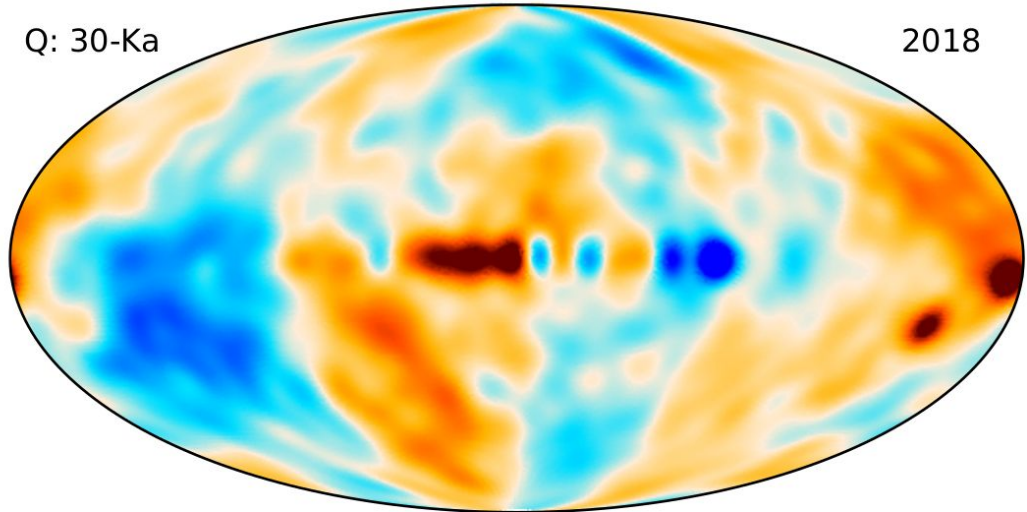
Q: 30-K

2018

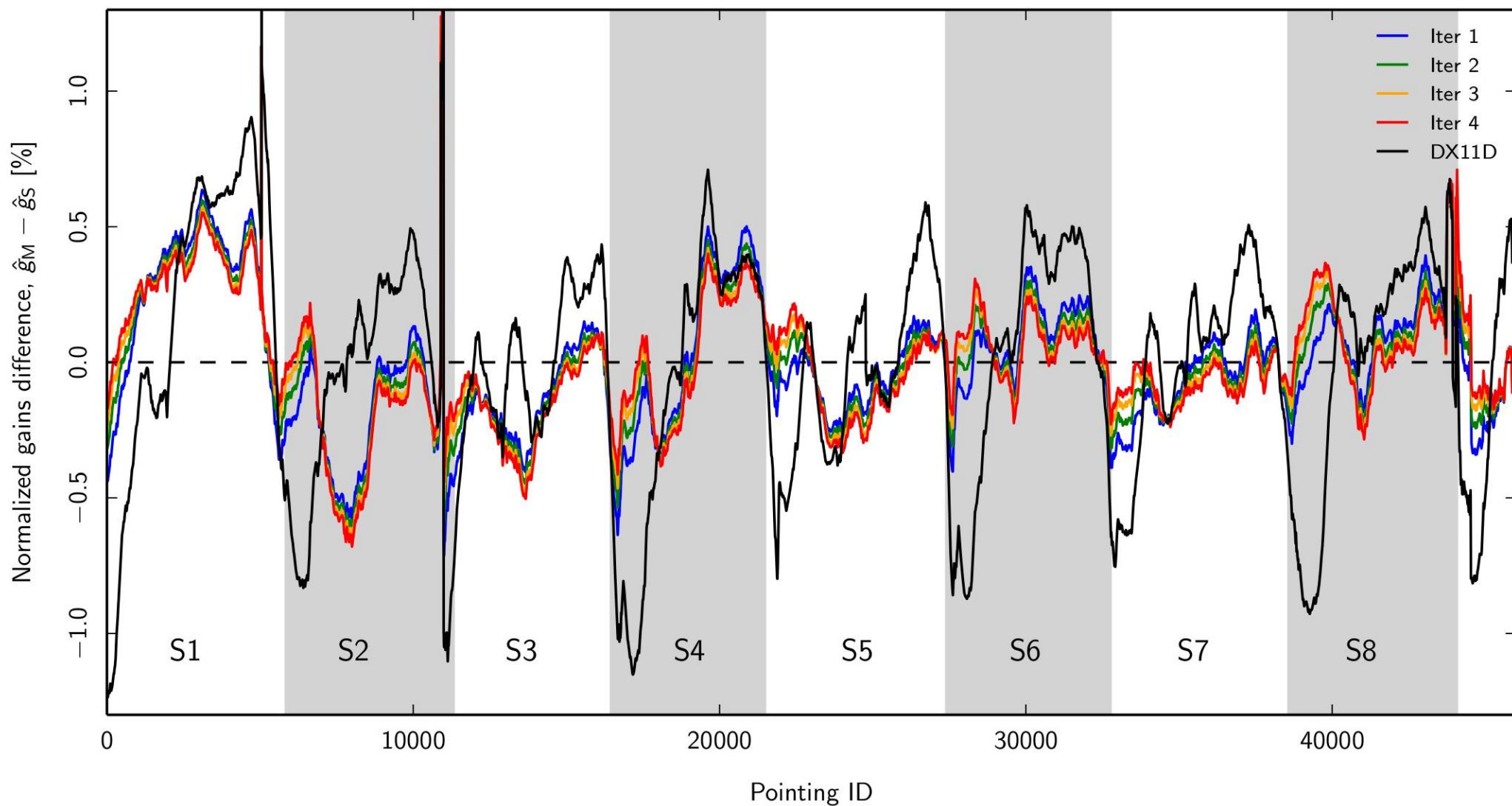


Q: 30-Ka

2018

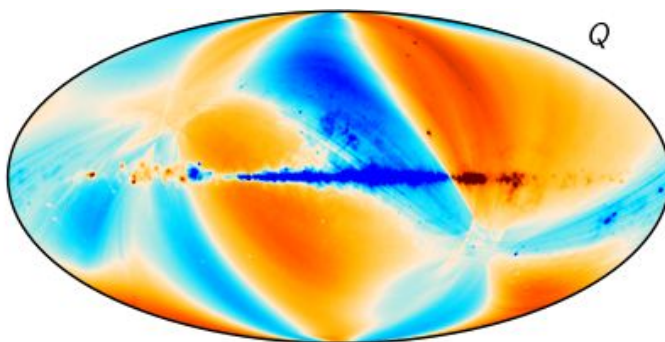
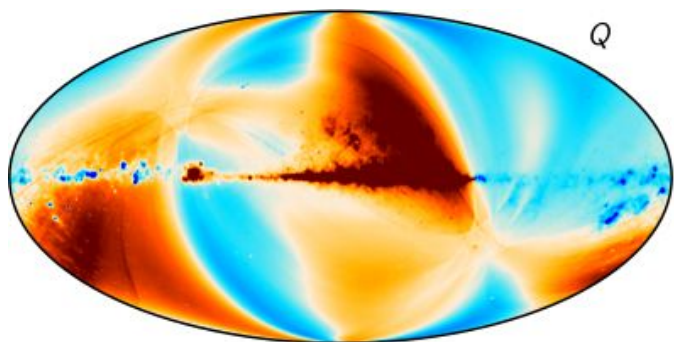


Critical question: How well do we really know the gain?

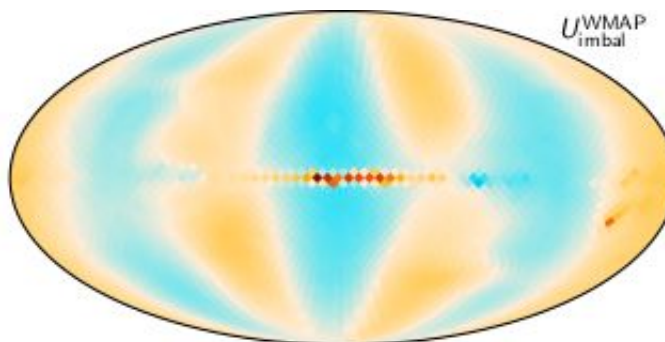
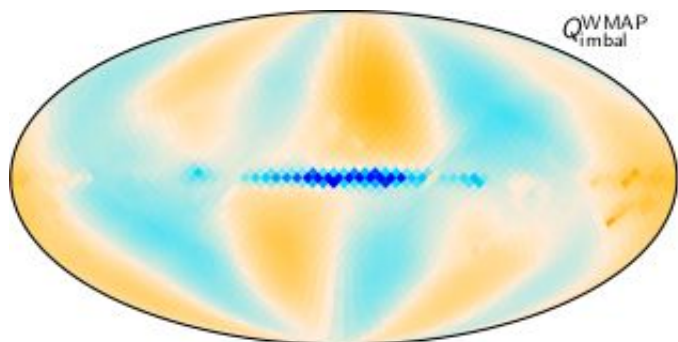


Planck (2018), A&A, 641, A2

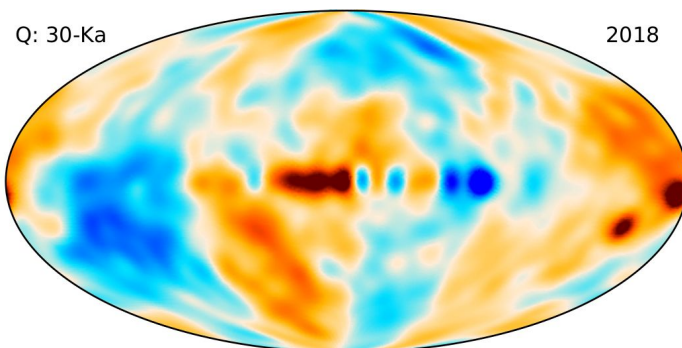
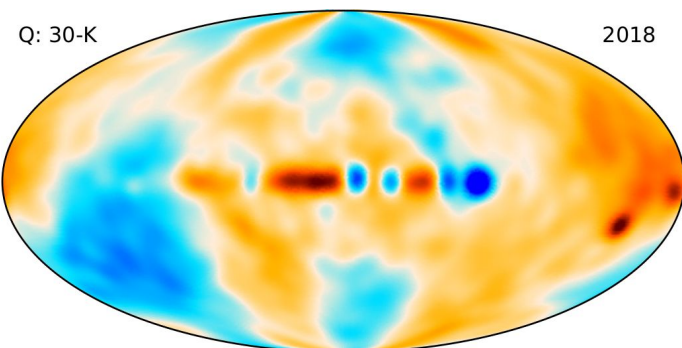
Known poorly measured modes in Planck and WMAP



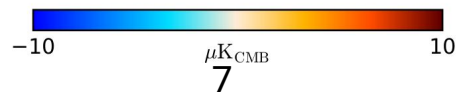
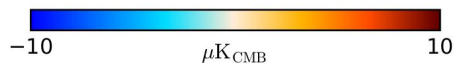
Planck 2018 30 GHz gain residual template



WMAP K-band transmission imbalance template



30 - K difference map



Can we address the outstanding issues seen in Planck LFI by:

1. speeding up the iteration process, and perform hundreds of component separation + calibration iterations, not just four?
2. break internal Planck-specific degeneracies using external data, in particular WMAP?

The name BeyondPlanck was chosen to

- recognize that this work builds on, and is a natural continuation of, the official Planck analysis effort
- emphasize that this involves not only Planck, but also other data sets

Why do we care?



Gravitational waves from black holes



LIGO

Gravitational waves from the Big Bang



CORE

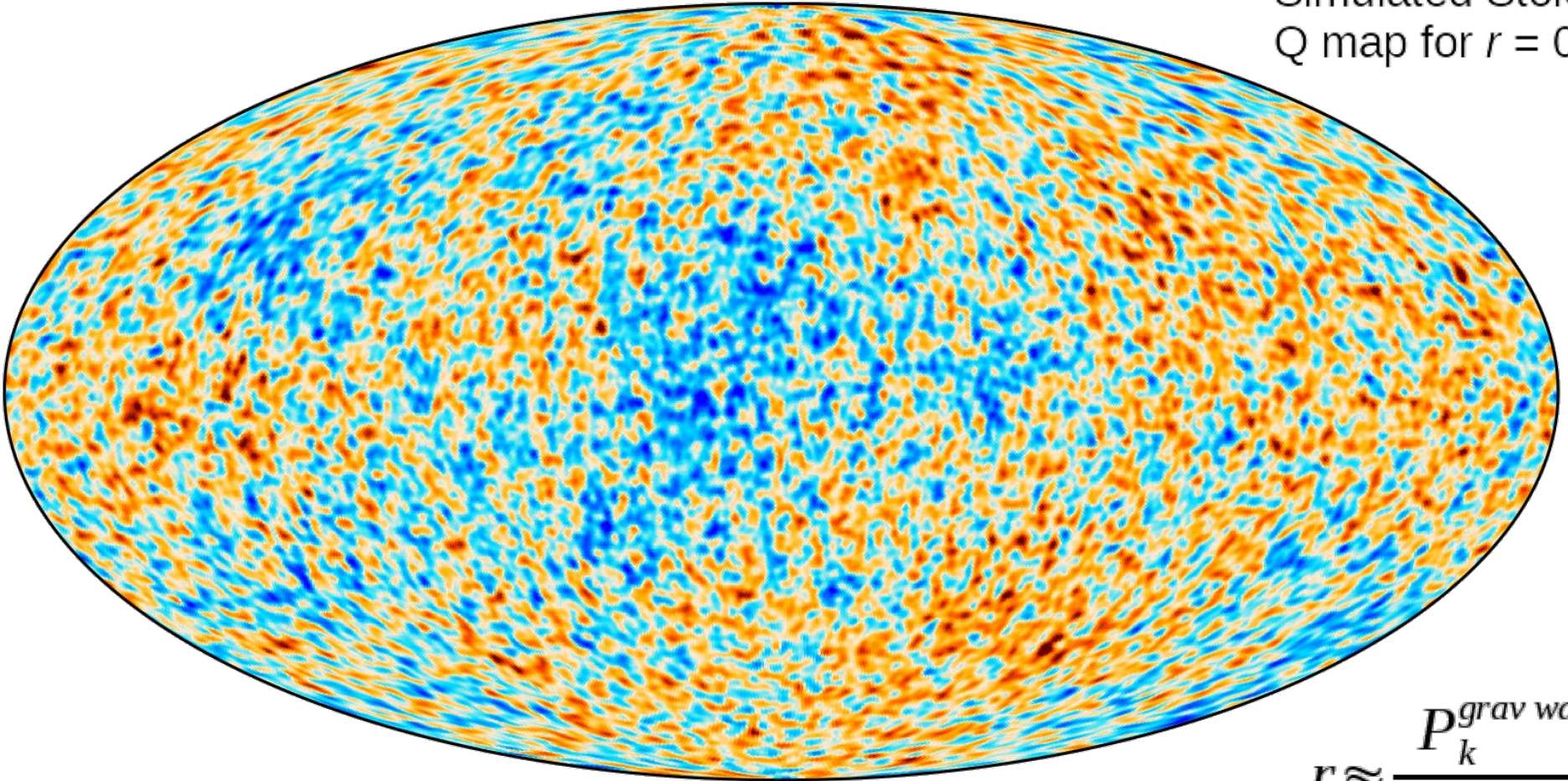
PICO

LiteBIRD

What sort of precision is required for gravitational waves?



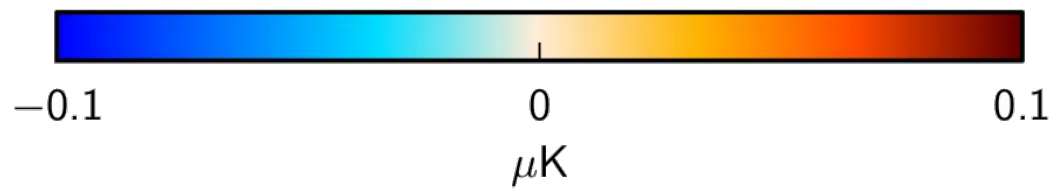
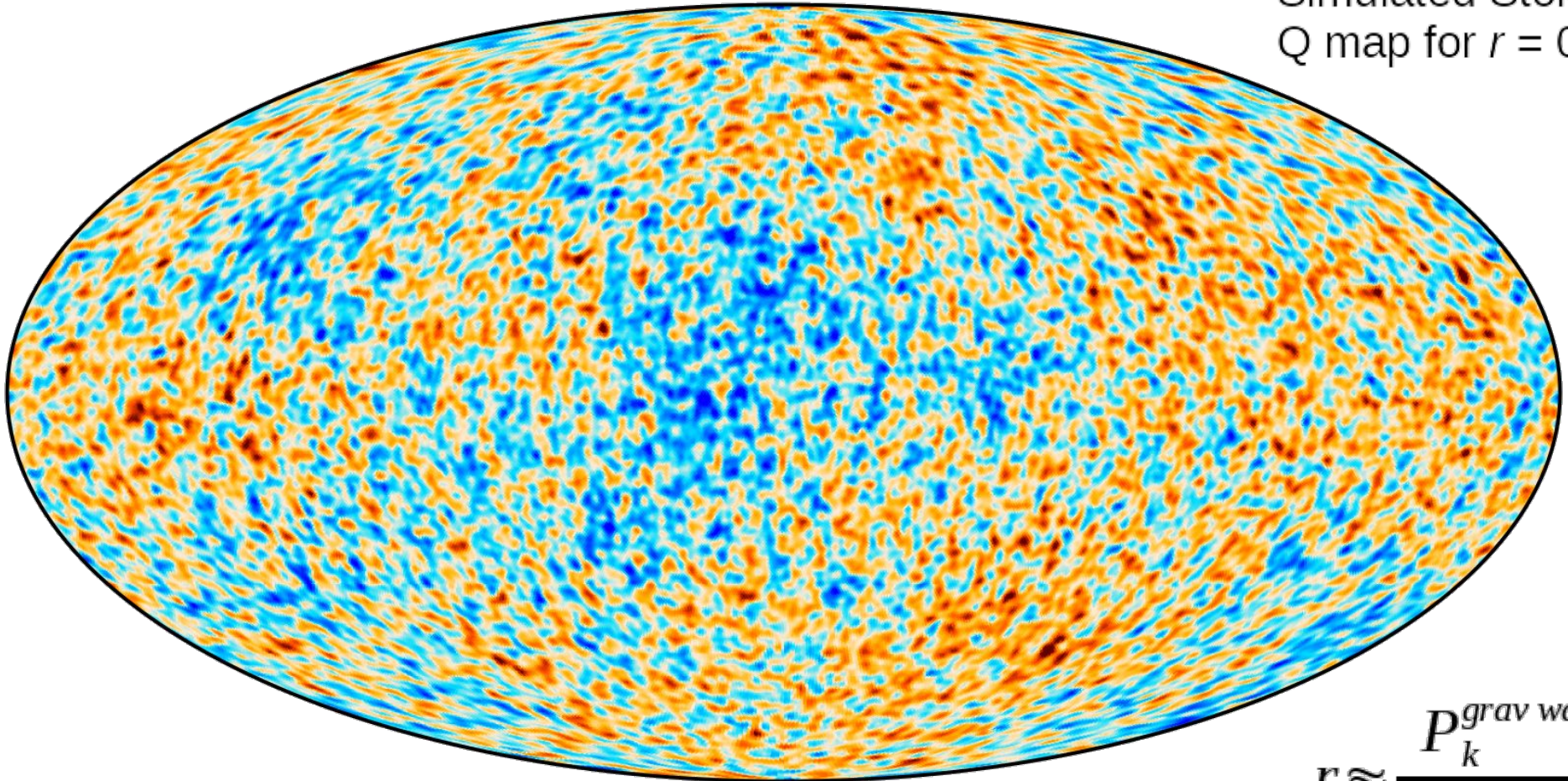
Simulated Stokes
Q map for $r = 0.01$



$$r \approx \frac{P_k^{\text{grav waves}}}{P_k^{\text{density waves}}}$$

What sort of precision is required for gravitational waves?

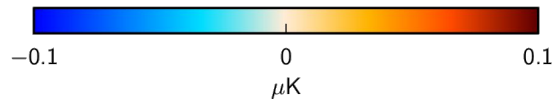
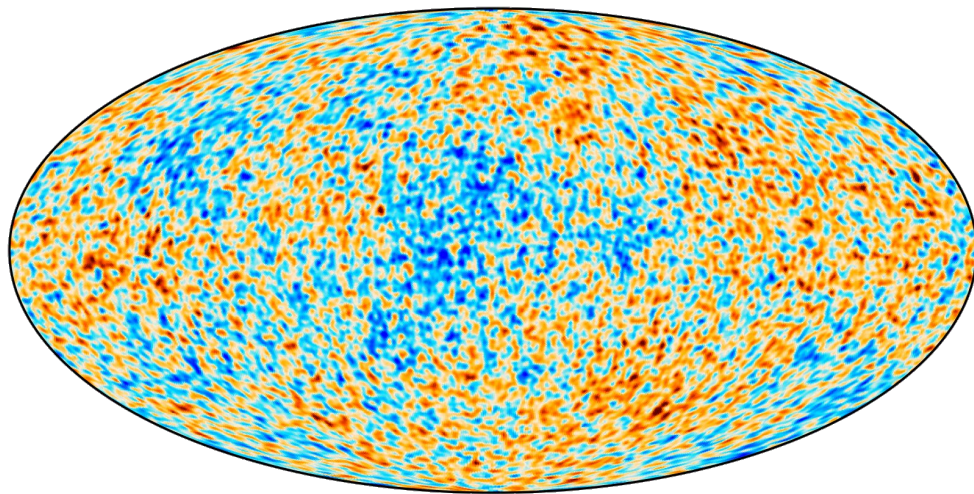
Simulated Stokes
Q map for $r = 0.01$



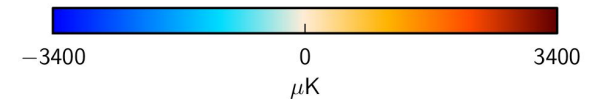
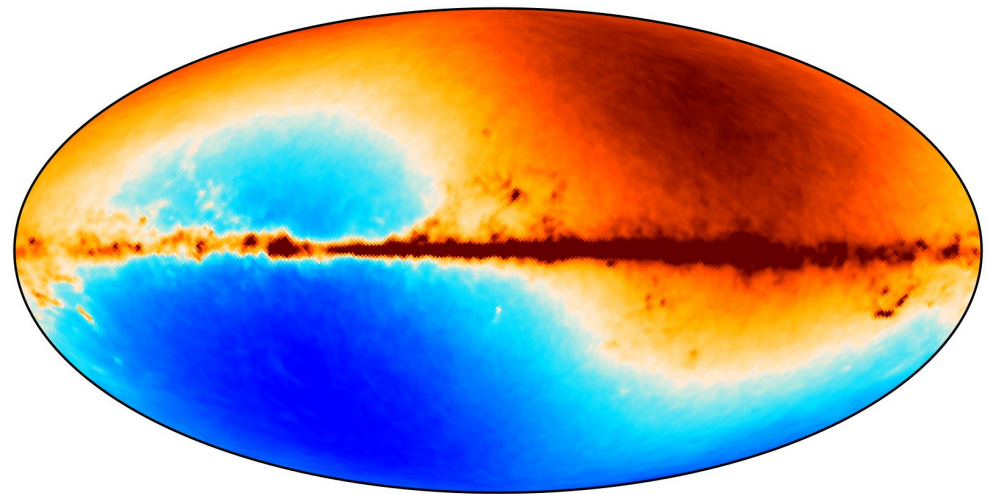
$$r \approx \frac{P_k^{\text{grav waves}}}{P_k^{\text{density waves}}}$$

What sort of precision is required for gravitational waves?

