

# Next steps in component separation for new CMB observables

Mathieu Remazeilles

Instituto de Fisica de Cantabria

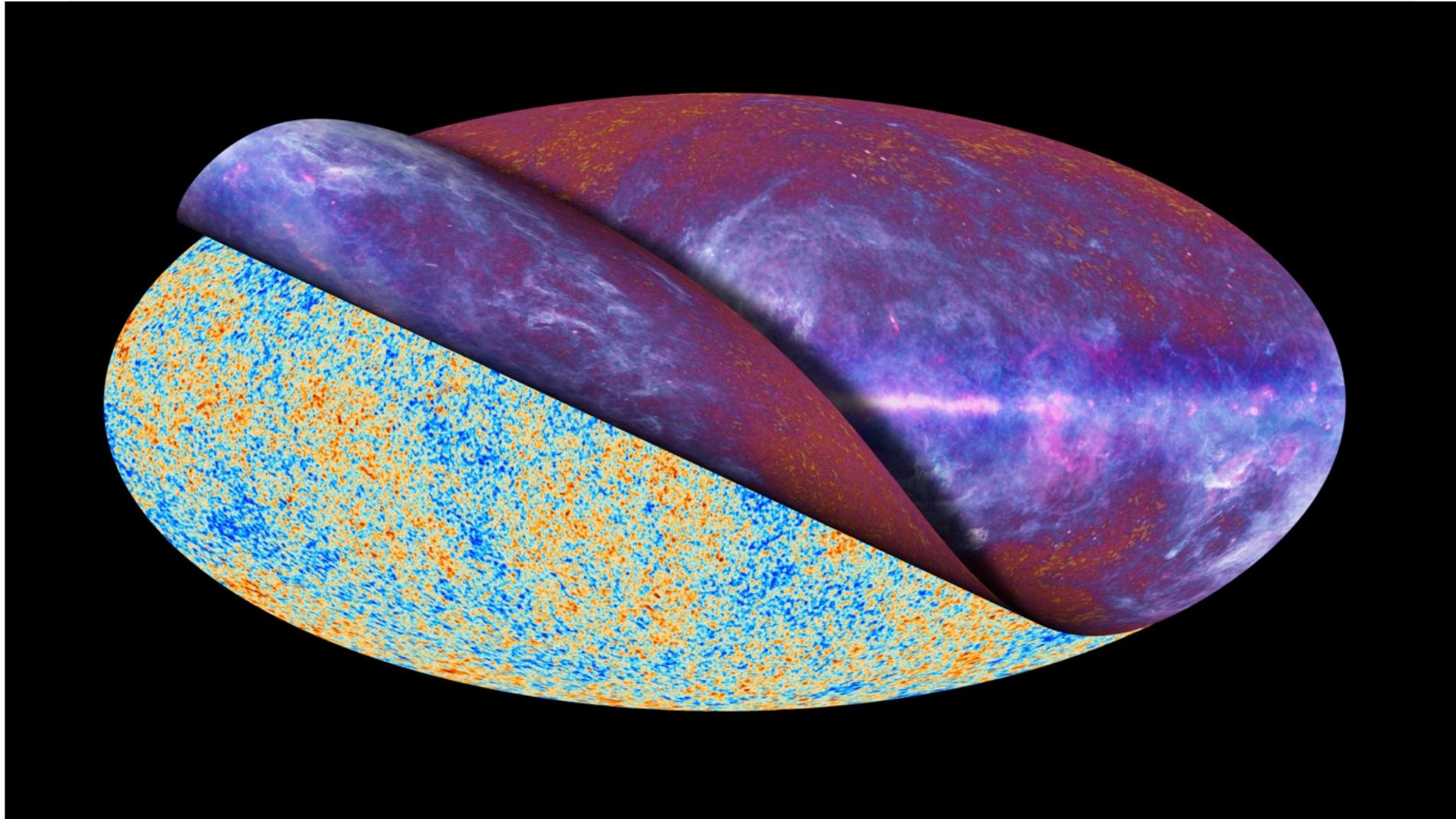
# Foregrounds obscure the cosmological signals