Expected signal



Actual sky



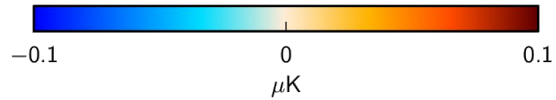
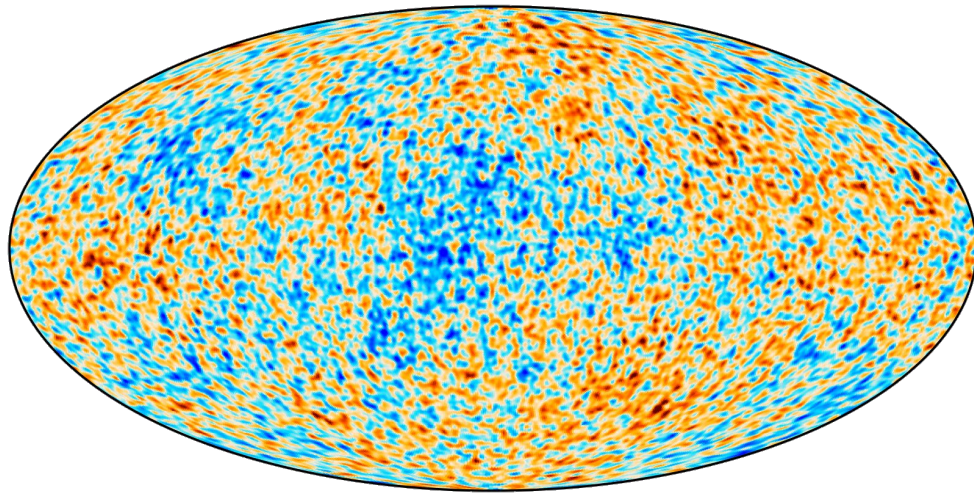
The sky is more than four orders of magnitude brighter than the signal!



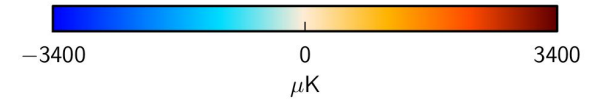
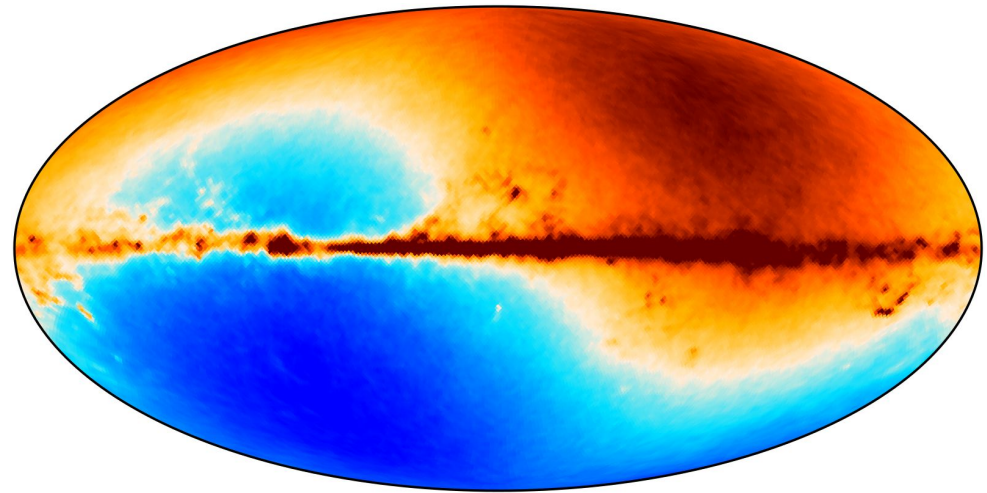
Need extremely accurate component separation
and control of instrumental systematic effects!

What sort of precision is required for gravitational waves?

Expected signal

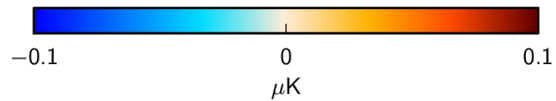
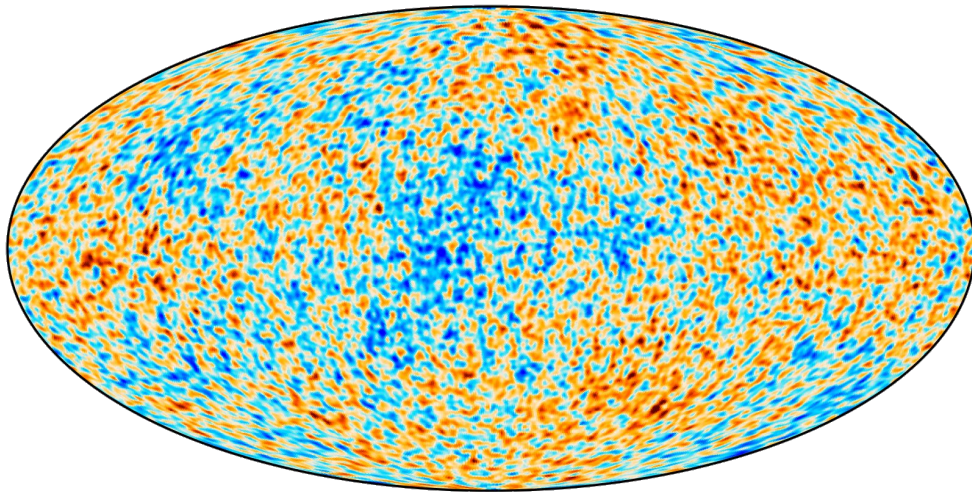


Actual sky

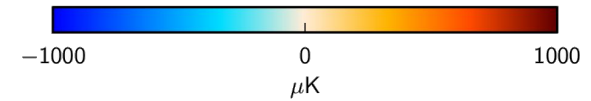
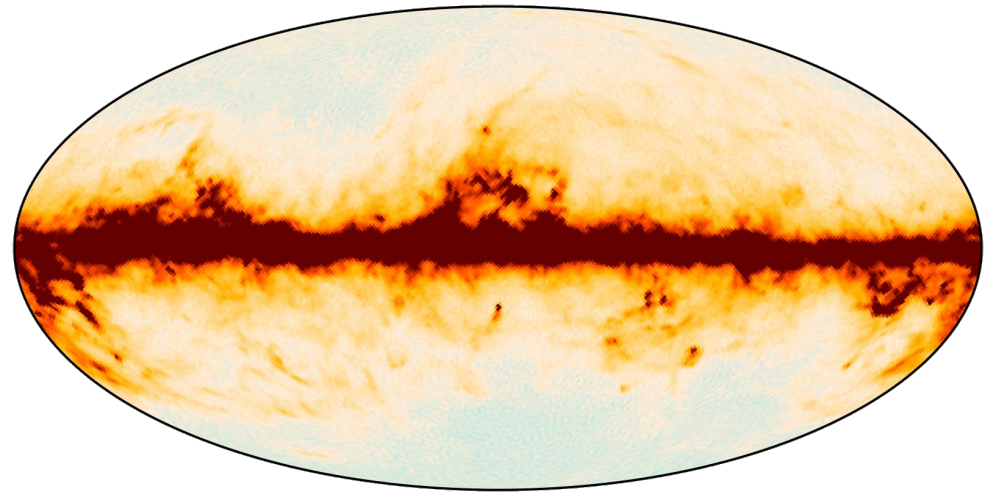


What sort of precision is required for gravitational waves?

Expected signal

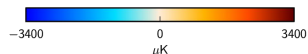
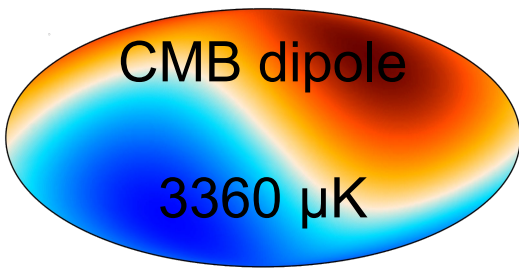


Actual sky - dipole



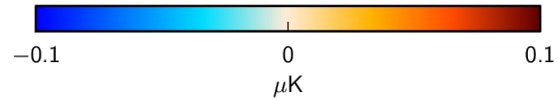
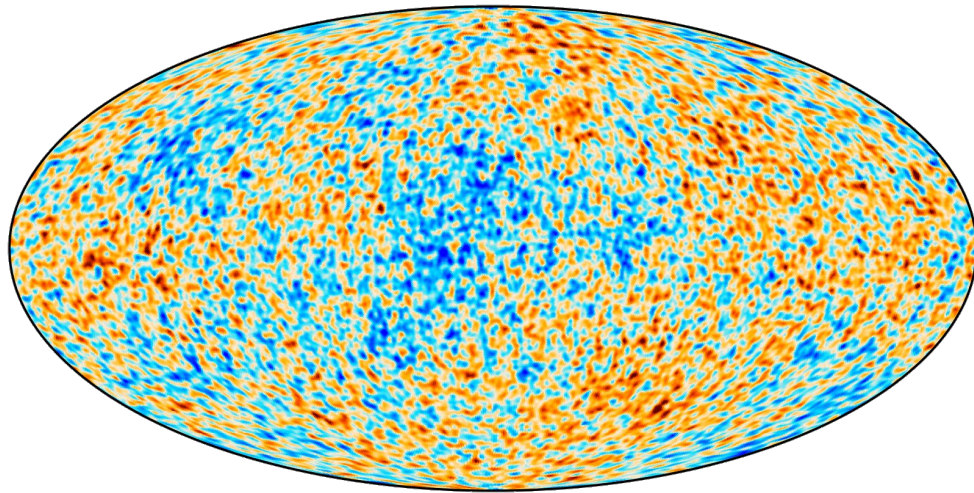
CMB dipole

3360 μK

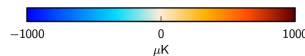
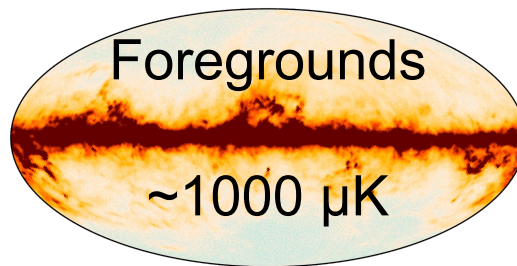
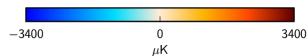
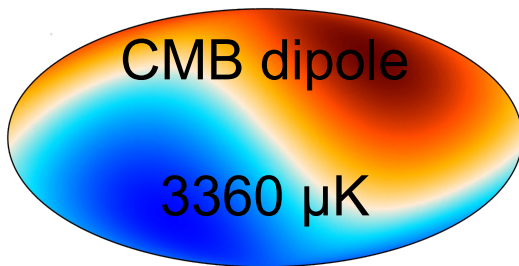
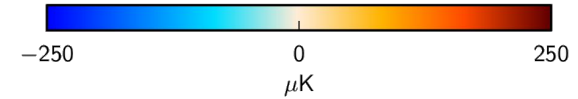
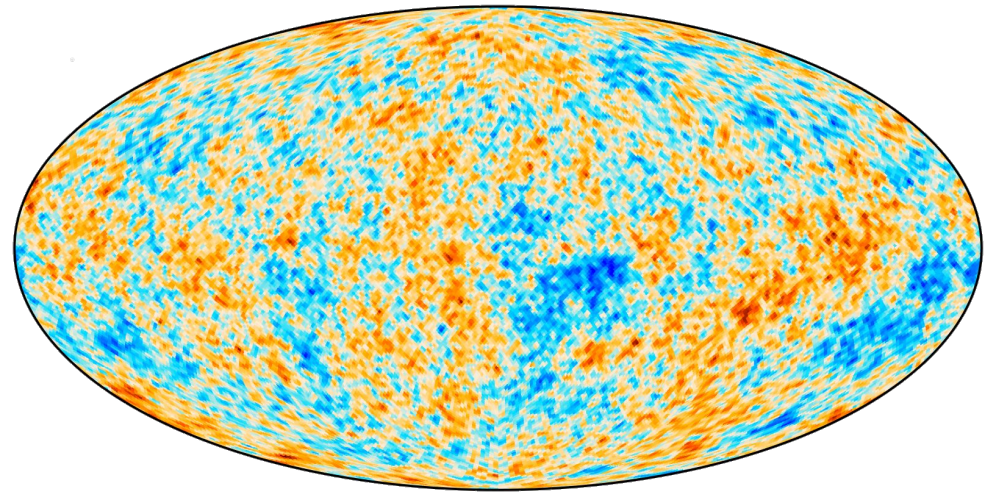


What sort of precision is required for gravitational waves?

Expected signal

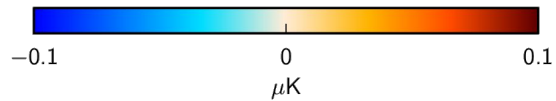
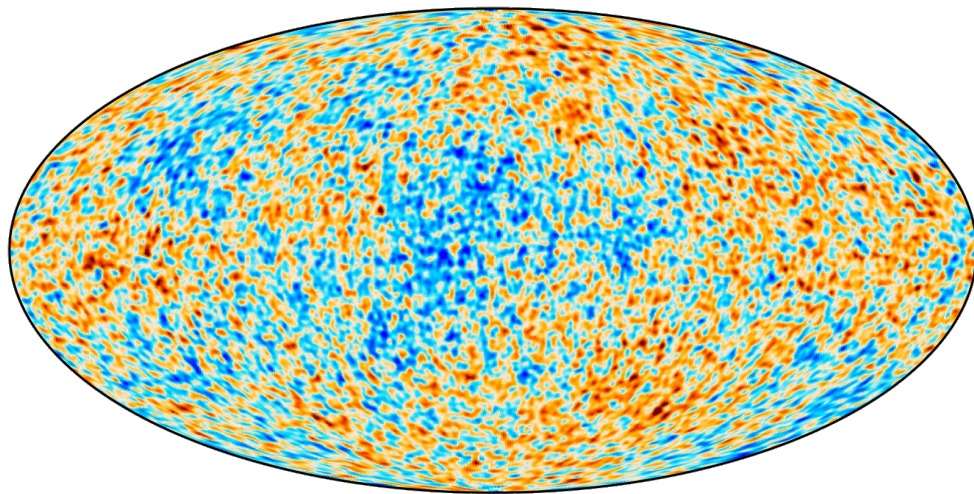


Actual sky - dipole - foregrounds

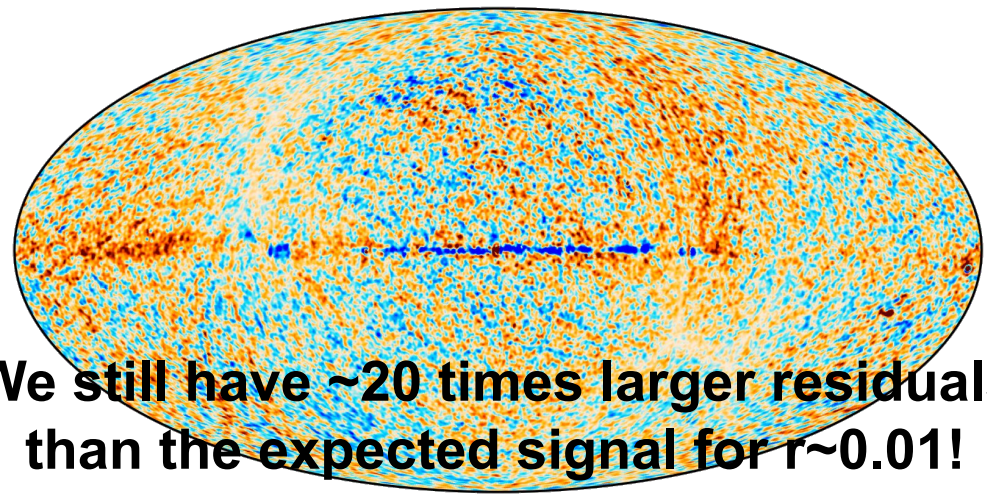


What sort of precision is required for gravitational waves?

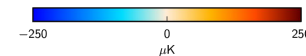
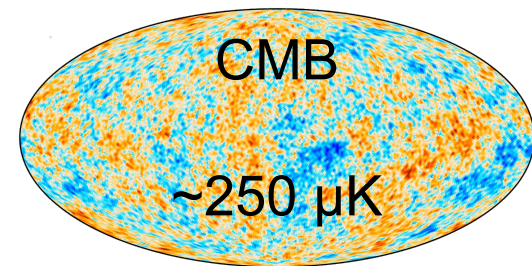
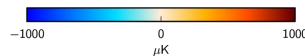
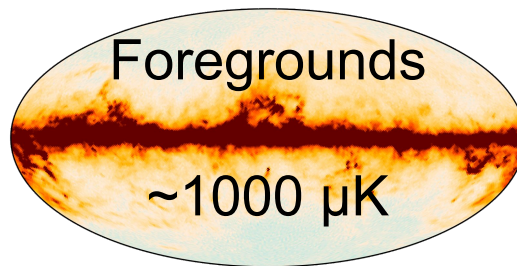
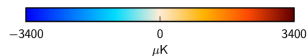
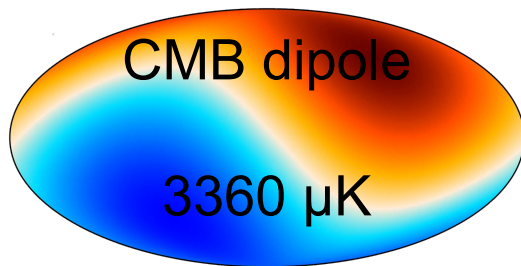
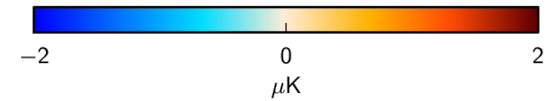
Expected signal



Actual sky - dipole - foregrounds



We still have ~20 times larger residuals than the expected signal for $r \sim 0.01$!



What sort of precision is required for gravitational waves?

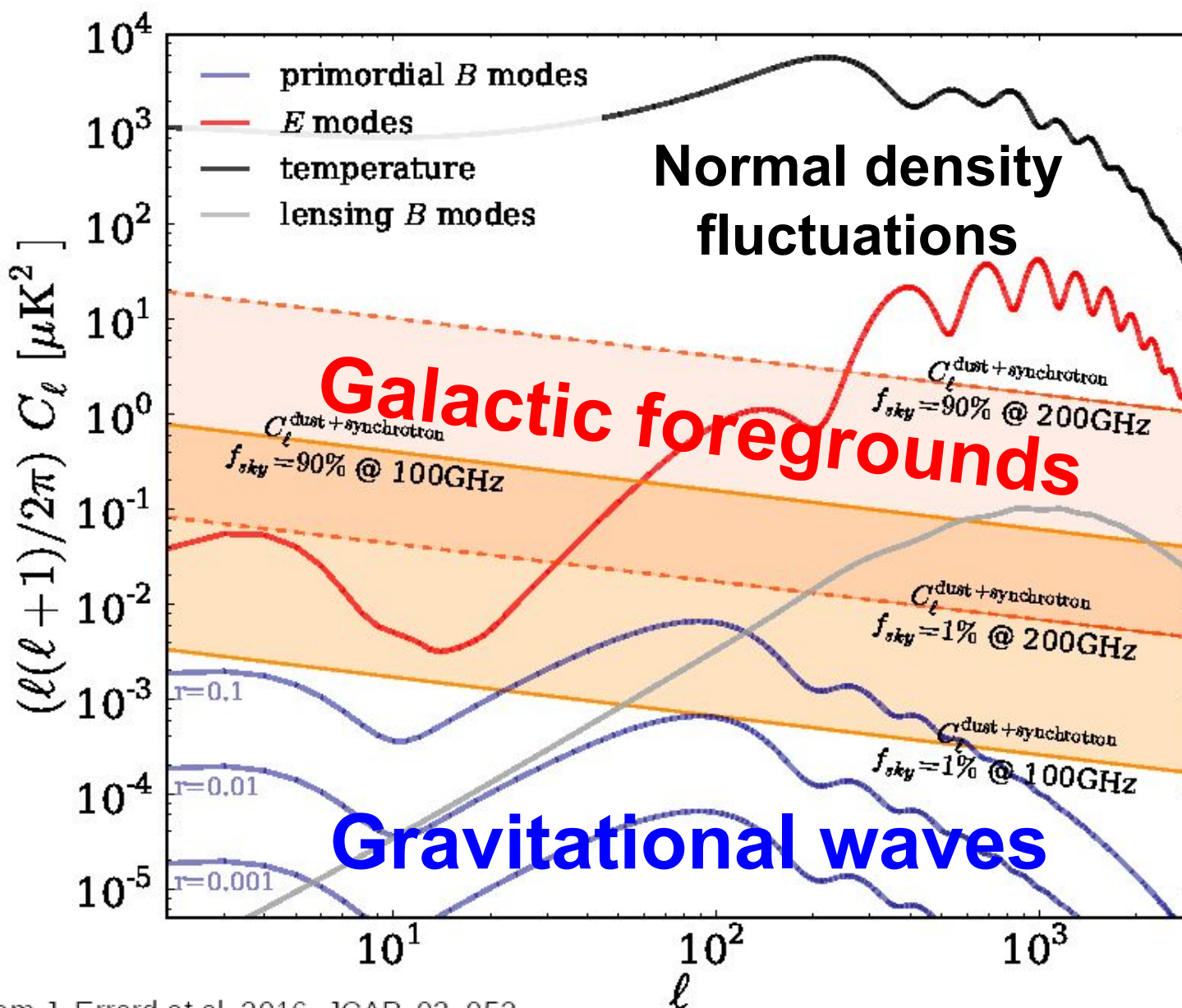


Figure from J. Errard et al. 2016, JCAP, 03, 052

What sort of precision is required for gravitational waves?