Credit: ESA

M. Remazeilles

# Planck: playground for component separation



Credit: ESA

M. Remazeilles

# Component separation methods

## Parametric

*Model-dependent*

**COMMANDER**

*Eriksen et al 2004, 2008*

## Blind

*No assumption  
about foregrounds*

**NILC**

*Delabrouille et al  
2009*

**SEVEM**

*Fernandez-Cobos  
et al 2012*

**SMICA**

*Cardoso et al 2008  
Delabrouille et al 2003*

**GNILC**

*Remazeilles et al  
2011b*

**CILC**

*Remazeilles et al  
2011a*

**MILCA**

*Hurier et al 2013*

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*de la Hoz et al 2020*

moments fitting

(not map-based)

*Vacher et al 2021, 2022*

*Mangili et al 2021*

*Azzoni et al 2020*

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Delta-map

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Foreground moment  
deprojection

cMILC

*Remazeilles et al 2021*

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## Machine learning

moment networks

*Jeffrey et al 2022*

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Foreground moment  
deprojection

cMILC

*Remazeilles et al 2021*

FastMEM

*Hobson et al  
1998*

CCA

*Bonaldi et al  
2006*

## Machine learning

moment networks

*Jeffrey et al 2022*

M. Remazeilles

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Delta-map

*Ichiki et al 2019*

# Component separation methods

## Parametric

Model-dependent

COMMANDER

*Eriksen et al 2004, 2008*

fqbuster

Talk by Giuseppe Puglisi  
*Hobson et al 2009, 2016*  
*Hazumi et al 2011, 2019*

B-SeCRET

Talk by Elena de la Hoz  
*Hazumi et al 2020*

moments fitting  
(not map-based)

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Foreground moment deprojection

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FastMEM

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Talk by Benjamin Beringue

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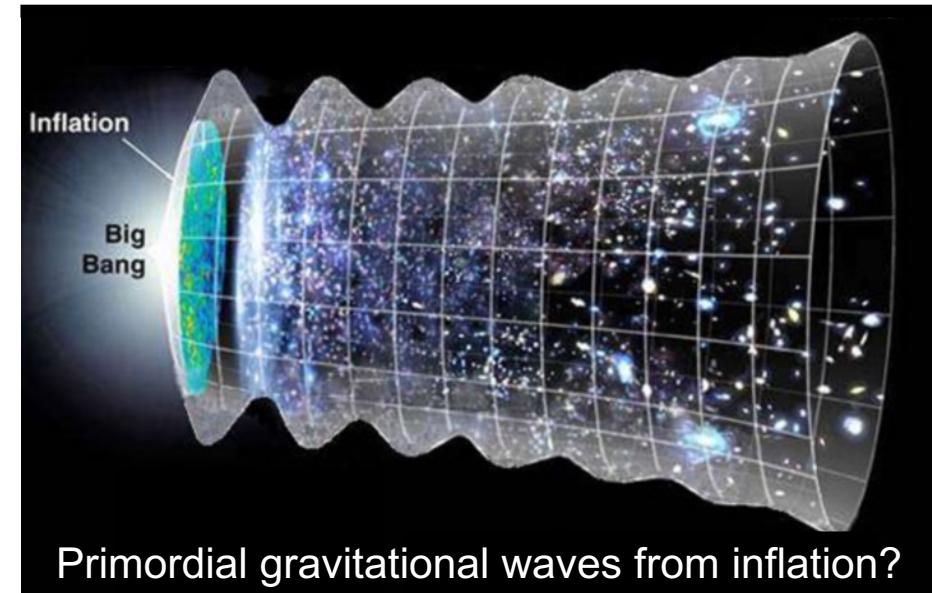
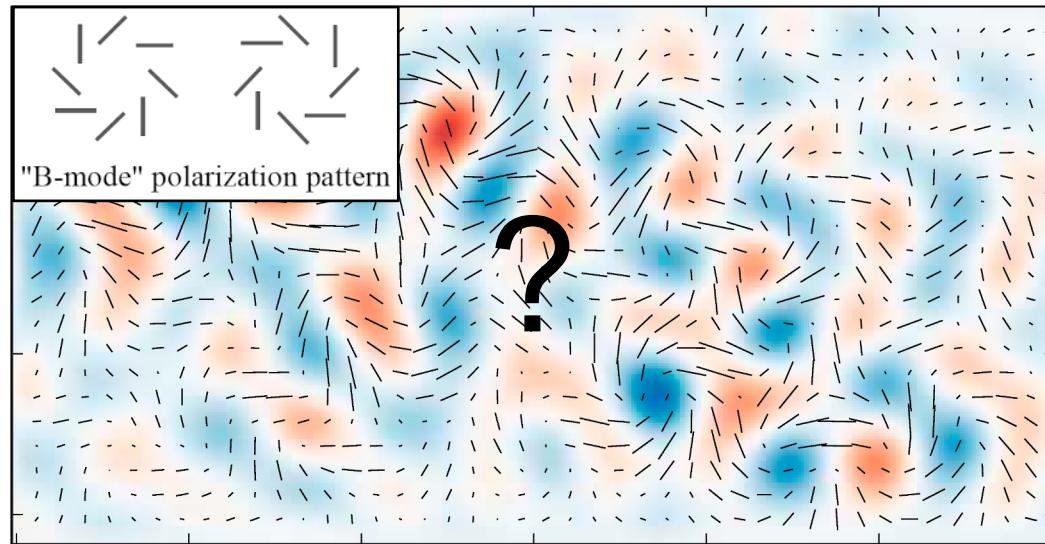
Delta-map