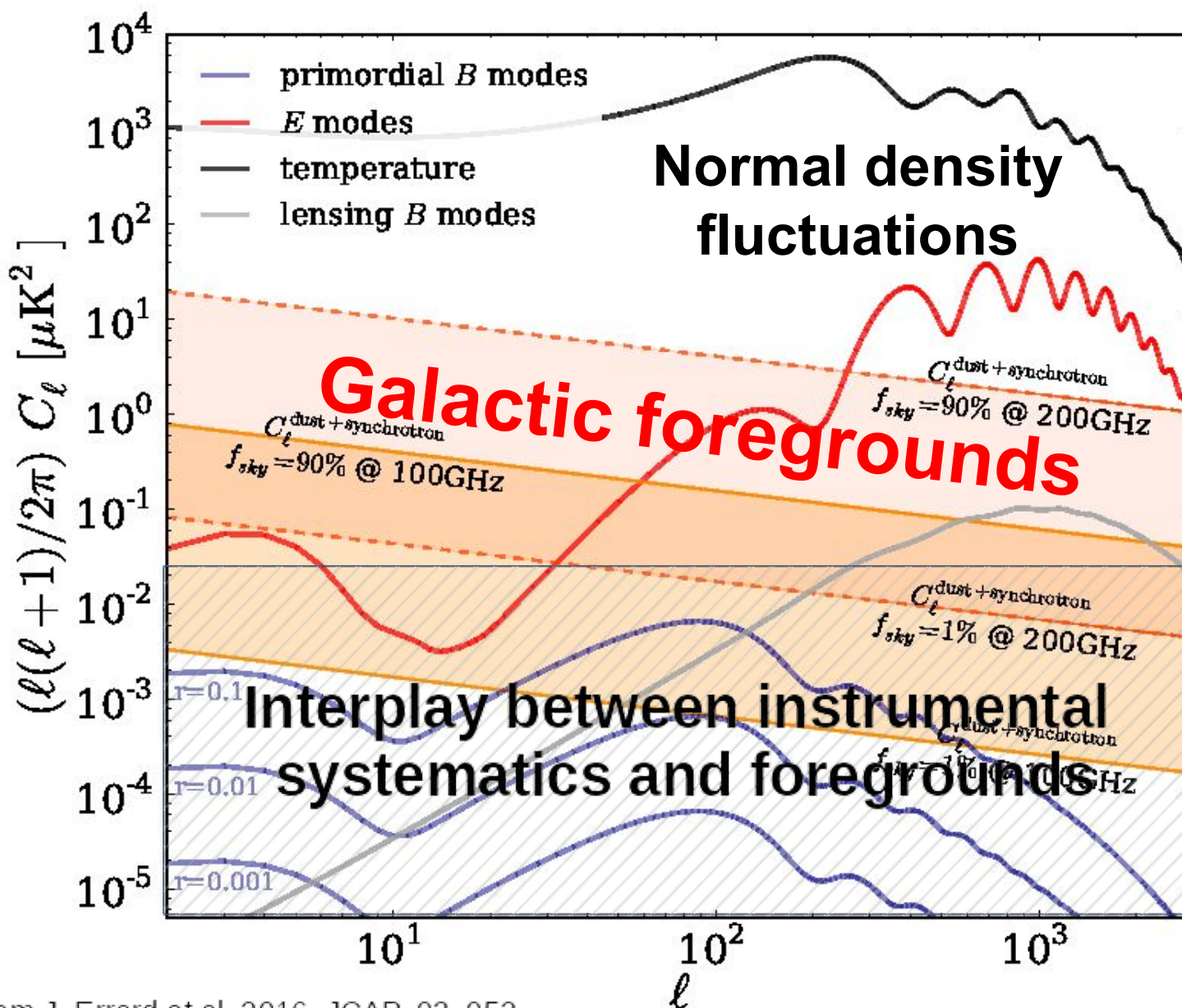
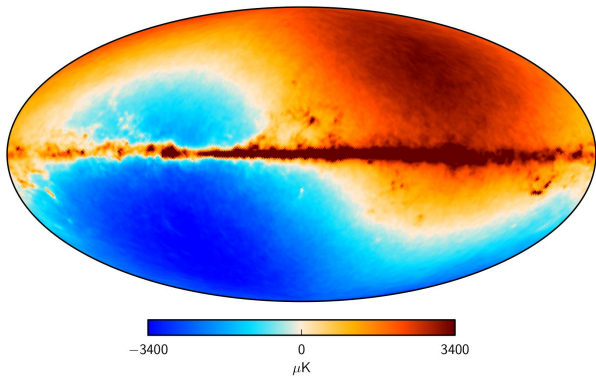


Figure from J. Errard et al. 2016, JCAP, 03, 052

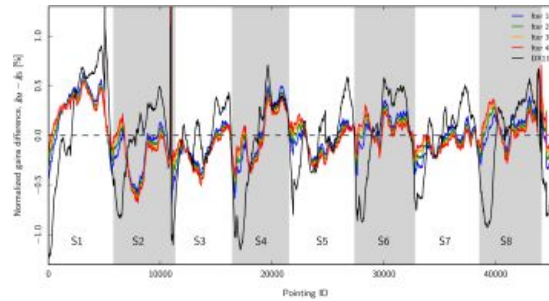
CMB's "chicken and egg" problem

Need to know the instrument to measure the sky...

Data

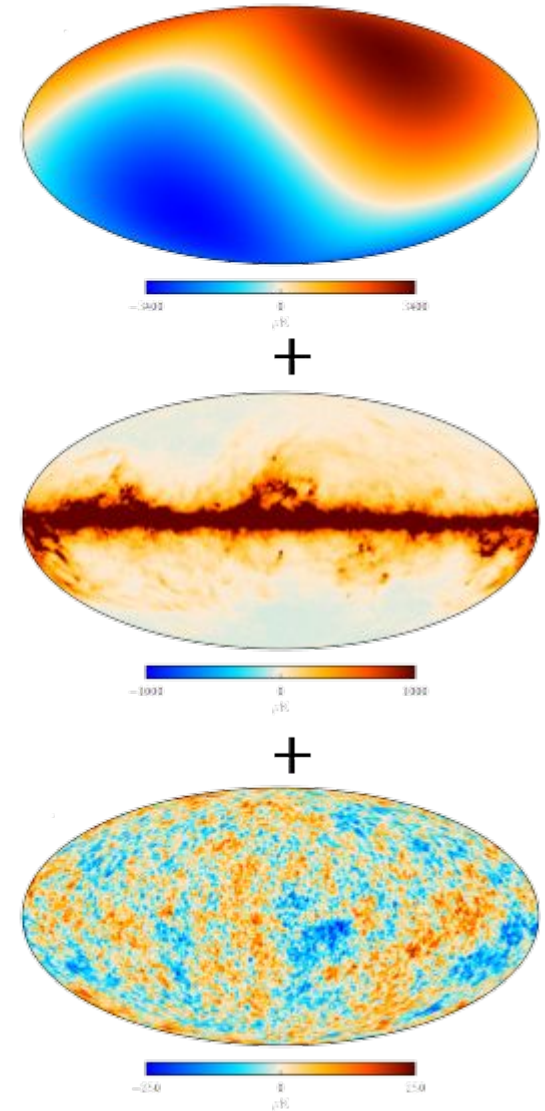


Instrument calibration

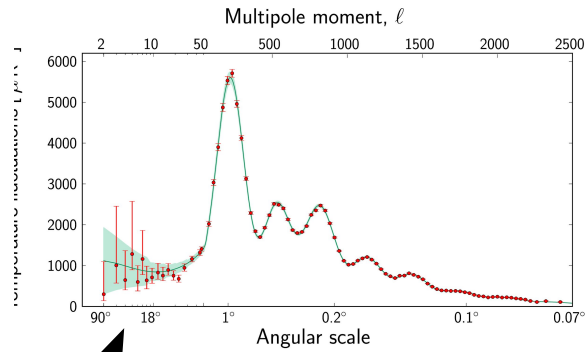


... but also need to know the sky in order to calibrate the instrument!

Sky

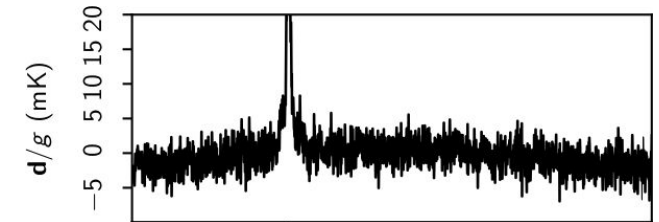
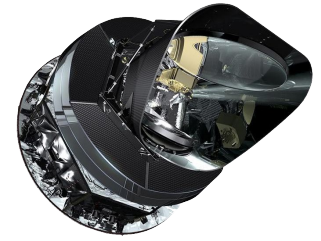
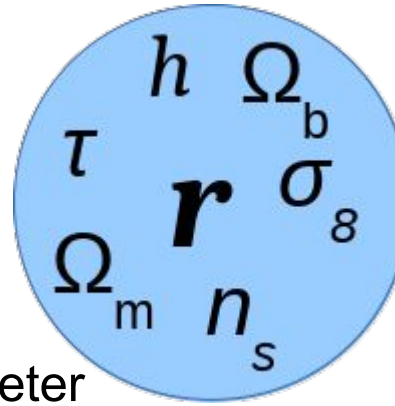


Classic CMB analysis



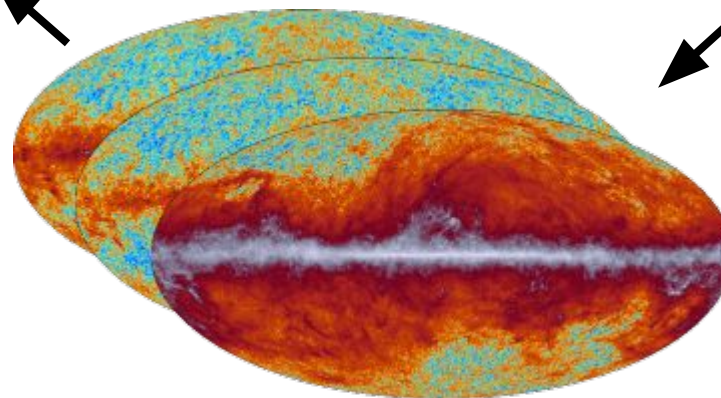
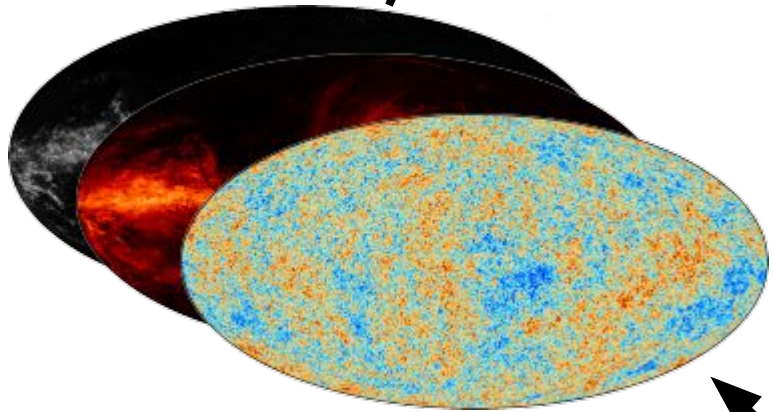
Power spectrum estimation

Parameter estimation

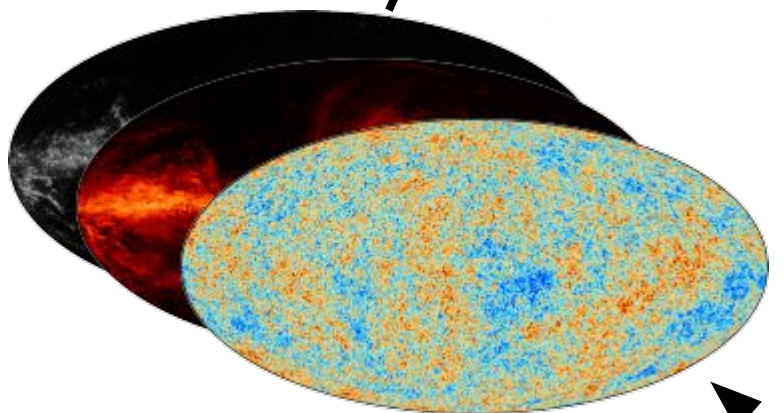
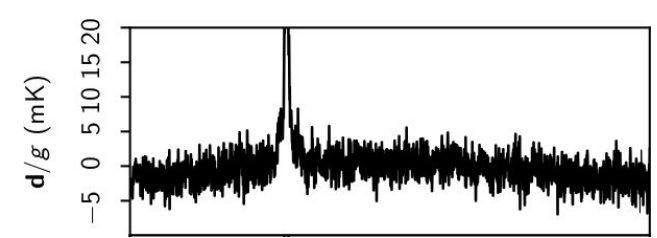
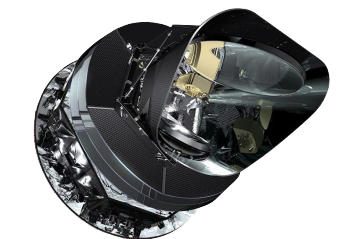
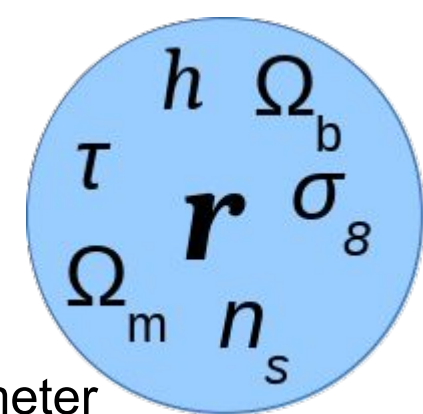
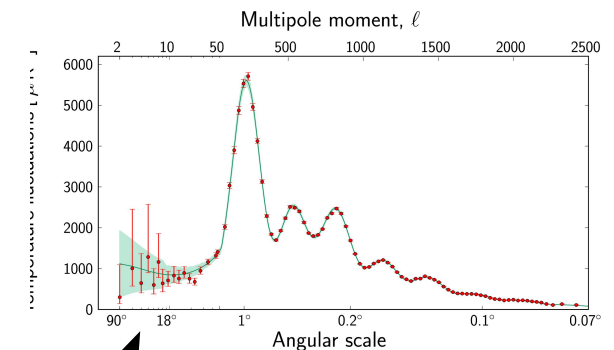


Calibration + mapmaking

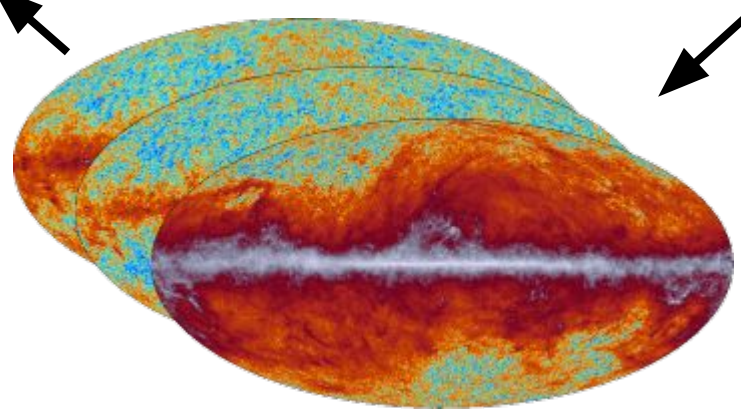
Component separation



End-to-end iterative analysis



Component separation



Main goals of the BeyondPlanck project:

- Implement an end-to-end analysis framework for current and future CMB experiments using Planck experience
- Demonstrate this framework with Planck LFI data
- Make software and results publicly available under an OpenSource license

The BeyondPlanck collaboration



EU-funded institutions



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Maksym Brilenkov
Hans Kristian Eriksen
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Marco Bersanelli
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Cristian Franceschet
Davide Maino
Aniello Mennella
Simone Paradiso

External collaborators



Brandon Hensley



Jeff Jewell



Reijo Keskitalo



Bruce Partridge



Martin Reinecke

1. Write down an explicit parametric model for the observed data:

$$d_{j,t} = g_{j,t} P_{tp,j} \left[\mathbf{B}_{pp',j}^{\text{symm}} \sum_c M_{cj}(\beta_{p'}, \Delta_{\text{bp}}^j) a_{p'}^c + \mathbf{B}_{j,t}^{\text{asymm}} (\mathbf{s}_j^{\text{orb}} + \mathbf{s}_t^{\text{fsl}}) \right] + n_{j,t}^{\text{corr}} + n_{j,t}^{\text{w}}.$$

Let $\omega = \{\text{all free parameters}\}$

2. Derive the joint posterior distribution with Bayes' theorem:

$$P(\omega | \mathbf{d}) = \frac{P(\mathbf{d} | \omega)P(\omega)}{P(\mathbf{d})} \propto \mathcal{L}(\omega)P(\omega).$$

3. Map out $P(\omega | \mathbf{d})$ with standard Markov Chain Monte Carlo (MCMC) methods

The BeyondPlanck data model



Data

Pointing

Bandpass

Sidelobe pickup

White noise

$$d_{j,t} =$$

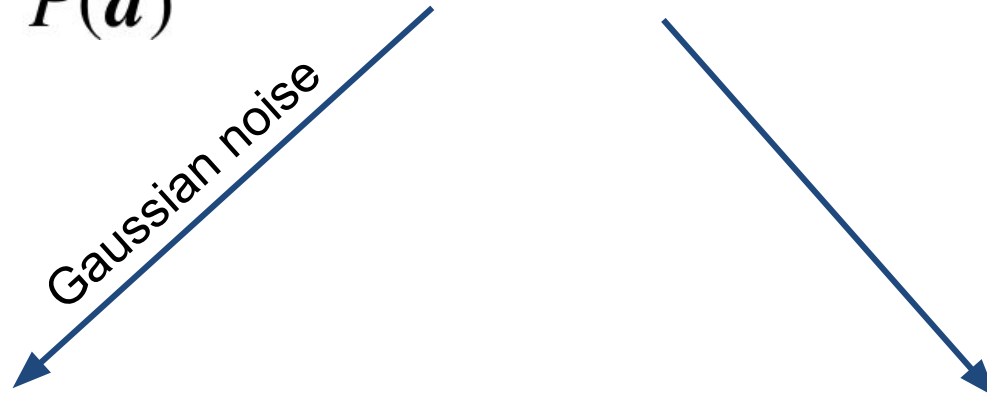
$$w_{i,t}$$

$$\omega \equiv \{g, \Delta_{\text{bp}}, \mathbf{n}_{\text{corr}}, \mathbf{a}_i, \beta_i, C_\ell, \dots\}$$

$$+ \sum_{j=1}^J \mathbf{a}_{\text{src}}^j \left(\frac{1}{\nu_{0,\text{src}}} \right)$$

Point sources

$$P(\omega | \mathbf{d}) = \frac{P(\mathbf{d} | \omega)P(\omega)}{P(\mathbf{d})} \propto \mathcal{L}(\omega)P(\omega).$$



$$\mathcal{L}(\omega) = \frac{e^{-\frac{1}{2}(\mathbf{d}-s(\omega))^t \mathbf{N}_{\text{wn}}^{-1}(\mathbf{d}-s(\omega))}}{\sqrt{|\mathbf{N}_{\text{wn}}|}}$$

- $P(f_{\text{knee}})$ = lognorm(DPC, 0.1)
- $P(\beta_{\text{synch}})$ = -3.1 ± 0.1
- $P(T_{\text{dust}})$ = $\delta(T_{\text{dust}} - T_{\text{dust, HFI}})$
- $P(a_{\text{ff}})$ = $N(a_{\text{ff, Planck}}, \sigma_{l, \text{ff}}^2)$
- $P(a_{\text{ame}})$ = $N(\alpha \cdot m_{857}, \sigma_{l, \text{ame}}^2)$

⋮

How to sample from *big* distributions?



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The Free Encyclopedia

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Gibbs sampling

From Wikipedia, the free encyclopedia

In [statistics](#), **Gibbs sampling** or a **Gibbs sampler** is a [Markov chain Monte Carlo \(MCMC\) algorithm](#) for obtaining a sequence of observations which are approximated from a specified [multivariate probability distribution](#), when direct sampling is difficult. This sequence can be used to approximate the joint distribution (e.g., to generate a histogram of the distribution); to approximate the [marginal distribution](#) of one of the variables, or some subset of the variables (for example, the unknown [parameters](#) or [latent variables](#)); or to compute an [integral](#) (such as the [expected value](#) of one of the variables). Typically, some of the variables correspond to observations whose values are known, and hence do not need to be sampled.

Part of a series on

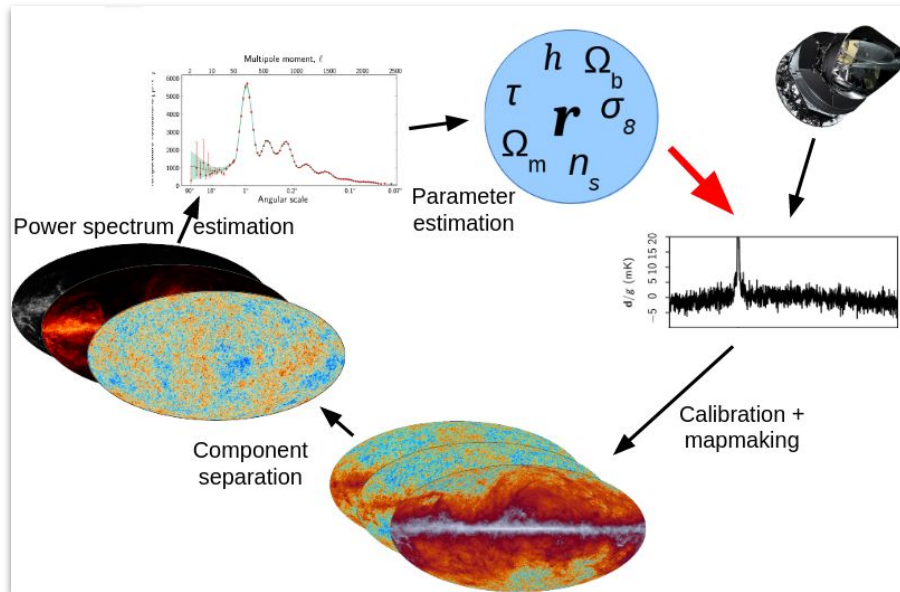
Bayesian statistics



Theory

- Admissible decision rule
- Bayesian efficiency
- Bayesian probability
- Probability interpretations
- Bayes' theorem
- Bayes factor
- Bayesian inference
- Bayesian network
- Prior
- Posterior
- Likelihood
- Conjugate prior
- Posterior predictive
- Hyperparameter
- Hyperprior

What we want to do:



How we actually do it:

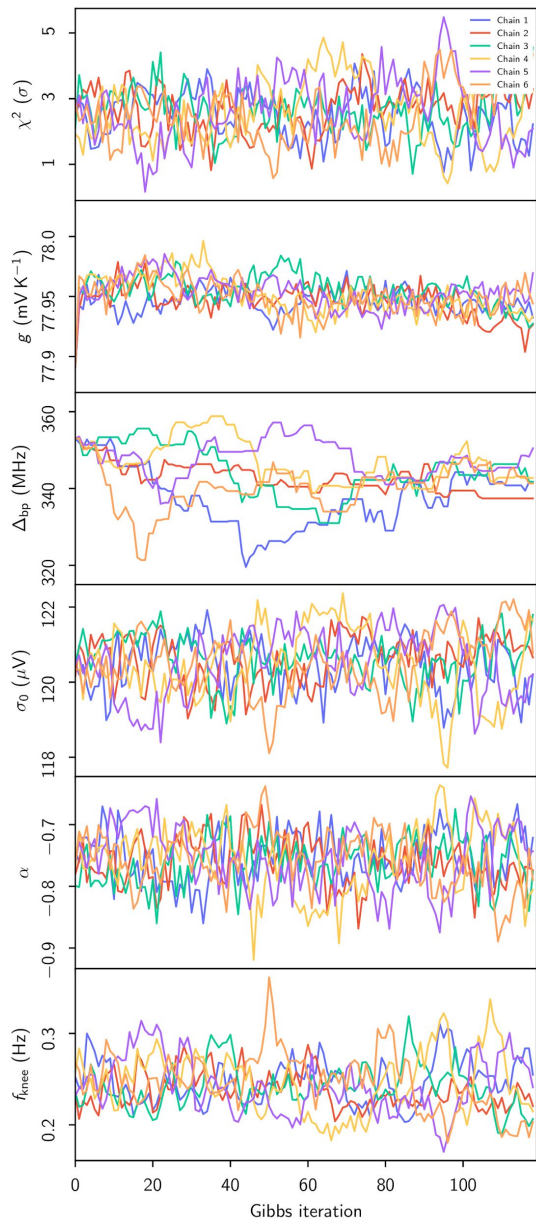
$$\begin{aligned}
 \mathbf{g} &\leftarrow P(\mathbf{g} \mid \mathbf{d}, \xi_n, \Delta_{\text{bp}}, \mathbf{a}, \beta, C_\ell) \\
 \mathbf{n}_{\text{corr}} &\leftarrow P(\mathbf{n}_{\text{corr}} \mid \mathbf{d}, \mathbf{g}, \xi_n, \Delta_{\text{bp}}, \mathbf{a}, \beta, C_\ell) \\
 \xi_n &\leftarrow P(\xi_n \mid \mathbf{d}, \mathbf{g}, \mathbf{n}_{\text{corr}}, \Delta_{\text{bp}}, \mathbf{a}, \beta, C_\ell) \\
 \Delta_{\text{bp}} &\leftarrow P(\Delta_{\text{bp}} \mid \mathbf{d}, \mathbf{g}, \mathbf{n}_{\text{corr}}, \xi_n, \mathbf{a}, \beta, C_\ell) \\
 \beta &\leftarrow P(\beta \mid \mathbf{d}, \mathbf{g}, \mathbf{n}_{\text{corr}}, \xi_n, \Delta_{\text{bp}}, C_\ell) \\
 \mathbf{a} &\leftarrow P(\mathbf{a} \mid \mathbf{d}, \mathbf{g}, \mathbf{n}_{\text{corr}}, \xi_n, \Delta_{\text{bp}}, \beta, C_\ell) \\
 C_\ell &\leftarrow P(C_\ell \mid \mathbf{d}, \mathbf{g}, \mathbf{n}_{\text{corr}}, \xi_n, \Delta_{\text{bp}}, \mathbf{a}, \beta)
 \end{aligned}$$