*Ichikawa et al 2019*

# New CMB observables

## CMB B-mode polarization

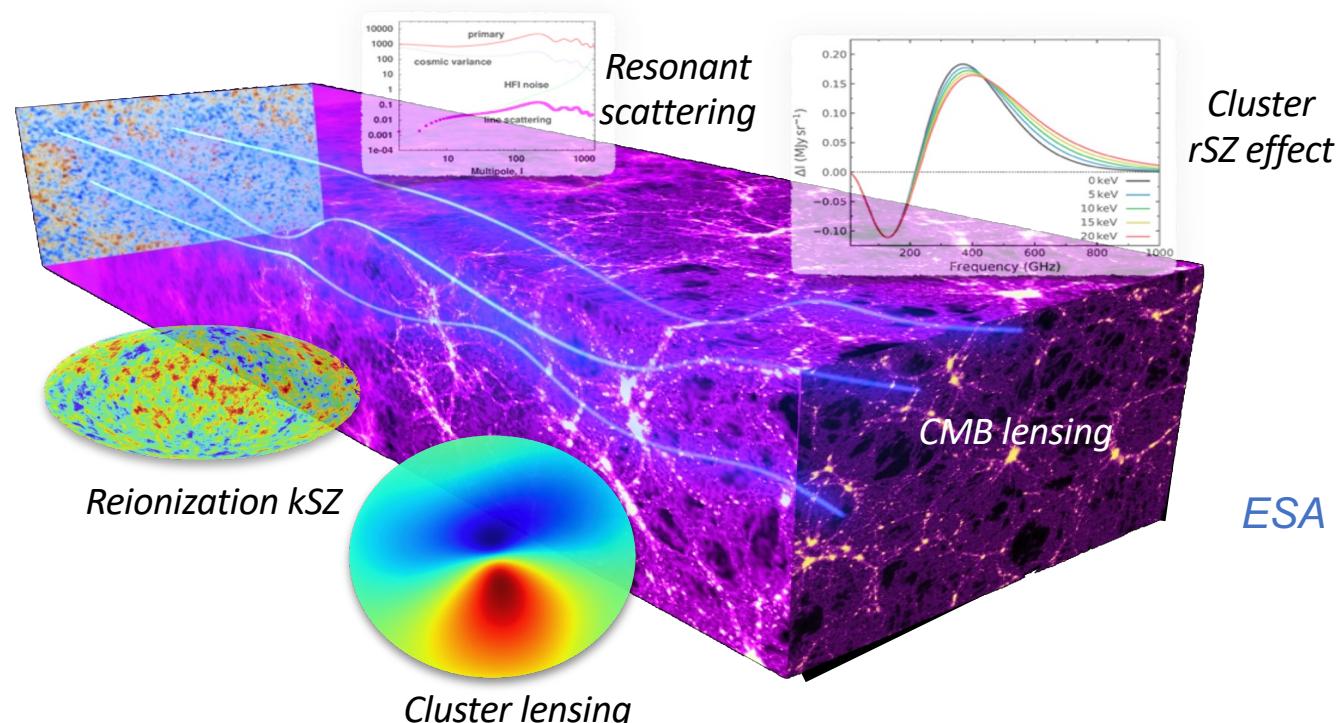
- *Footprint of primordial gravitational waves of quantum origin*
- *Direct evidence for inflation epoch in the early Universe*
- *Informs on the energy scale of inflation (tensor-to-scalar ratio)*