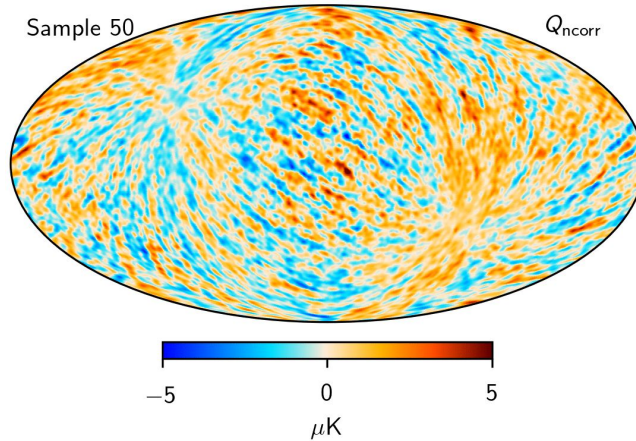
REFERENCE	TITLE
<i>Pipeline</i>	
BeyondPlanck Collaboration (2020) . . .	I. Global Bayesian analysis of the <i>Planck</i> Low Frequency Instrument data
Keihänen et al. (2020)	II. CMB mapmaking through Gibbs sampling
Galloway et al. (2020a)	III. Computational infrastructure and Commander3
Brilenkov et al. (2020)	IV. Time-ordered data simulations
Gerakakis et al. (2020)	V. Open Science and reproducibility
<i>Instrument characterization</i>	
Ihle et al. (2020)	VI. Noise characterization and modelling
Gjerløw et al. (2020)	VII. Calibration
Galloway et al. (2020b)	VIII. Sidelobe corrections
Svalheim et al. (2020a)	IX. Bandpass and beam leakage corrections
<i>Cosmological and astrophysical results</i>	
Suur-Uski et al. (2020)	X. LFI frequency map posteriors
Colombo et al. (2020)	XI. CMB constraints
Paradiso et al. (2020)	XII. Cosmological parameter estimation with end-to-end error propagation
Andersen et al. (2020)	XIII. Intensity foregrounds, degeneracies and priors
Svalheim et al. (2020b)	XIV. Polarized synchrotron emission
Herman et al. (2020)	XV. Limits on polarized anomalous microwave emission
<i>External analysis</i>	
Aurlien et al. (2020)	XVI. Application to simulated <i>LiteBIRD</i> observations
Watts et al. (2020)	XVII. Application to <i>WMAP</i>
Galeotta et al. (2020)	XVIII. End-to-end validation of BEYONDPLANCK

- To highlight the method itself, only the following data are included in the current analysis:
 - ***Planck LFI 30, 44 and 70 GHz time-ordered data***
 - ***Planck 857 GHz*** to constrain thermal dust intensity
 - ***Planck 353 GHz*** polarization-only to constrain thermal dust polarization
 - ***WMAP 33-61 GHz*** in T+P to constrain low-frequency foregrounds
 - ***Haslam 408 MHz*** to constrain synchrotron intensity
- Intermediate *Planck HFI* and *WMAP 23 GHz* data are ***not*** included, because they have higher signal-to-noise ratios than Planck LFI

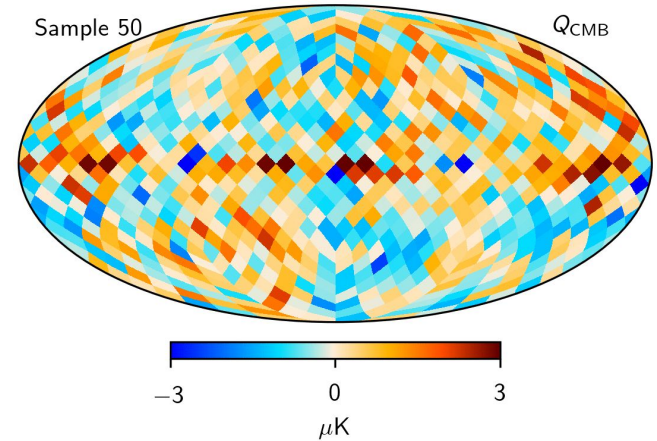
Instrument



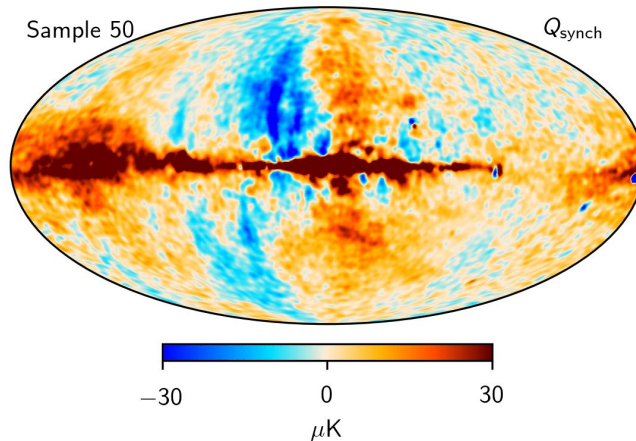
Correlated noise



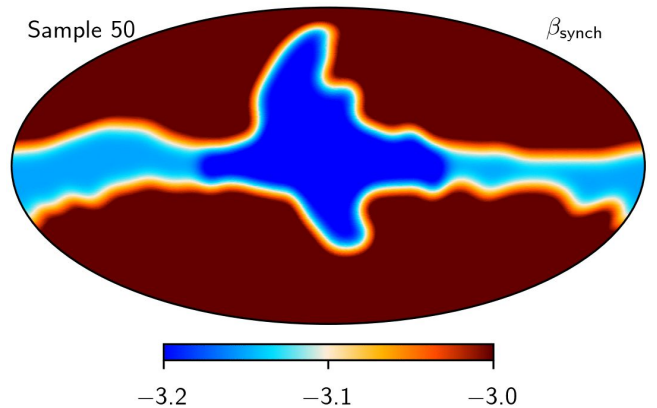
CMB Stokes Q



Synch Stokes Q



Synch pol β



...

Computational resource requirements

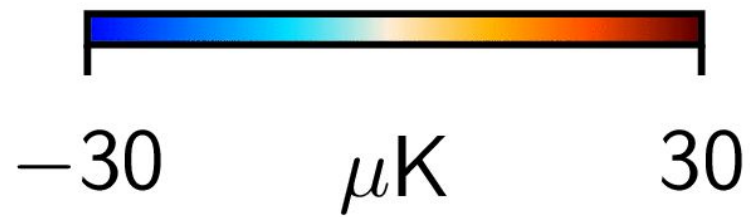
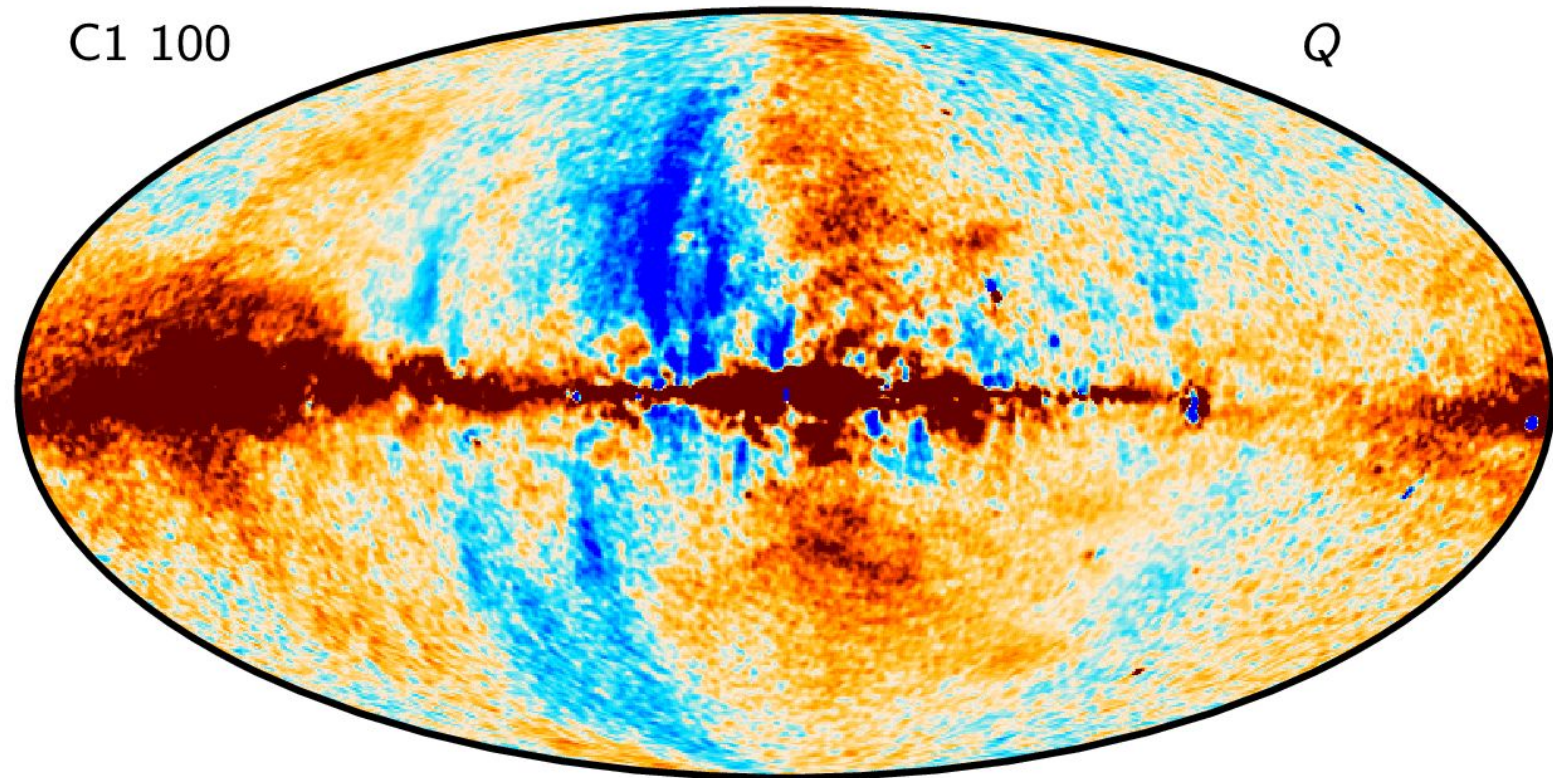
ITEM	30 GHz	44 GHz	70 GHz	SUM
<i>Data volume</i>				
Uncompressed data volume	761 GB	1 633 GB	5 522 GB	7 915 GB
Compressed data volume/RAM requirements	86 GB	178 GB	597 GB	861 GB
<i>Processing time (cost per run)</i>				
TOD initialization/IO time	176 sec	288 sec	753 sec	1217 sec
Other initialization				663 sec
Total initialization				1880 sec
<i>Gibbs sampling steps</i>				
Data decompression				393 sec
TOD projection				330 sec
Sidelobe evaluation				480 sec
Orbital dipole				449 sec
Gain sampling				94 sec
Correlated noise				3138 sec
TOD binning				498 sec
Loss due to point sources				502 sec
Sum of other TOD steps				306 sec
TOD processing cost per sample				6396 sec
Amplitude sampling, $P(\mathbf{a} \mathbf{d}, \omega \setminus \mathbf{a})$				527 sec
Spectral index sampling, $P(\beta \mathbf{d}, \omega \setminus \beta)$				1080 sec
Other steps				149 sec
Total cost per sample				8168 sec

2.3 hours/sample
on
72-core node with 1.5 TB RAM

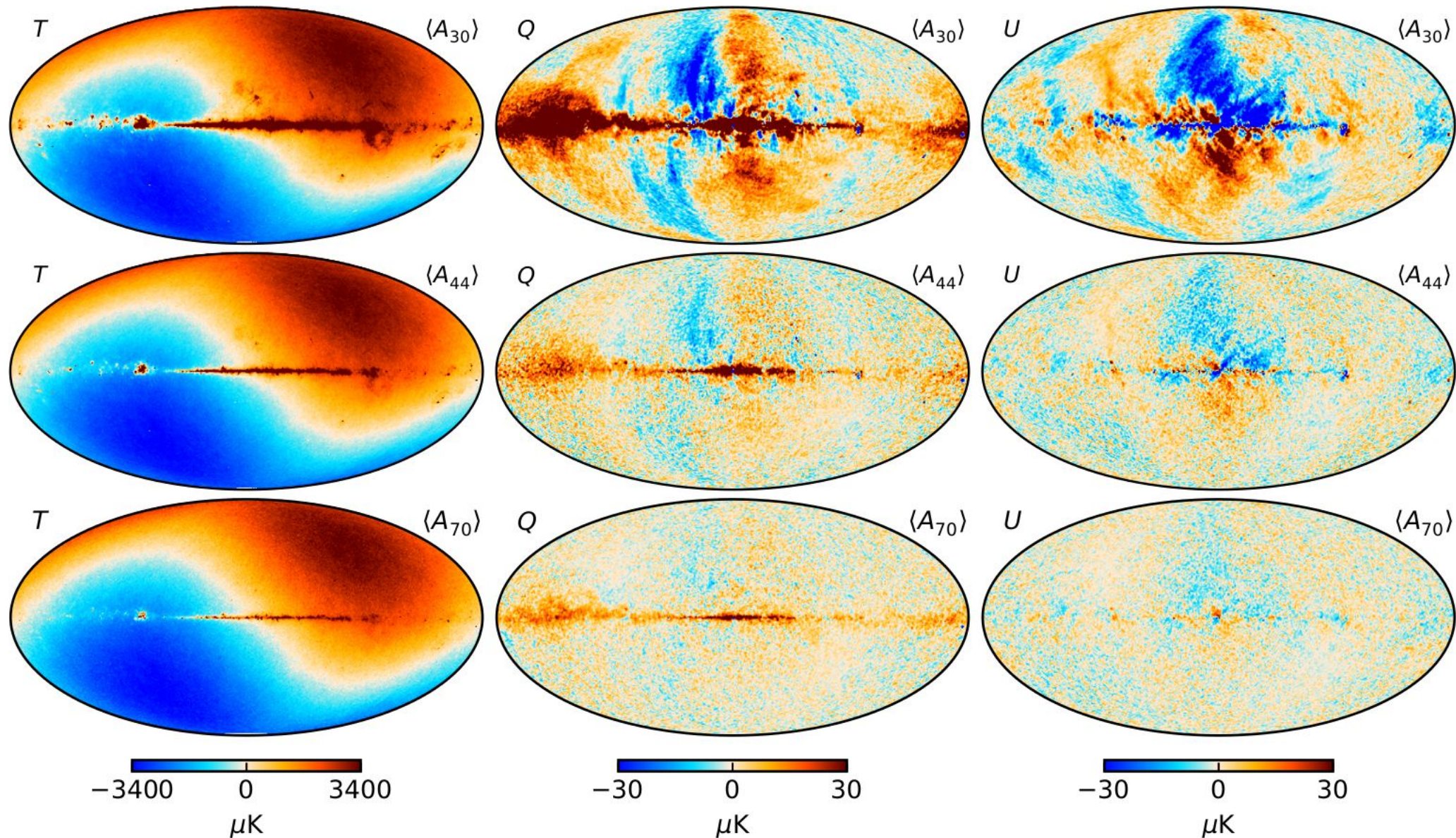
Galloway et al. (2020)

- **Six independent Gibbs chains of each 200 samples** were generated on 6 compute nodes
- Total wall production time for main run was **3 weeks**
- Total CPU cost for main run was **220,000 CPU hours**
 - For comparison, simulating one single traditional Planck Full Focal Plane 70 GHz realization costs $O(10^4)$ CPU hours (Planck Collaboration 2016, A&A, 596, A12)

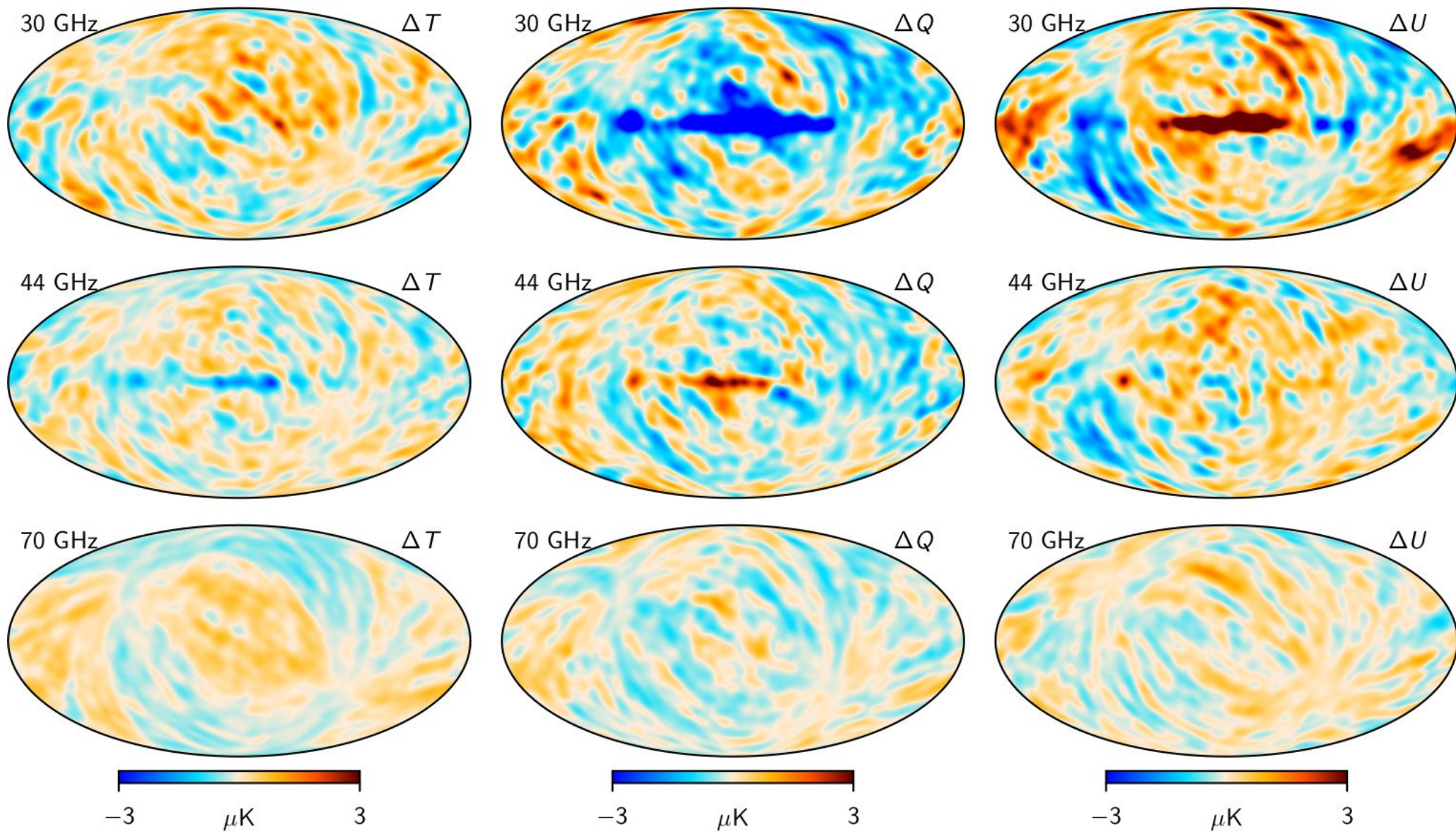
Frequency maps: 30 GHz Stokes Q



Frequency maps: Posterior mean

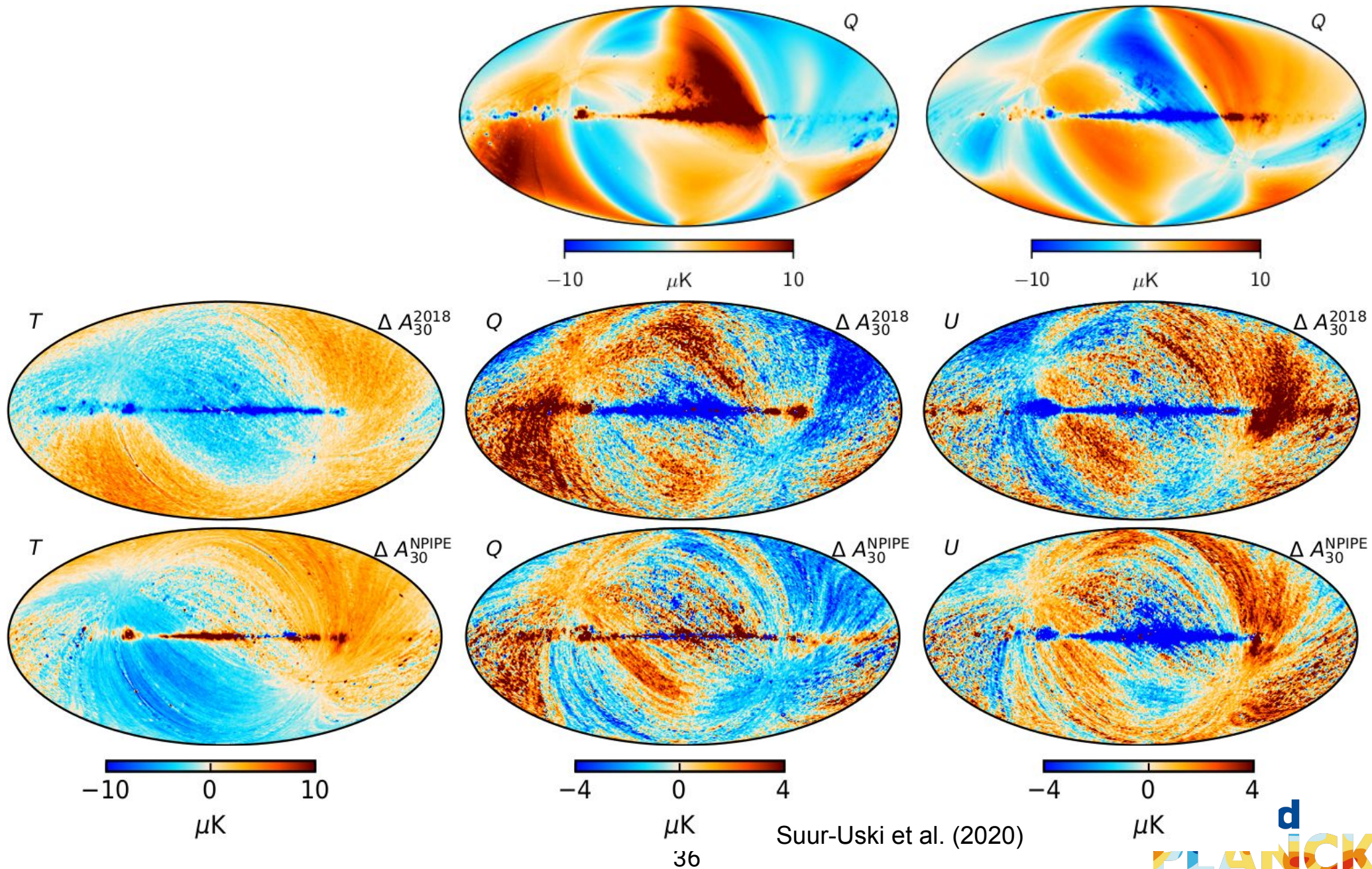


Frequency maps: Difference between two samples

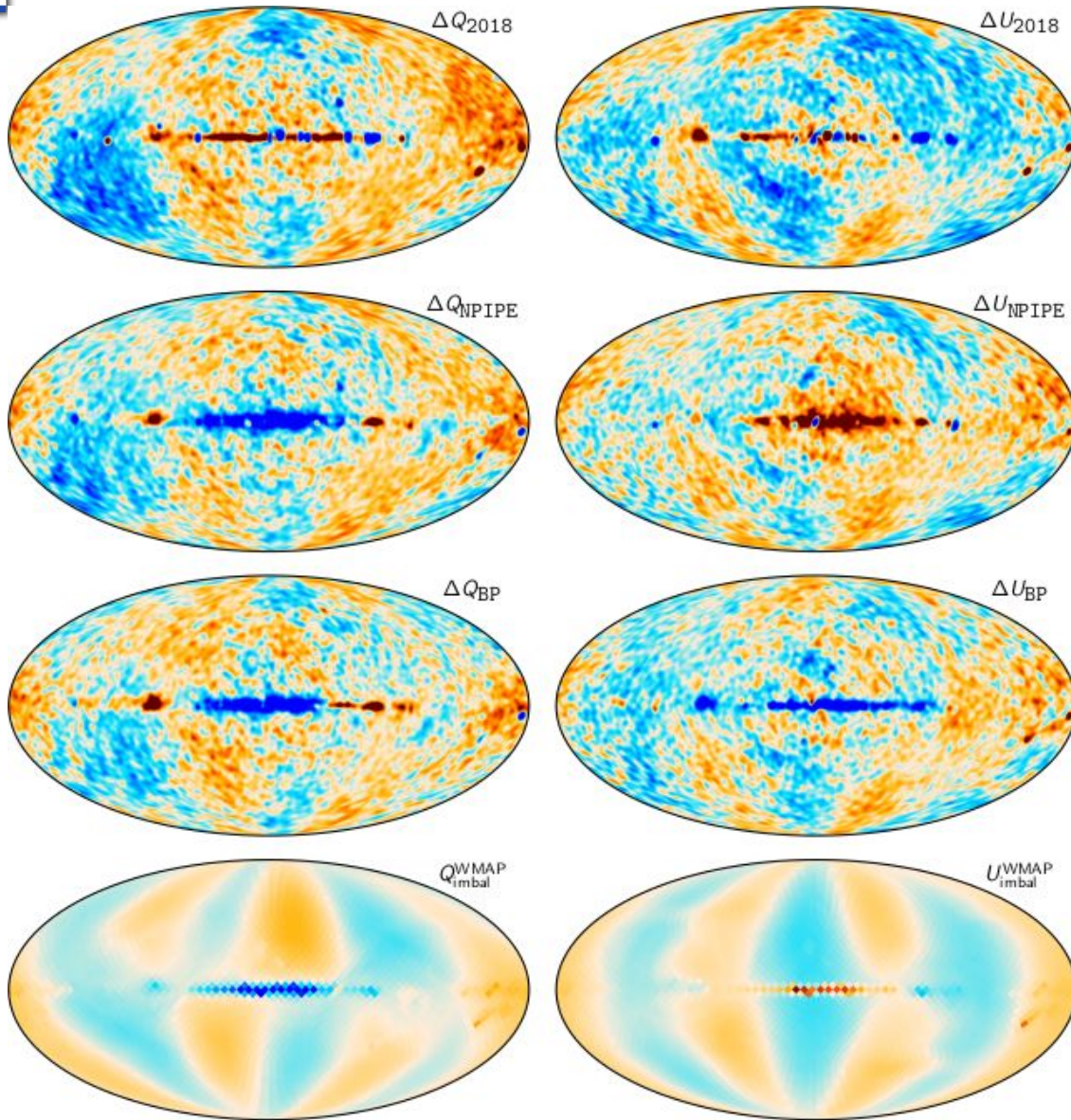


Frequency maps: 30 GHz minus NPIPE/Planck 2018

LFI DPC gain template (Planck Collaboration 2020, A&A, 641, A2)



Frequency maps: 30 GHz minus WMAP K-band

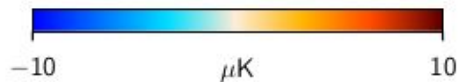


Planck 2018

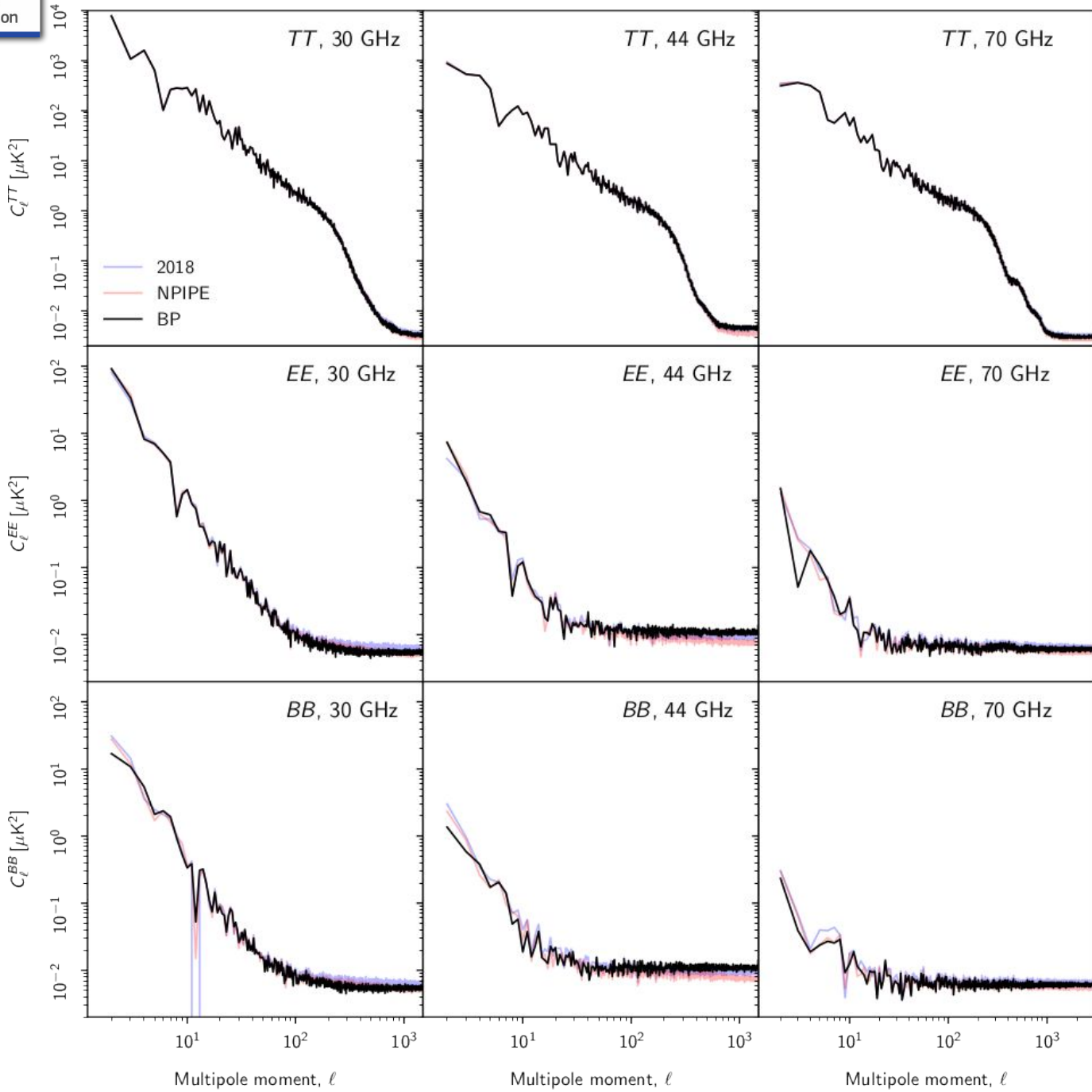
NPIPE

BeyondPlanck

WMAP transmission imbalance template (Jarosik et al. 2007)



Frequency maps: Power spectrum



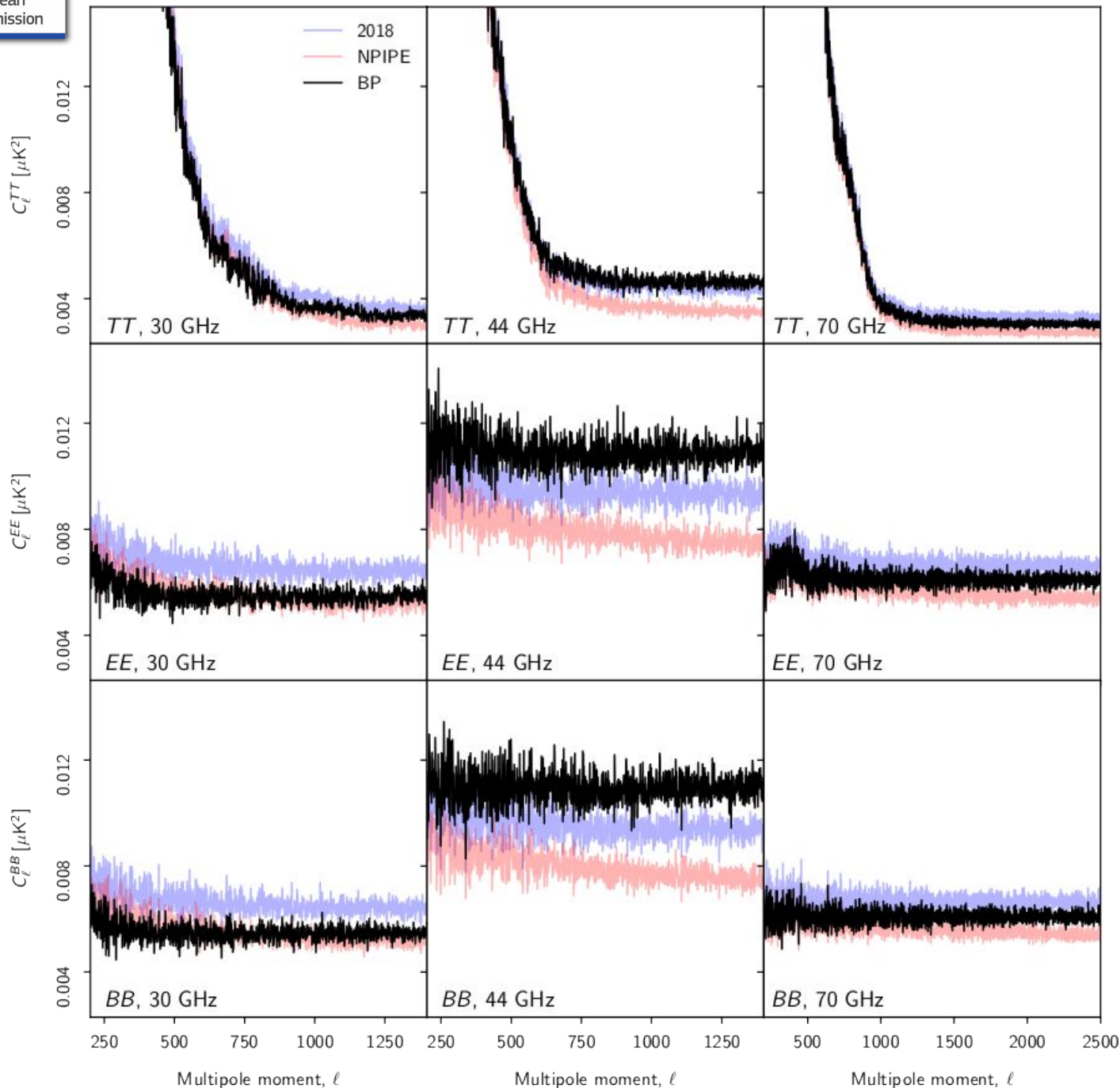
Suur-Uski et al. (2020)



Frequency maps: Power spectrum



European Commission



Flatter spectrum

=

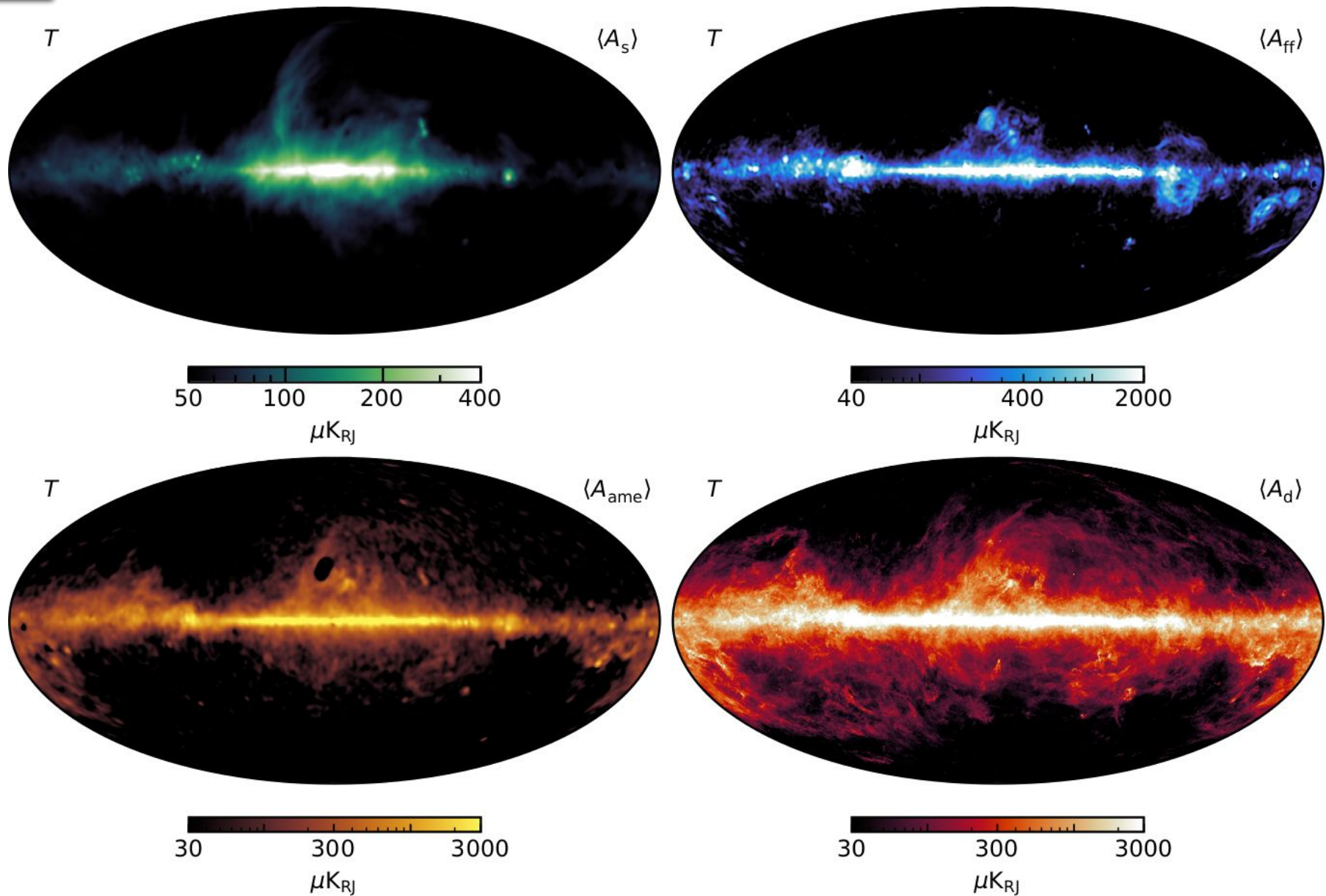
Less correlated noise due to joint multi-frequency signal estimation

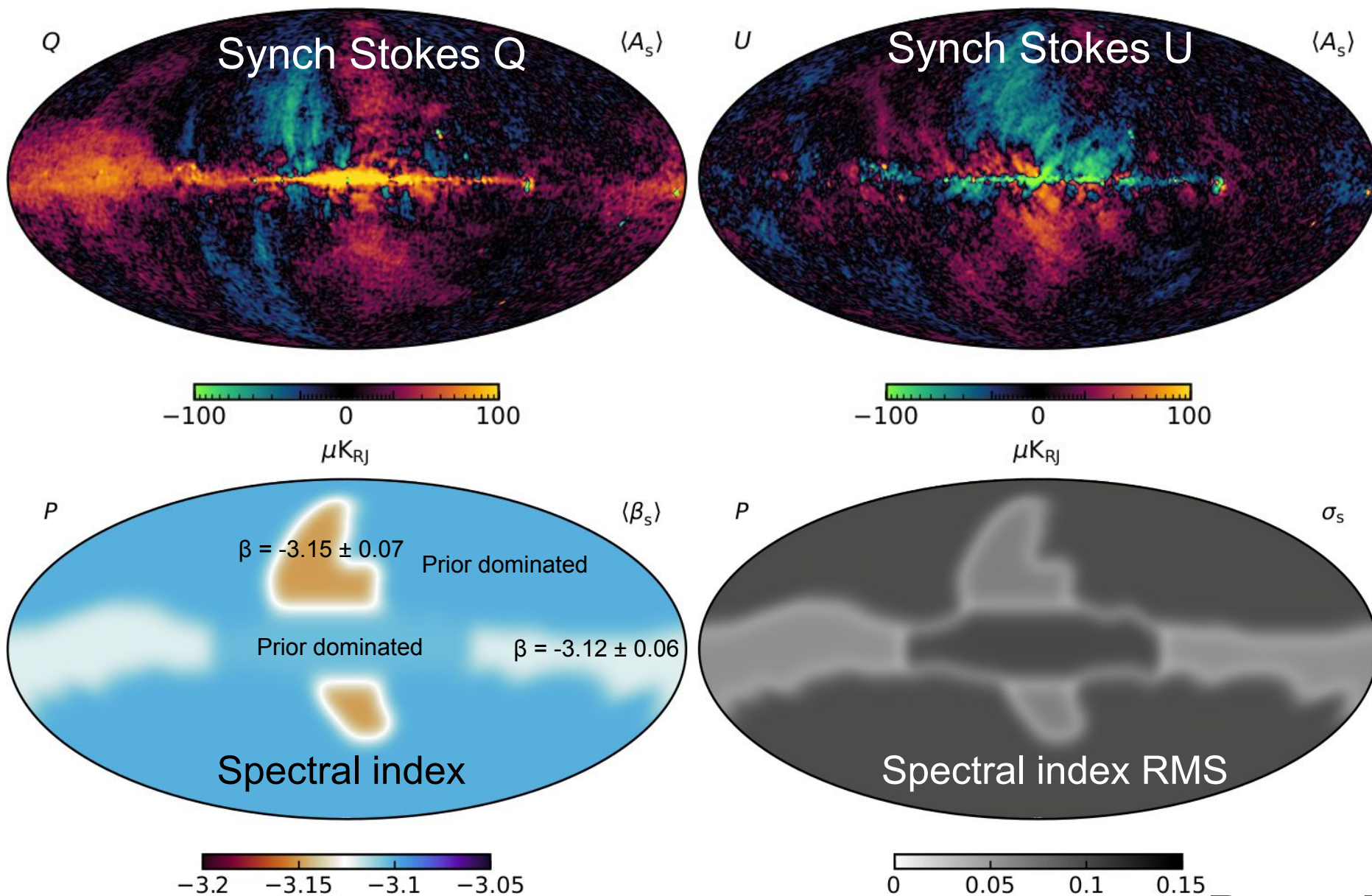
Higher white noise at 44 GHz

because we discard more data

Suur-Uski et al. (2020)

Astrophysical foregrounds: Temperature sky



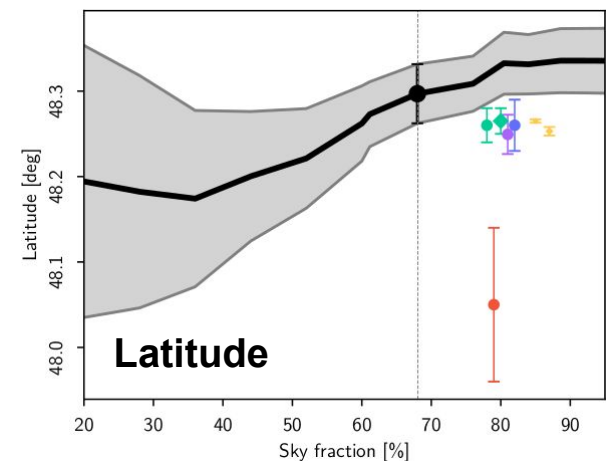
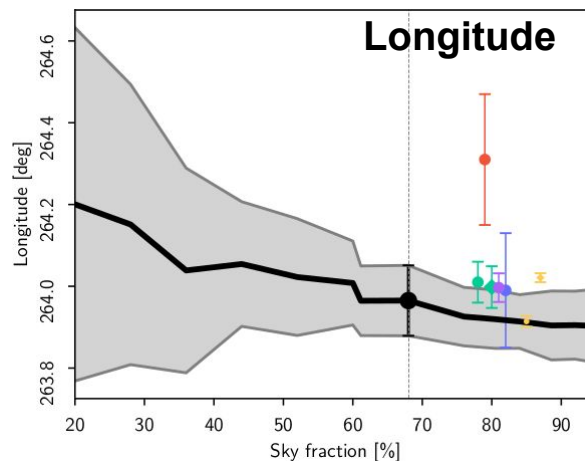
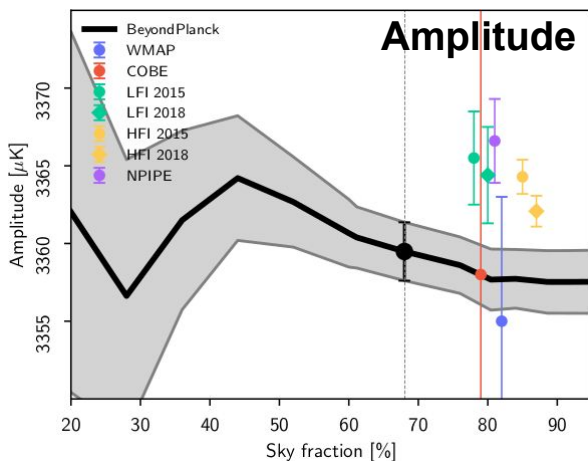