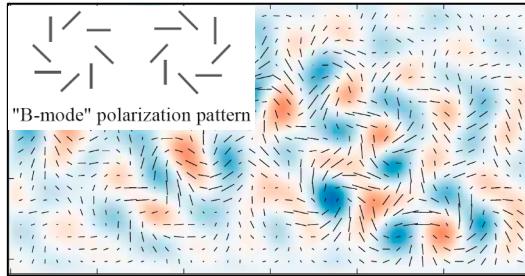
# New CMB observables

## CMB backlight (secondary spectral / spatial distortions)

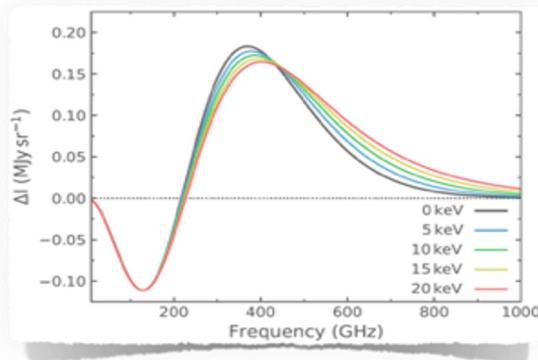
- *Spectral and spatial distortions imprinted on the CMB by the cosmic web*
- *Scattering by free electrons, atoms and molecules / gravitational lensing by LSS*
- *CMB “backlight” to probe baryonic and dark matter across the Universe*



# New era of faint signal regimes



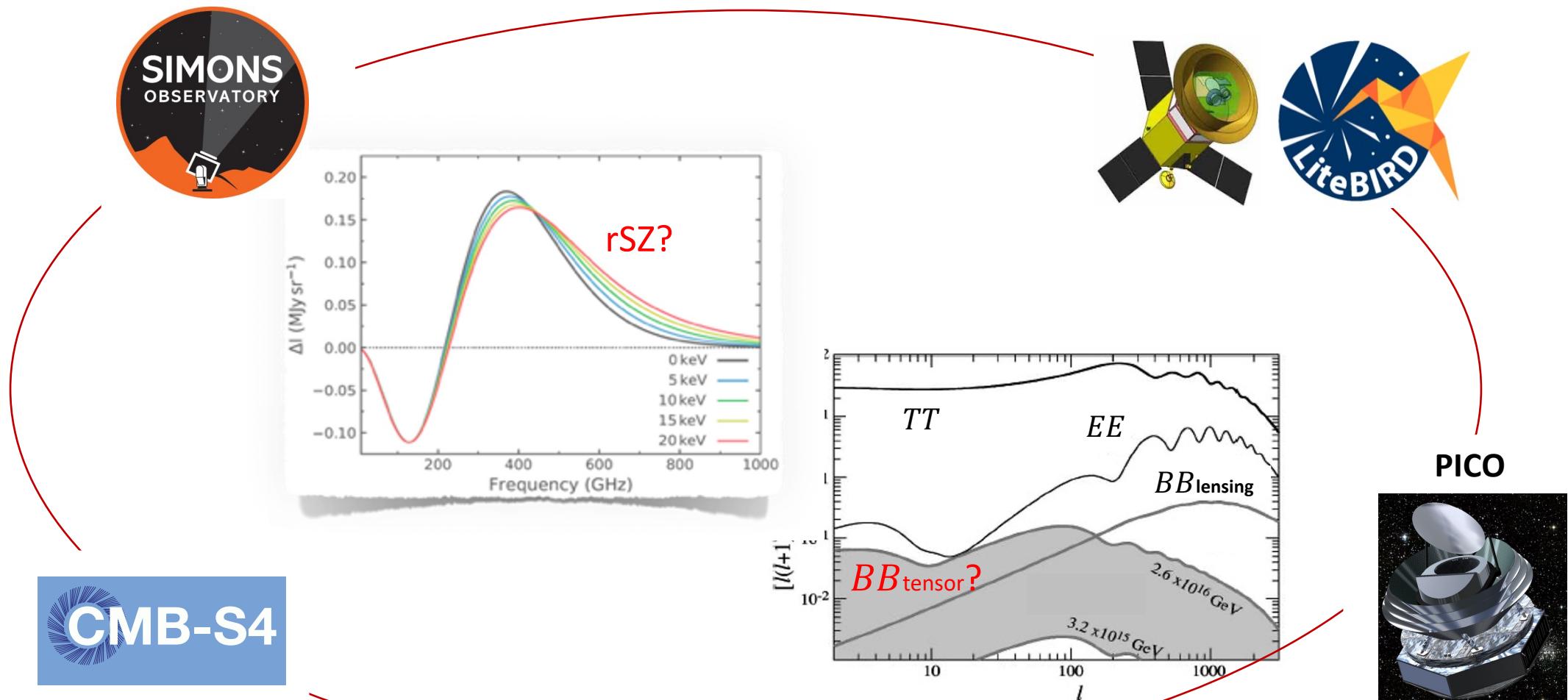
*Primary B-mode  $\sim 10 \text{ nK} \ll \text{Galactic foreground B-mode} \sim 10^7 \text{ nK}$*