CMB: Solar dipole

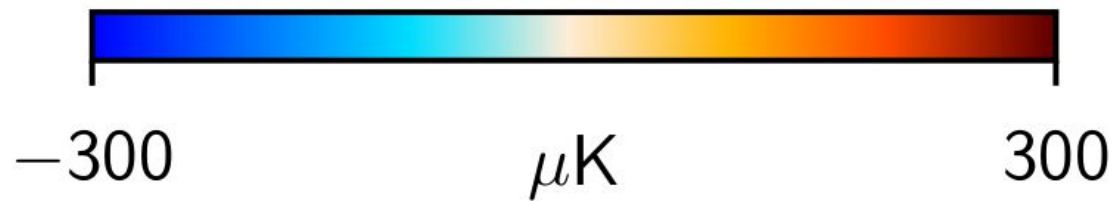
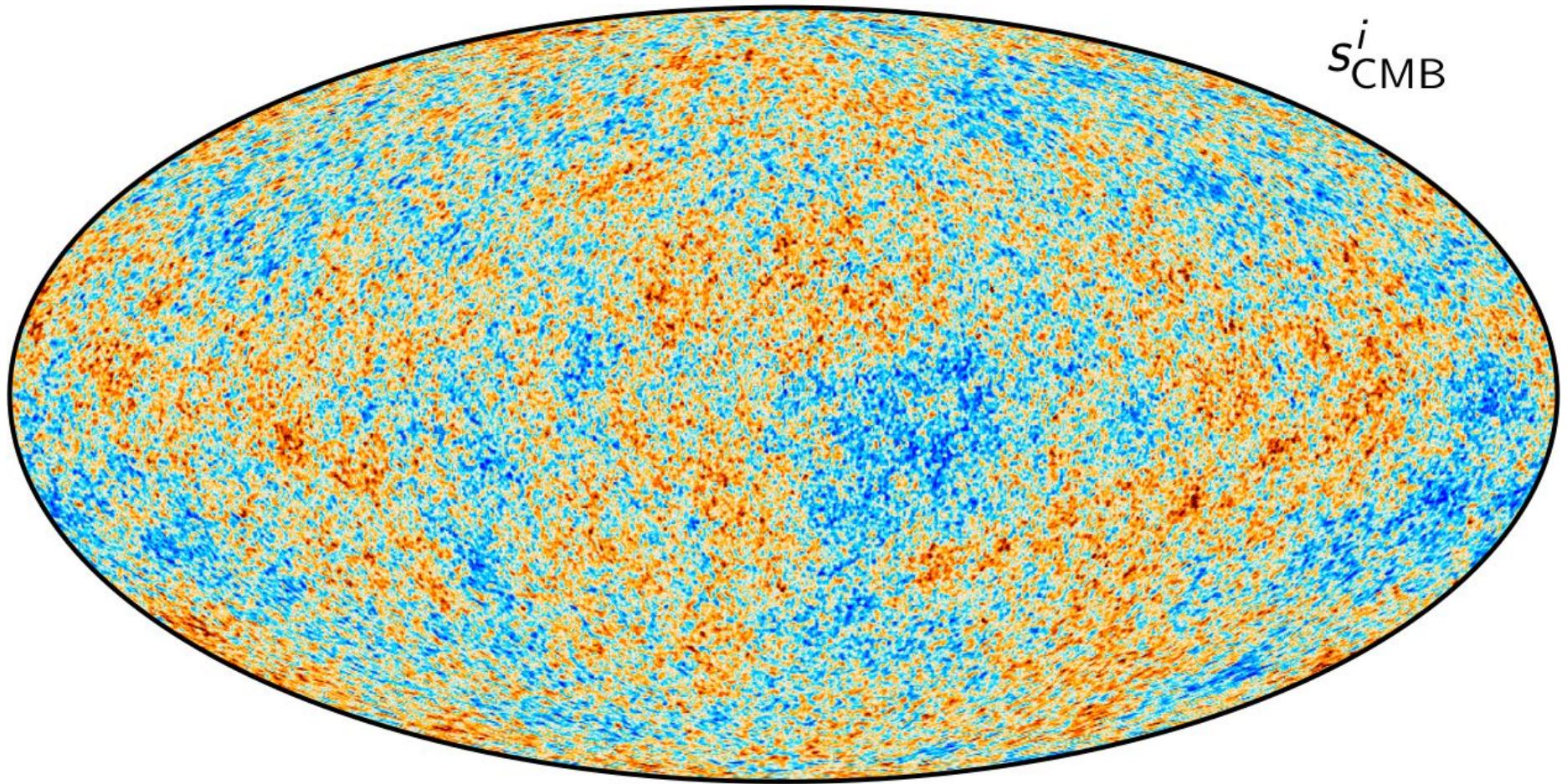


GALACTIC COORDINATES

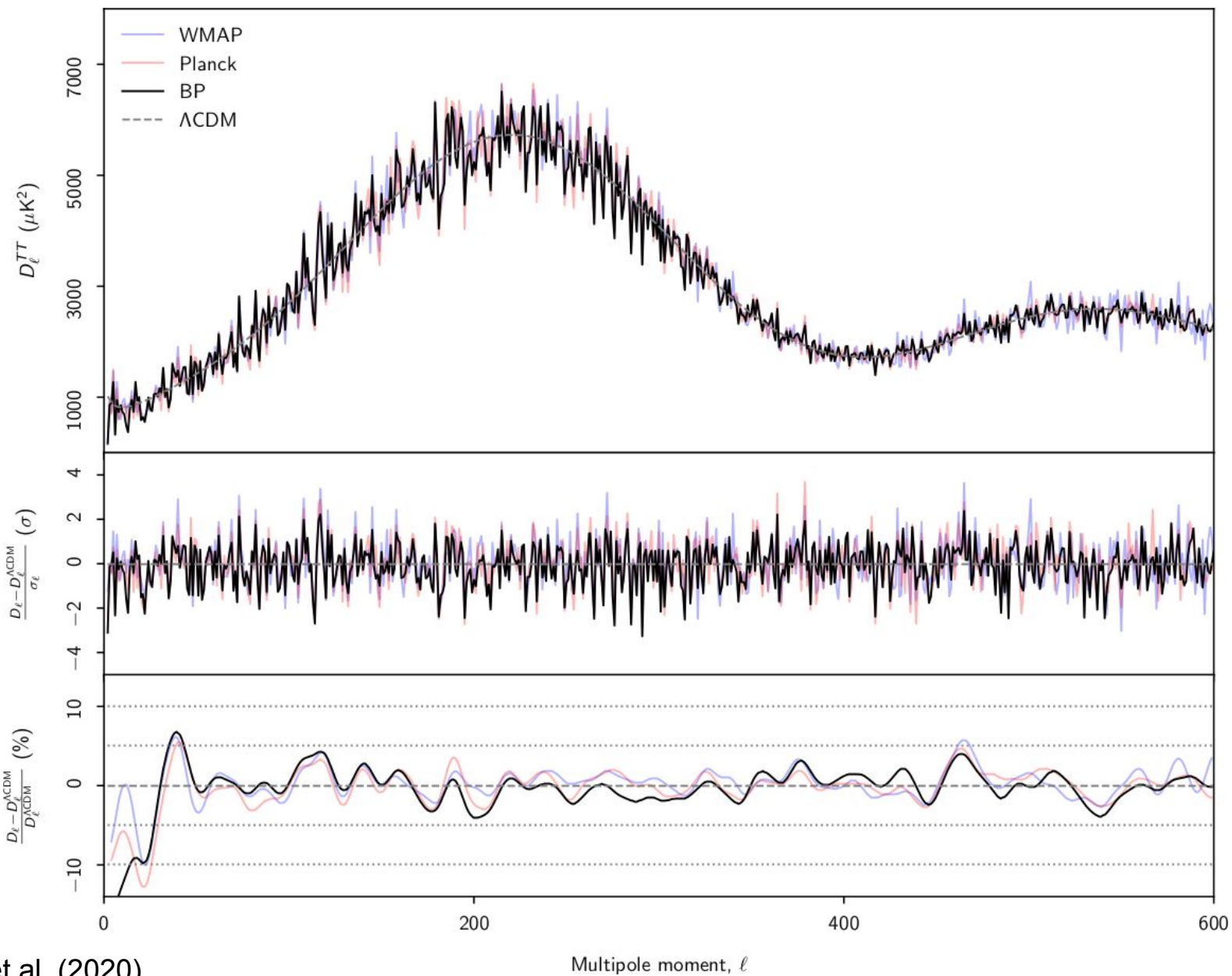
EXPERIMENT	AMPLITUDE		GALACTIC COORDINATES		REFERENCE
	[μK_{CMB}]		l	b	
			[deg]	[deg]	
<i>COBE</i> ^{a,b}	3358	± 23	264.31 ± 0.16	48.05 ± 0.09	Lineweaver et al. (1996)
<i>WMAP</i> ^c	3355	± 8	263.99 ± 0.14	48.26 ± 0.03	Hinshaw et al. (2009)
LFI 2015 ^b	3365.5	± 3.0	264.01 ± 0.05	48.26 ± 0.02	Planck Collaboration II (2016)
HFI 2015 ^d	3364.29	± 1.1	263.914 ± 0.013	48.265 ± 0.002	Planck Collaboration VIII (2016)
LFI 2018 ^b	3364.4	± 3.1	263.998 ± 0.051	48.265 ± 0.015	Planck Collaboration II (2020)
HFI 2018 ^d	3362.08	± 0.99	264.021 ± 0.011	48.253 ± 0.005	Planck Collaboration III (2020)
NPIPE ^{a,c}	3366.6	± 2.6	263.986 ± 0.035	48.247 ± 0.023	Planck Collaboration (2020)
BEYONDPLANCK ^e ..	3359.5	± 1.9	263.97 ± 0.09	48.30 ± 0.03	Section 9.5



CMB temperature sample



CMB: High- l TT spectrum



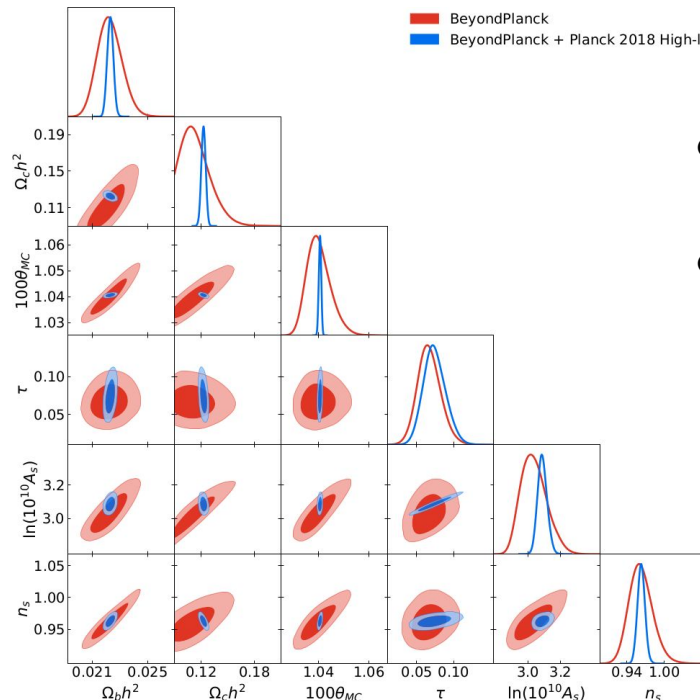
Colombo et al. (2020)

Cosmological parameters



Paradiso et al. (2020)

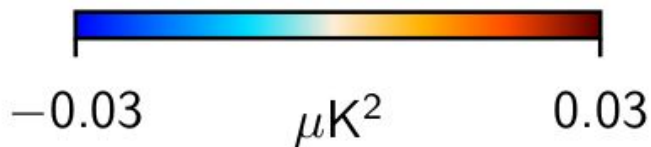
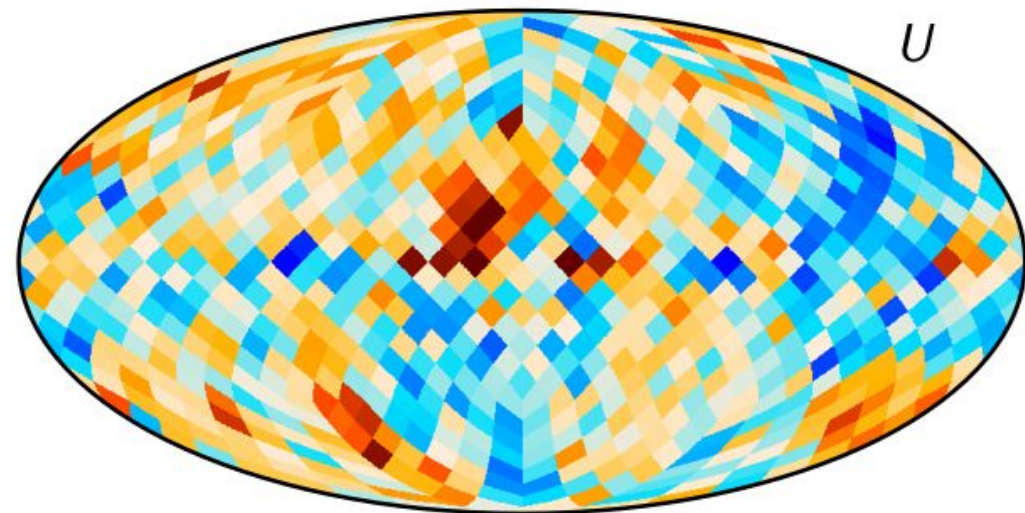
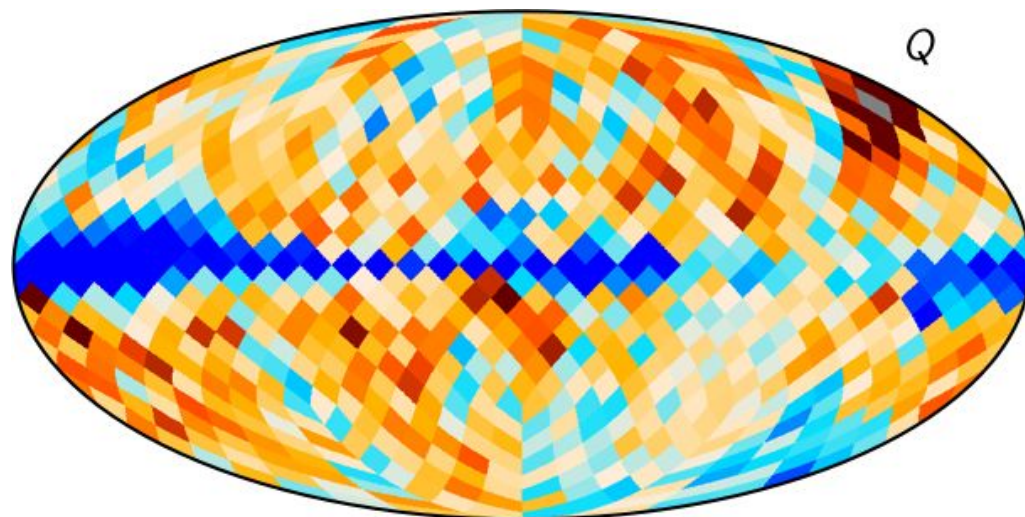
PARAMETER	BEYONDPLANCK		Planck 2018		WMAP	
	$\ell \leq 600$	+Planck $\ell > 600$	ESTIMATE	$\Delta(\sigma)$	ESTIMATE	$\Delta(\sigma)$
$\Omega_b h^2$	0.02226 ± 0.00088	0.02230 ± 0.00022	0.02237 ± 0.00015	-0.1	0.02243 ± 0.00050	-0.2
$\Omega_c h^2$	0.115 ± 0.016	0.1227 ± 0.0025	0.1200 ± 0.0012	-0.3	0.1147 ± 0.0051	0
Ω_Λ	0.721 ± 0.025	...
$100\theta_{MC}$	1.0402 ± 0.0048	1.04064 ± 0.00048	1.04092 ± 0.00031	-0.2
τ	0.067 ± 0.016	0.074 ± 0.015	0.054 ± 0.007	0.8	0.089 ± 0.0014	-1.4
$10^9 \Delta_{\mathcal{R}}^2$	2.41 ± 0.10	...
$\ln(10^{10} A_s)$	3.035 ± 0.079	3.087 ± 0.029	3.044 ± 0.014	-0.1
n_s	0.962 ± 0.019	0.9632 ± 0.0060	0.9649 ± 0.0042	-0.1	0.972 ± 0.013	-0.5



- Statistically consistent with previous estimates
- Larger error bars since we only use LFI and WMAP data
 - Formally speaking, we also marginalize over a much richer instrument and foreground model, but this is negligible in temperature compared to cosmic variance



Low-resolution CMB map and covariance matrix



Compute low-resolution CMB map and covariance matrix directly from samples:

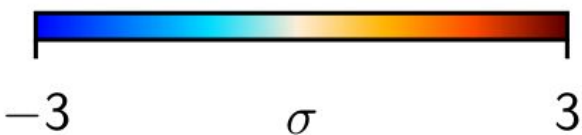
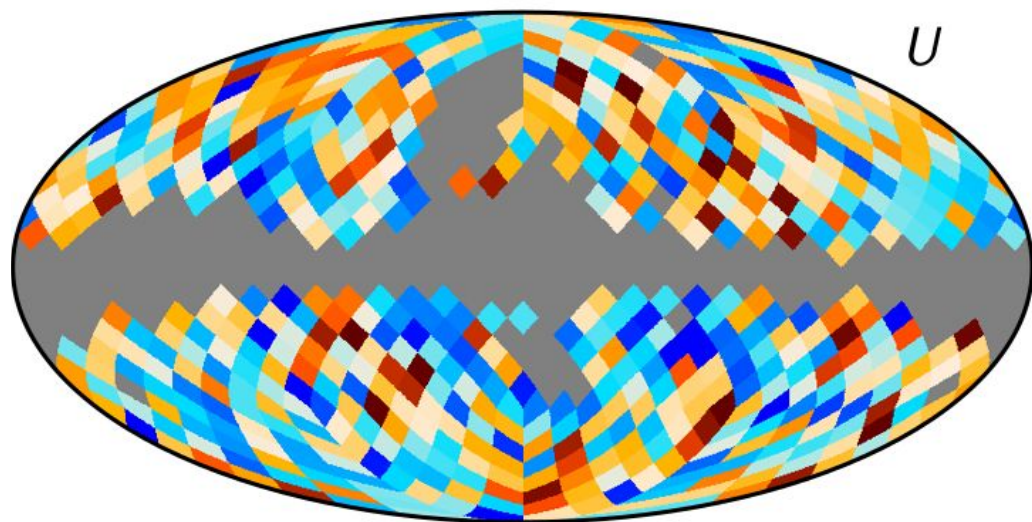
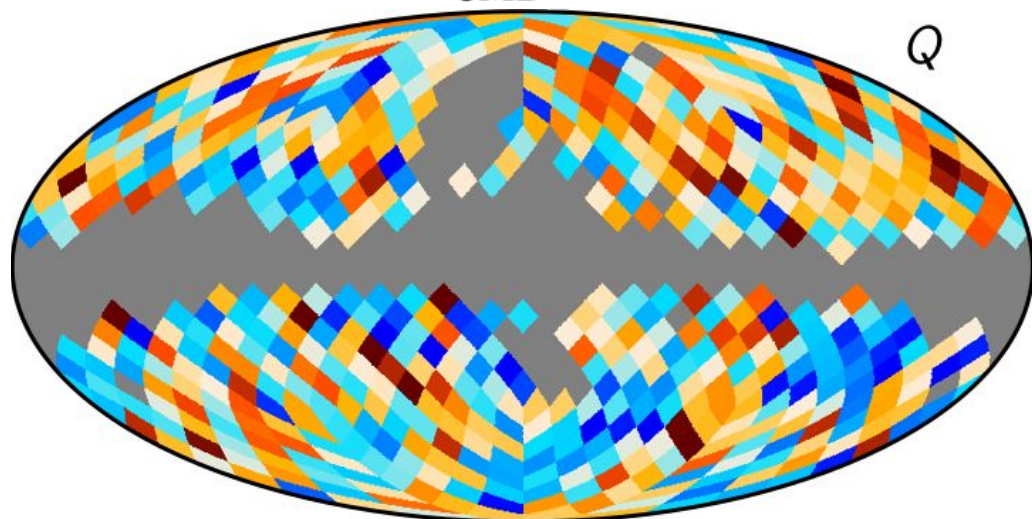
$$\hat{s}_{\text{CMB}} = \langle s_{\text{CMB}}^i \rangle$$

$$\mathbf{N} = \langle (s_{\text{CMB}}^i - \hat{s}_{\text{CMB}})(s_{\text{CMB}}^i - \hat{s}_{\text{CMB}})^t \rangle$$

*This is the first time uncertainties from **gain, bandpass and a fine-grained foreground model** have been consistently propagated into **CMB low- l likelihood inputs!***