*Relativistic SZ  $\sim \frac{1}{10}$  thermal SZ  $\ll \text{(Extra)galactic foregrounds}$*

Tiny modelling errors on foregrounds = Large error / bias on the signal!

# New sensitivities



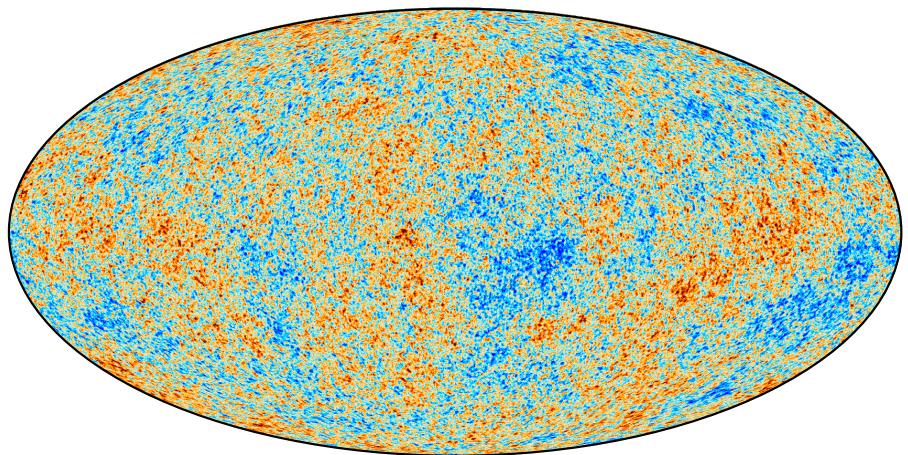
Much more sensitive to imperfect foreground modelling!  
Can we mitigate biases due to foregrounds down to below the noise?

# New layers of foregrounds complexity

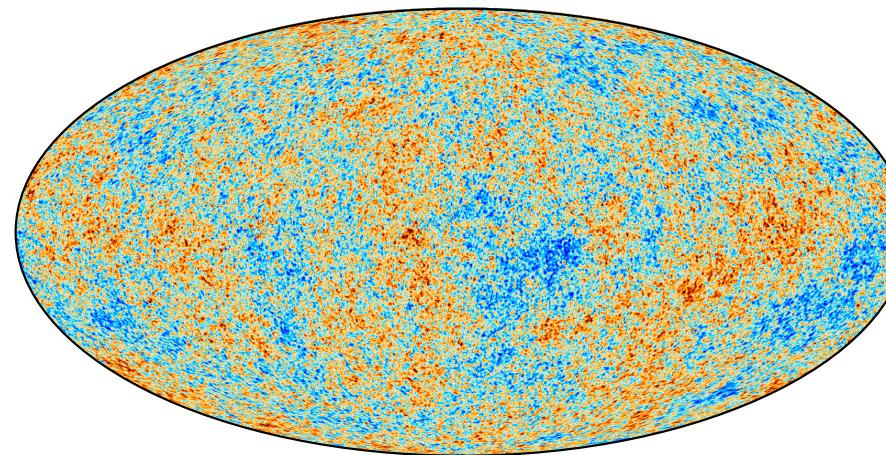
- **Lack of spectral information:** foregrounds are poorly known at the sensitivity levels required for these signals
- **Spectral distortions of the foregrounds**
- **Spectral degeneracies**
- **Foregrounds correlated with the signal of interest**