Low-resolution CMB map and covariance matrix

$$N_{\text{CMB}}^{-1/2} s_{\text{CMB}}$$



Compute low-resolution CMB map and covariance matrix directly from samples:

$$\hat{s}_{\text{CMB}} = \langle s_{\text{CMB}}^i \rangle$$

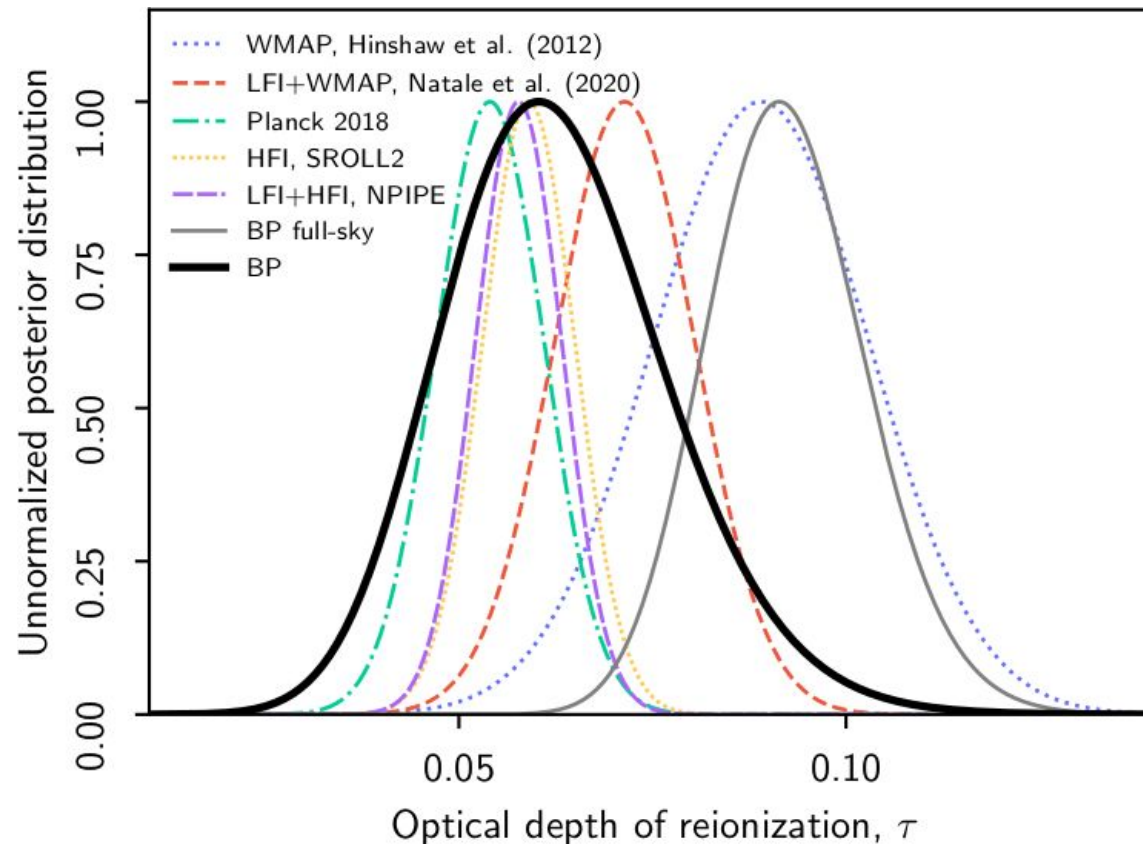
$$N = \langle (s_{\text{CMB}}^i - \hat{s}_{\text{CMB}})(s_{\text{CMB}}^i - \hat{s}_{\text{CMB}})^t \rangle$$

*This is the first time uncertainties from **gain, bandpass and a fine-grained foreground model** have been consistently propagated into **CMB low- l likelihood inputs!***

CMB: Low- l polarization likelihood, τ and r

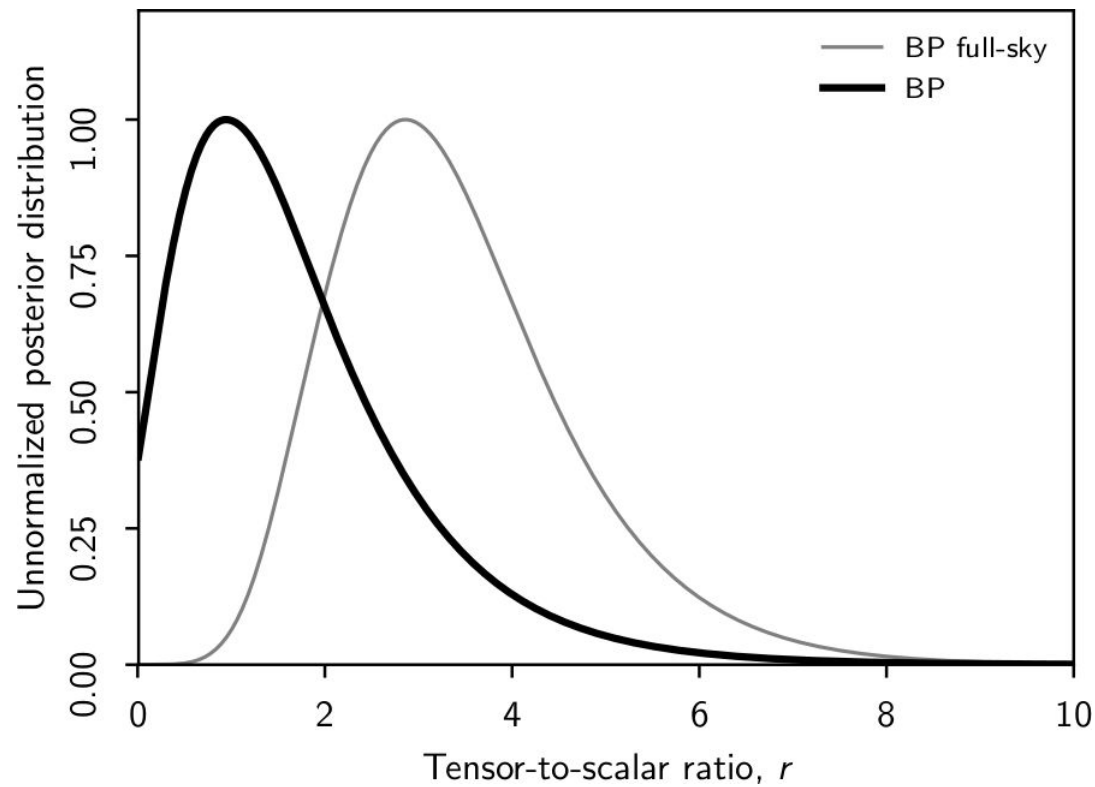


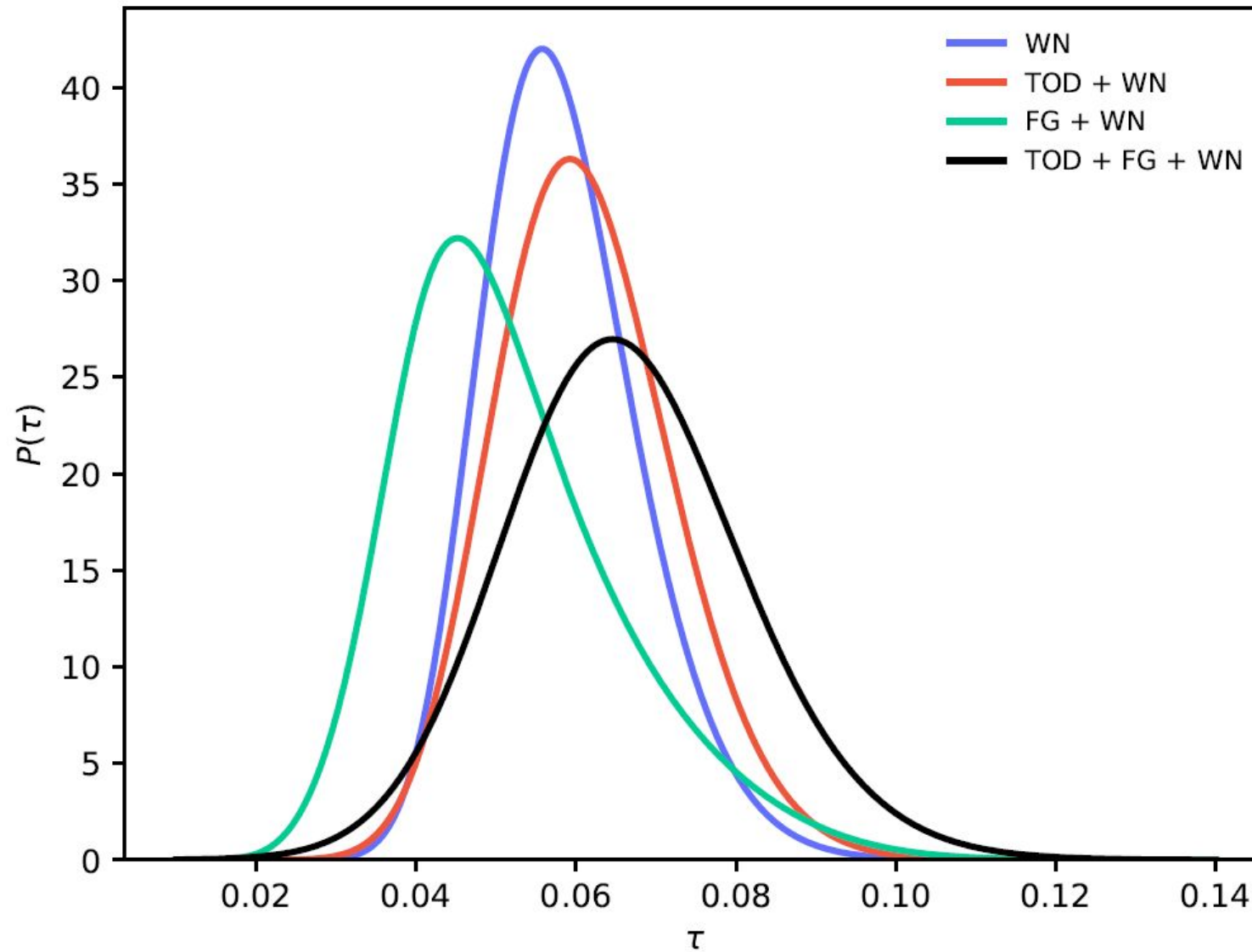
$$P(C_\ell | \hat{\mathbf{s}}_{\text{CMB}}) \propto \frac{e^{-\frac{1}{2} \hat{\mathbf{s}}_{\text{CMB}}^t (\mathbf{S}(C_\ell) + \mathbf{N})^{-1} \hat{\mathbf{s}}_{\text{CMB}}}}{\sqrt{|\mathbf{S}(C_\ell) + \mathbf{N}|}}$$



Paradiso et al. (2020)

$$P(C_\ell | \hat{\mathbf{s}}_{\text{CMB}}) \propto \frac{e^{-\frac{1}{2} \hat{\mathbf{s}}_{\text{CMB}}^t (\mathbf{S}(C_\ell) + \mathbf{N})^{-1} \hat{\mathbf{s}}_{\text{CMB}}}}{\sqrt{|\mathbf{S}(C_\ell) + \mathbf{N}|}}$$



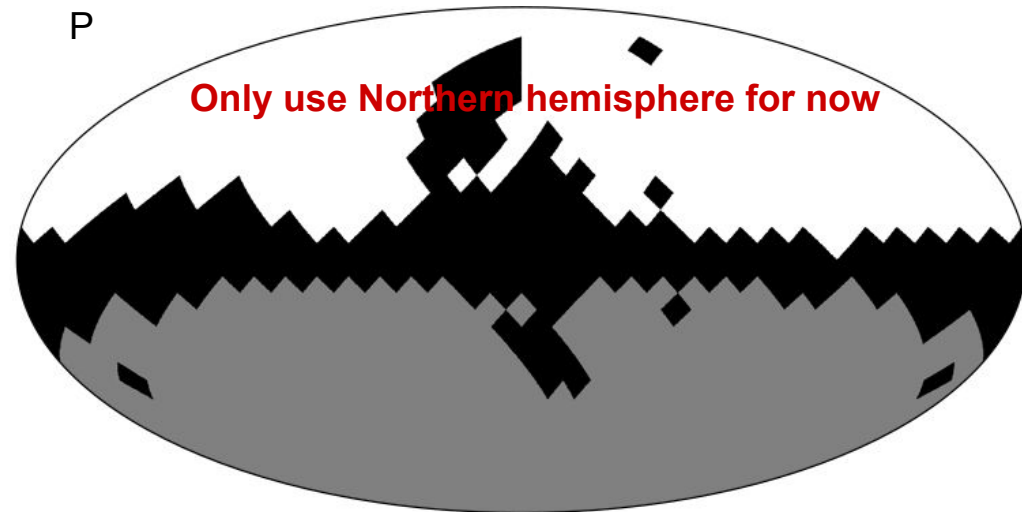
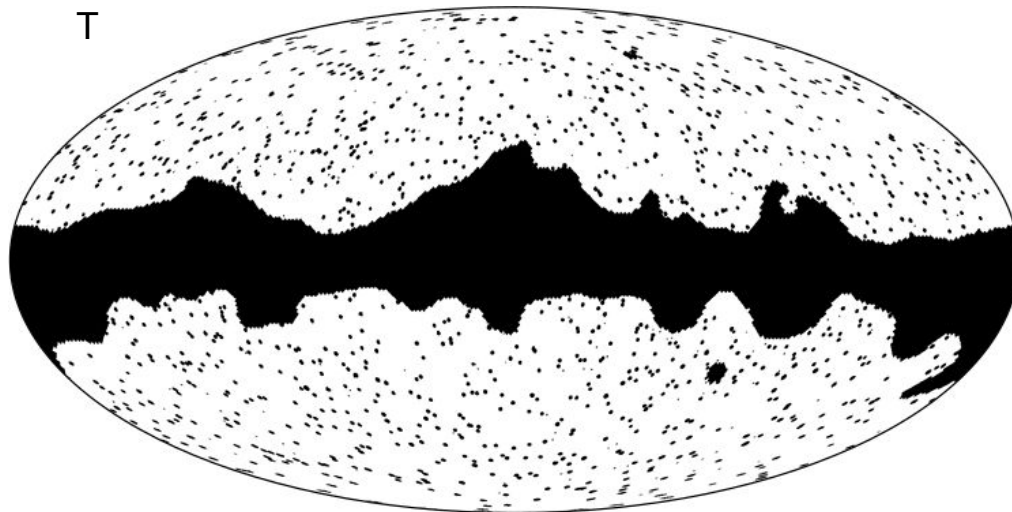


CMB: Goodness-of-fit and masking



ANALYSIS NAME	DATA SETS	$f_{\text{sky}}^{\text{pol}}$	τ	$r_{95\%}^{BB}$	χ^2 PTE	REFERENCE
BEYONDPLANCK, $\ell = 2-8$	LFI, WMAP $Ka-V$	0.36	$0.060^{+0.015}_{-0.013}$	< 4.3	0.16	Paradiso et al. (2020)
BEYONDPLANCK, $\ell = 3-8$	LFI, WMAP $Ka-V$	0.36	$0.061^{+0.015}_{-0.014}$	< 5.4	0.16	Paradiso et al. (2020)
BEYONDPLANCK, $\ell = 2-8$, full-sky . .	LFI, WMAP $Ka-V$	0.74	$0.091^{+0.010}_{-0.098}$	$2.9^{+1.3}_{-1.0}$	$5 \cdot 10^{-4}$	Paradiso et al. (2020)
WMAP 9-yr	WMAP $Ka-V$	0.76	0.089 ± 0.014			Hinshaw et al. (2013)
Natale et al.	LFI 70, WMAP $Ka-V$	0.54	0.071 ± 0.009			Natale et al. (2020)
Planck 2018	HFI 100×143	0.50	0.051 ± 0.009	< 0.41		Planck Collaboration V (2020)
SROLL2	HFI 100×143	0.50	0.059 ± 0.006			Pagano et al. (2020)
NPIPE (Commander CMB)	LFI+HFI	0.50	0.058 ± 0.006	< 0.16		Tristram et al. (2020)

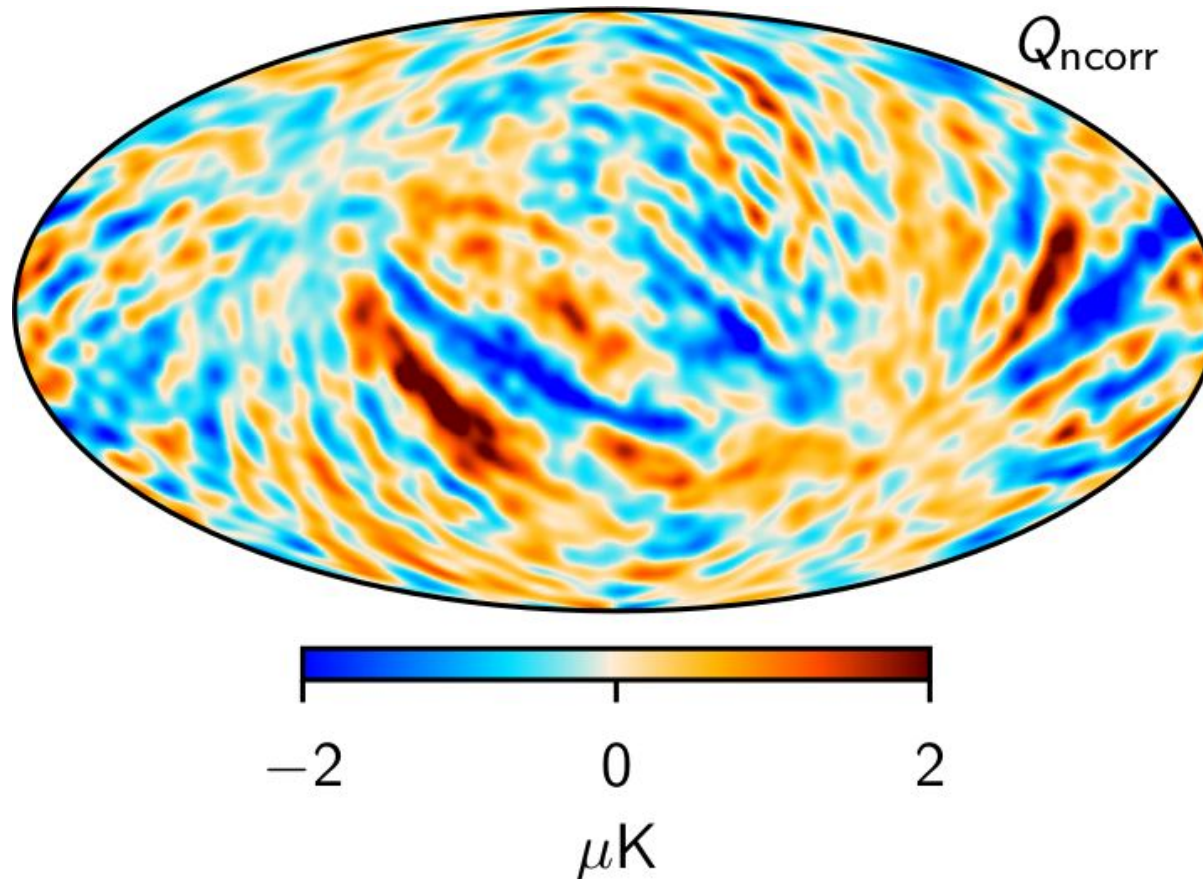
Paradiso et al. (2020)



Full-sky polarization mask has unacceptable χ^2 !



Outstanding issues 1: Stripes in 44 GHz



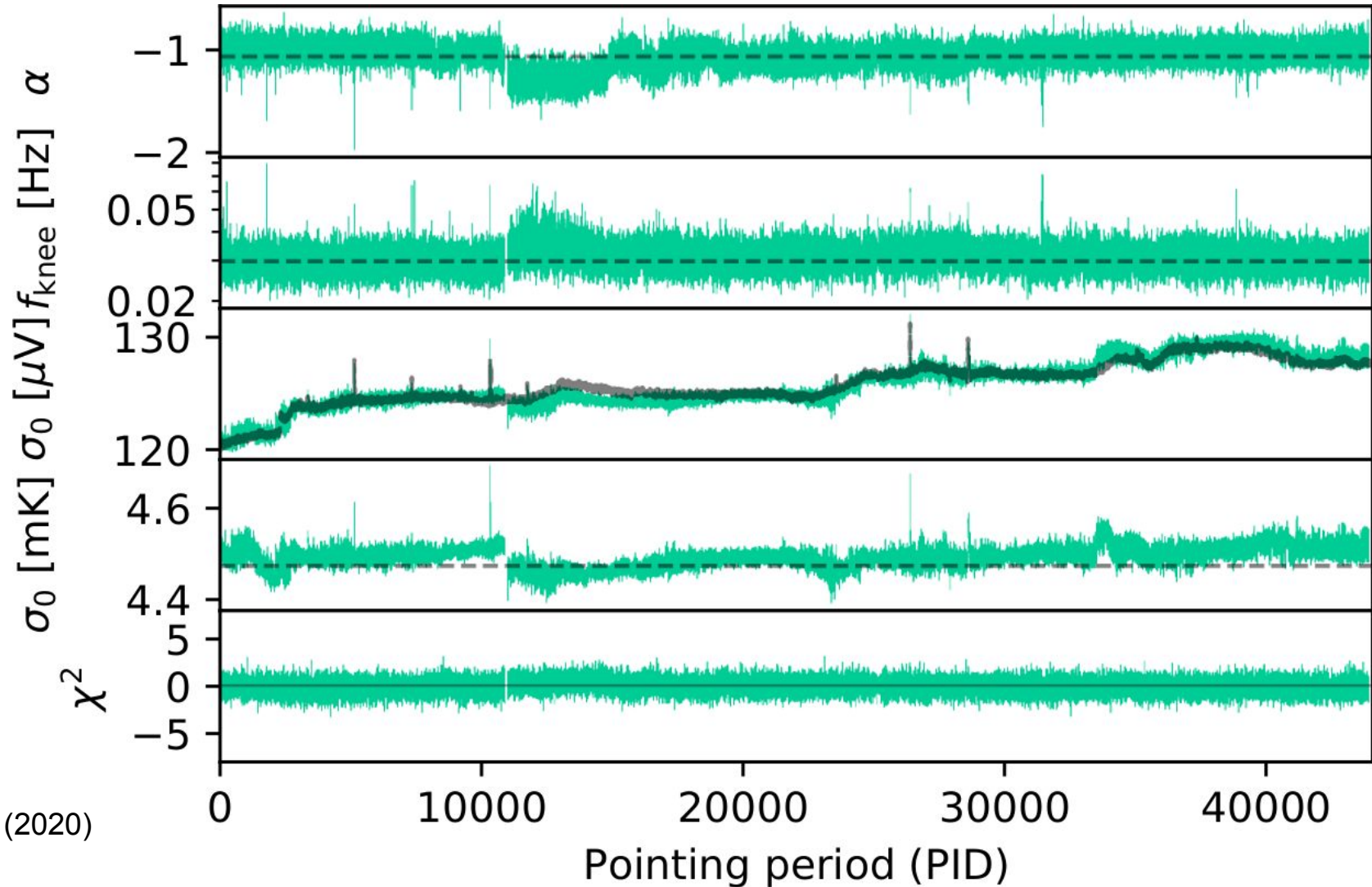
Ihle et al. (2020)

- Correlated noise map at 44 GHz shows strong stripes in Southern hemisphere
- Origin not yet understood, but being actively investigated
- Seems associated with poor gain model for some Planck scanning rings
 - Sub-optimal processing mask?
 - Undetected gain jumps?