# Four consistent Planck CMB maps

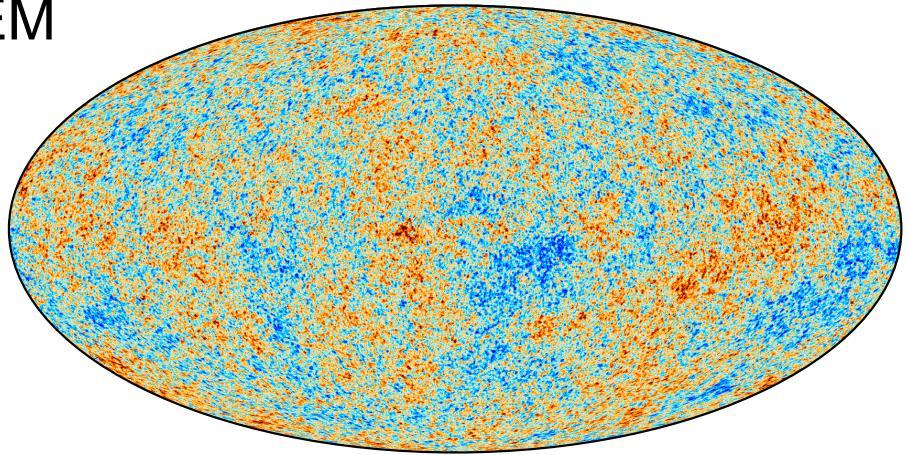
NILC



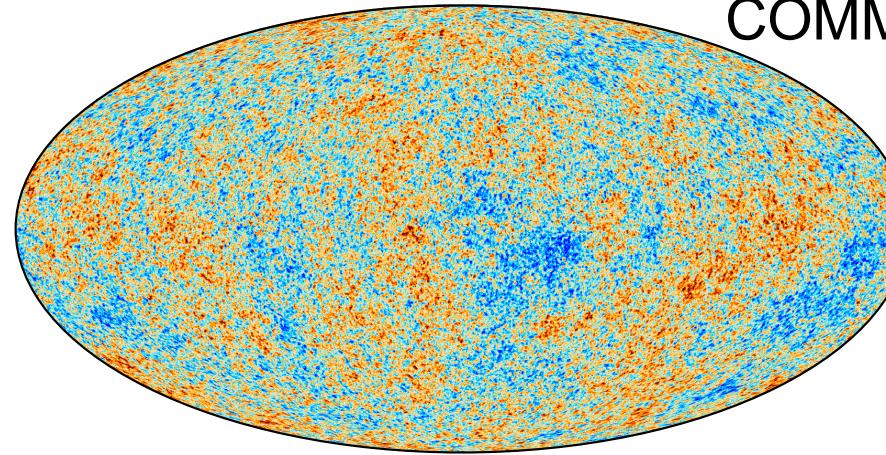
SMICA



SEVEM



COMMANDER



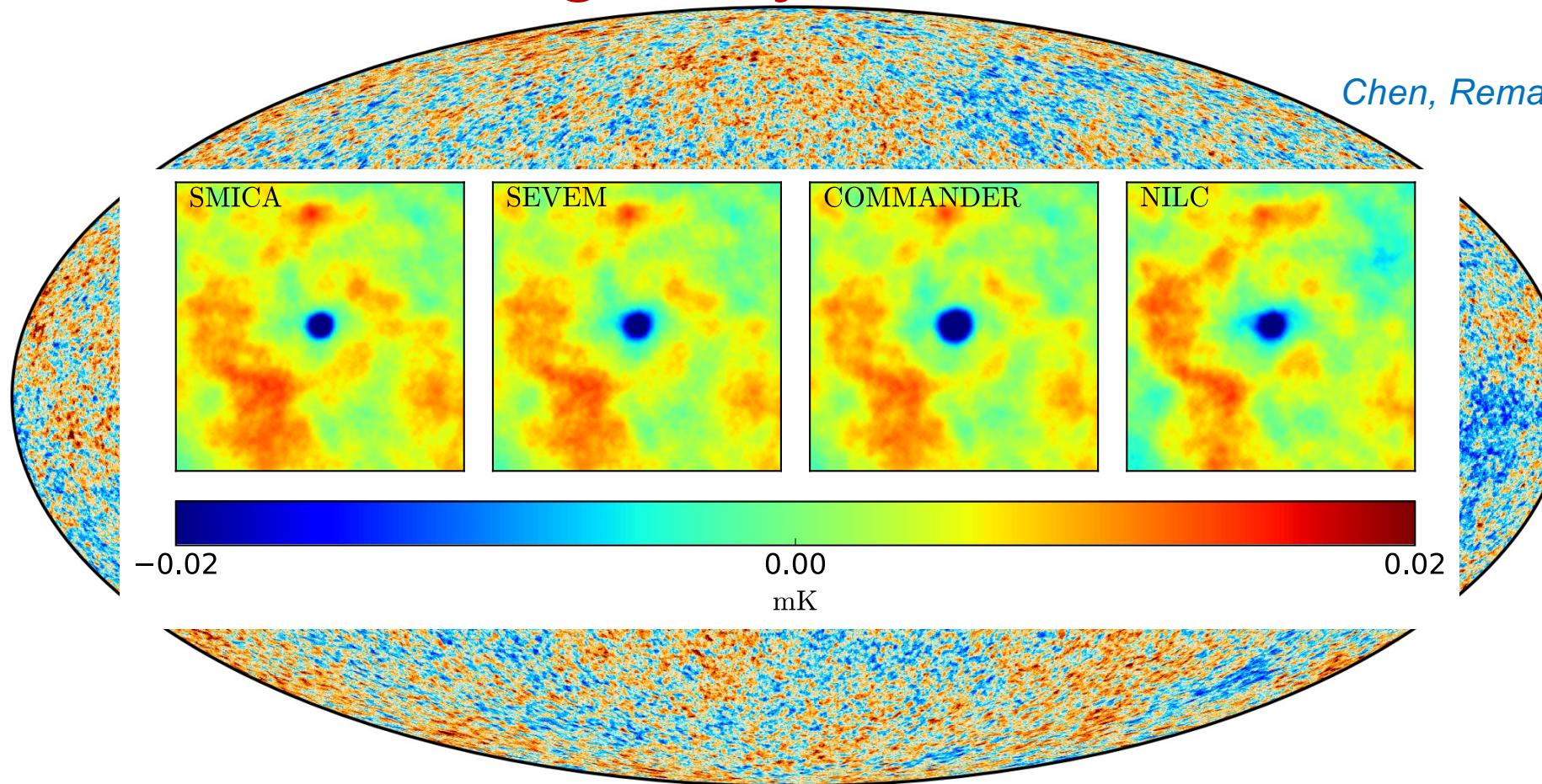
-300                     $\mu\text{K}$                     300

-300                     $\mu\text{K}$                     300

-300                     $\mu\text{K}$                     300

-300                     $\mu\text{K}$                     300