1/f model at 70 GHz fits well



Correlated noise parameters for 70GHz 23M radiometer



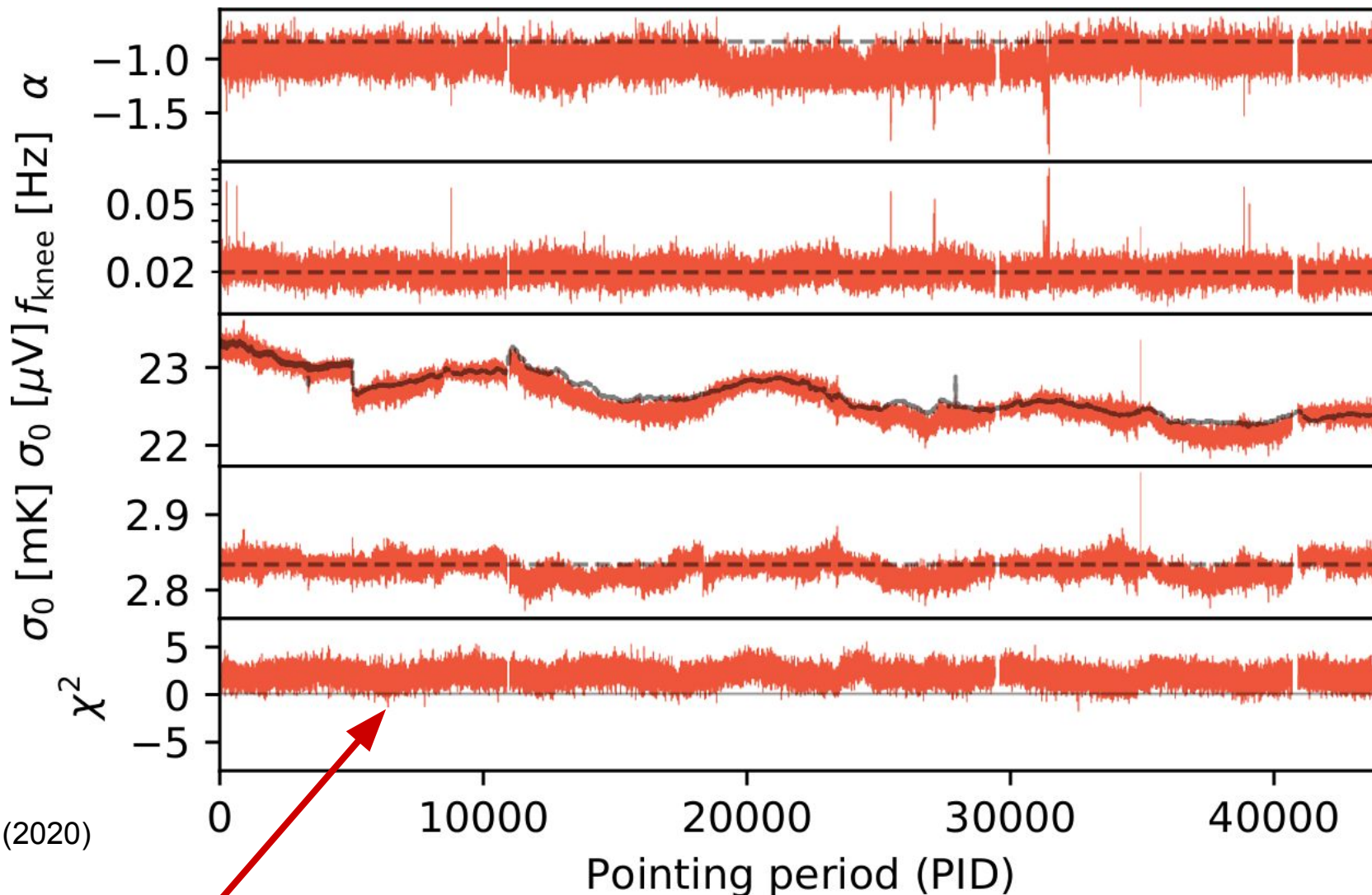
Ihle et al. (2020)

Outstanding issues 2: $1/f$ model at 30 and 44 GHz



European Commission

Correlated noise parameters for 44GHz 25M radiometer



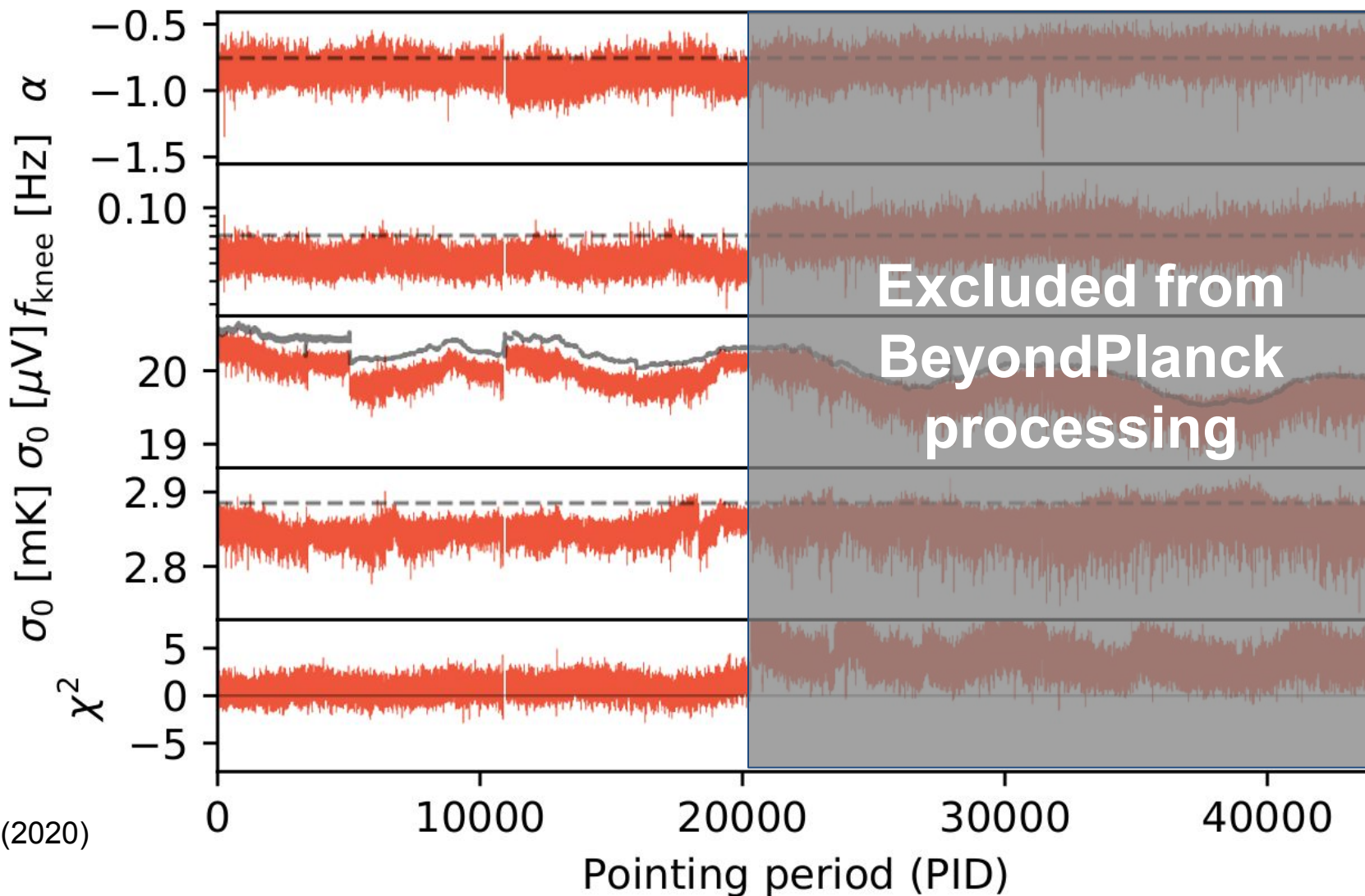
Ihle et al. (2020)

χ^2 excess of 2-3 sigma per PID!

Outstanding issues 2: $1/f$ model at 30 and 44 GHz



Correlated noise parameters for 44GHz 26S radiometer

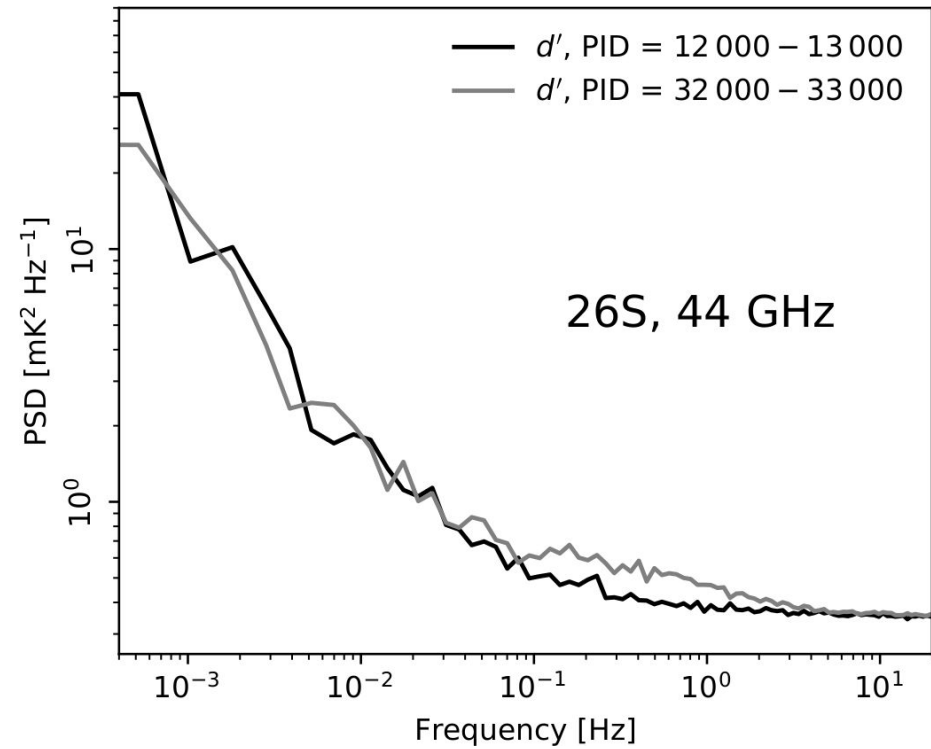
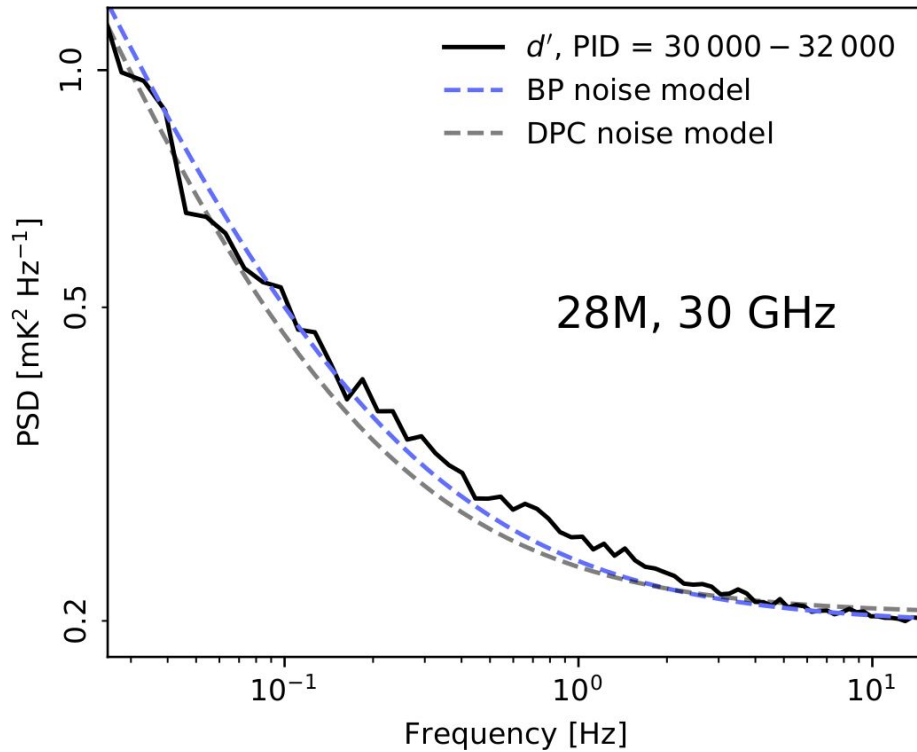


Ihle et al. (2020)

Outstanding issues 2: 1/f model at 30 and 44 GHz

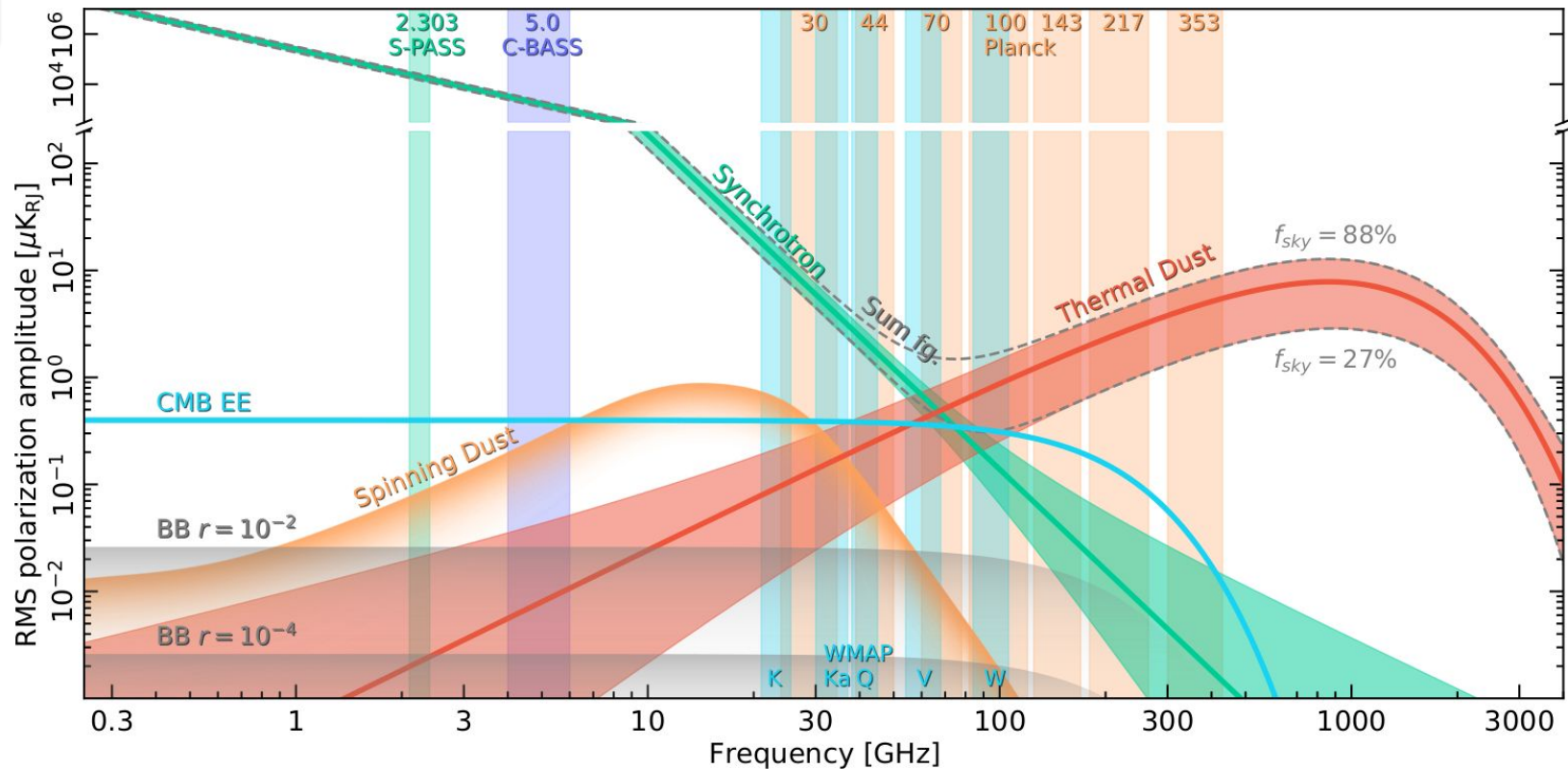


Ihle et al. (2020)



- Correlated noise is fitted using a standard 1/f model: $P(f) = \sigma_0^2 \left[1 + \left(\frac{f}{f_{\text{knee}}} \right)^\alpha \right]$
- Not a statistically sufficient model for 30 and 44 GHz channels
- Significant and time-variable **excess between 0.1 and 5 Hz**, corresponding to angular scales between **1 and 60 degrees on the sky**
 - Appears non-thermal in origin. Electrical issue? Investigation on-going

The future: Cosmoglobe



- BeyondPlanck has successfully implemented an efficient end-to-end analysis framework for global CMB analysis
 - So far, only LFI has been fully integrated
- Now it needs to be populated with complementary datasets:
 - Public: Planck HFI, WMAP, FIRAS, DIRBE...
 - Proprietary: BICEPx, C-BASS, CLASS, COMAP, PASIPHAE, QUIJOTE, QUIET, S-PASS, SPIDER...?
- Obviously a community effort, and will rely on active participation from interested experiments
- This effort will be organized by the **Cosmoglobe** project; see talk by Ingunn Wehus on Friday

- The main BeyondPlanck computer code is called `Commander3`
 - Direct generalization of `Commander2`, as used in the Planck 2018 analysis

- Commander3 is publicly released under a GPL3 license:

<https://github.com/Cosmoglobe/Commander>

- BeyondPlanck products, software and documentation are available through the project home page:

<https://beyondplanck.science>

- Caveats:
 - All software is provided as is, with no guarantees of any kind
 - This is a software platform for cutting-edge research, and therefore by nature a continuous work-in-progress
 - Support is provided on a strictly voluntary basis; there is no “help desk”
 - If you want hands-on assistance, proposing a joint research project with one or more experienced BeyondPlanck/Cosmoglobe team members is a good idea

The BeyondPlanck release conference



European
Commission

Schedule

Full program of the BeyondPlanck Release Conference. All times are expressed in CET (UTC+1).

Day 1 - Wednesday, 18 Nov		Day 2 - Thursday, 19 Nov	
14:45	Zoom connection opens	14:45	Zoom connection opens
15:00	BeyondPlanck: Motivation and main results (Hans Kristian Eriksen)	15:00	Noise estimation (Håvard Tveit Ihle)
16:00	Planck overview (Charles Lawrence)	15:30	Calibration (Eirik Gjerløw)
	Break	15:55	Sidelobes (Mathew Galloway)
17:30	The Planck LFI instrument (Marco Bersanelli)	16:10	Leakage corrections (Trygve Leithe Svalheim)
18:05	Planck LFI DPC processing and final status (Samuele Galeotta)		Break
18:40	NPIPE (Reijo Keskitalo)	17:00	Commander and computational aspects (Mathew Galloway)
19:10	CMB mapmaking by Gibbs sampling (Elina Keihänen)	17:20	BeyondPlanck frequency maps (Anna-Stiina Suur-Uski)
		17:55	CMB constraints (Loris Colombo)
		18:25	Cosmological parameters (Simone Paradiso)
Day 3 - Friday, 20 Nov			
14:45	Zoom connection opens		
15:00	LiteBIRD (Masashi Hazumi, TBC)		
15:30	Intensity foregrounds (Kristian Joten Andersen)		
16:00	Polarization foregrounds (Trygve Leithe Svalheim)		
16:25	Limits on polarized AME (Daniel Herman)		
	Break		
17:15	Preliminary analysis of external data sets (Duncan Watts)		
17:30	BeyondPlanck and LiteBIRD (Ragnhild Aurlien)		
18:00	Cosmoglobe (Ingunn Kathrine Wehus)		
18:20	Planck products and PLA (Marcos López-Caniego)		
18:40	Summary and questions (Bruce Partridge)		

Schedule

Time	Monday, 23 Nov	Tuesday, 24 Nov
15:00	Introduction	c3pp and postprocessing
15:30	The BP data model	BeyondPlanck products
16:00	Jupyter I	Jupyter III
16:30	Jupyter II	Jupyter IV
17:00	Break	Break
17:30	Installing Commander	Adding external sky maps
18:00	The Commander parameter file	Adding external time-ordered data
18:30	Running Commander	Cosmoglobe
19:00	Q&A	Q&A



BeyondPlanck project

- Main webpage: <https://beyondplanck.science>
Products: <https://products.beyondplanck.science>
<https://pla.esac.esa.int> (subset; when papers are accepted)
Papers: <https://beyondplanck.science/products/publications>
Discussion forum: <https://forums.beyondplanck.science>

Commander

- Source code : <https://github.com/cosmoglobe/Commander>
Documentation: <https://docs.beyondplanck.science>

Cosmoglobe

- Main webpage: <http://cosmoglobe.uio.no>

Planck Legacy Archive (selected BeyondPlanck products coming soon)

- Link: <https://pla.esac.esa.int>

- **BeyondPlanck has successfully implemented a framework for global end-to-end Bayesian CMB analysis, and demonstrated this using Planck LFI**
- Important advantages of this framework include:
 - Joint instrument and foreground modelling ⇒ more robust results
 - End-to-end error propagation ⇒ reliable uncertainties
 - Physically motivated models ⇒ intuitive interpretation
 - Multi-experiment analysis ⇒ naturally breaking degeneracies
 - Multi-level goodness-of-fit tests ⇒ detailed systematics monitoring
 - No intermediate human interaction ⇒ less room for mistakes
 - High computational efficiency ⇒ can run on inexpensive computers
- Next steps are to generalize and populate this framework with many more datasets, both public and proprietary
 - All interested parties are invited to join Cosmoglobe, working together toward a global model of the Universe in an Open Science-based community!

The BeyondPlanck collaboration



EU-funded institutions



Kristian Joten Andersen
Ragnhild Aurlien
Ranjay Banerji
Maksym Brilenkov
Hans Kristian Eriksen
Johannes Røsok Eskilt
Marie Kristine Foss
Unni Fuskeland
Eirik Gjerløw
Mathew Galloway
Daniel Herman
Ata Karakci
Håvard Tveit Ihle
Metin San
Trygve Leithe Svalheim
Harald Thommesen
Duncan Watts
Ingunn Kathrine Wehus



Sara Bertocco
Samuele Galeotta
Gianmarco Maggio
Michele Maris
Daniele Tavagnacco
Andrea Zacchei



Elina Keihänen
Anna-Stiina Suur-Uski



Stelios Bollanos
Stratos Gerakakis
Maria Ieronymaki
Ilias Ioannou



Marco Bersanelli
Loris Colombo
Cristian Franceschet
Davide Maino
Aniello Mennella
Simone Paradiso

External collaborators



Brandon Hensley



Jeff Jewell



Reijo Keskitalo



Bruce Partridge



Martin Reinecke

The BeyondPlanck collaboration



EU-funded institutions

10 postdocs
10 PhD students



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Ragnhild Aurlien
Ranajoy Banerji
Maksym Brilenkov
Hans Kristian Eriksen
Johannes Røsoek Eskilt
Marie Kristine Foss

Unni Fuskeland
Eirik Gjerløw
Mathew Galloway
Daniel Herman
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Stratos Gerakakis
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Ilias Ioannou

External collaborators



Brandon Hensley



Jeff Jewell



Reijo Keskitalo



Bruce Partridge



Martin Reinecke



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- “*BeyondPlanck*”
 - COMPET-4 program
 - PI: Hans Kristian Eriksen
 - Grant no.: 776282
 - Period: Mar 2018 to Nov 2020

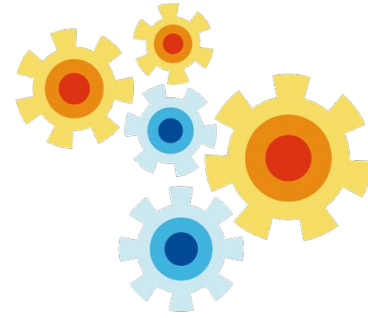
Collaborating projects:

- “*bits2cosmology*”
 - ERC Consolidator Grant
 - PI: Hans Kristian Eriksen
 - Grant no: 772 253
 - Period: April 2018 to March 2023
- “*Cosmoglobe*”
 - ERC Consolidator Grant
 - PI: Ingunn Wehus
 - Grant no: 819 478
 - Period: June 2019 to May 2024

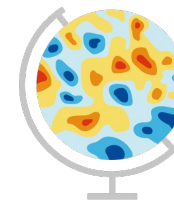


Questions?

Beyond PLANCK



Commander



Cosmoglobe Beyond PLANCK