# Four consistent Planck CMB maps stacked on galaxy cluster locations



Chen, Remazeilles, Dickinson  
MNRAS 2018

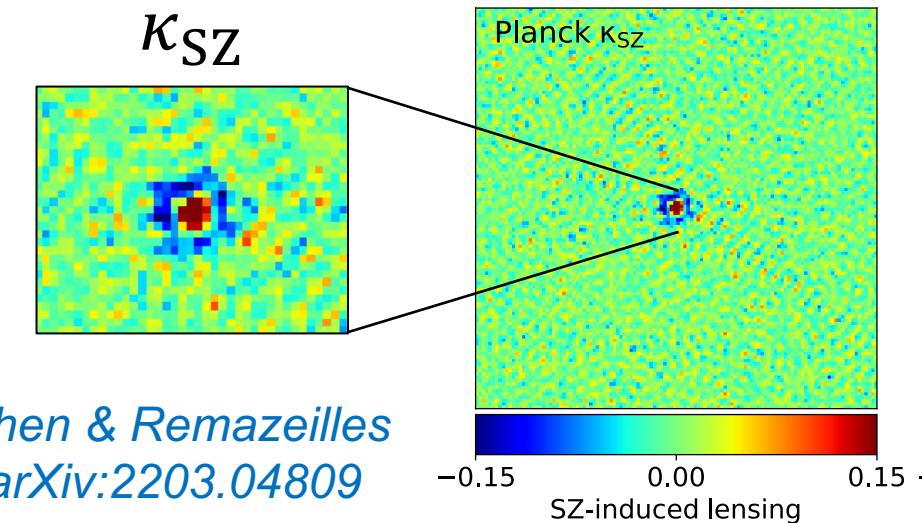
**Thermal SZ cluster residuals in all Planck CMB maps!**

*Major issue for kinetic SZ, CMB lensing, non-Gaussianity, cross-correlations with LSS*

# Planck residual SZ-induced lensing field

Increment of lensing convergence  
in the central part of the clusters

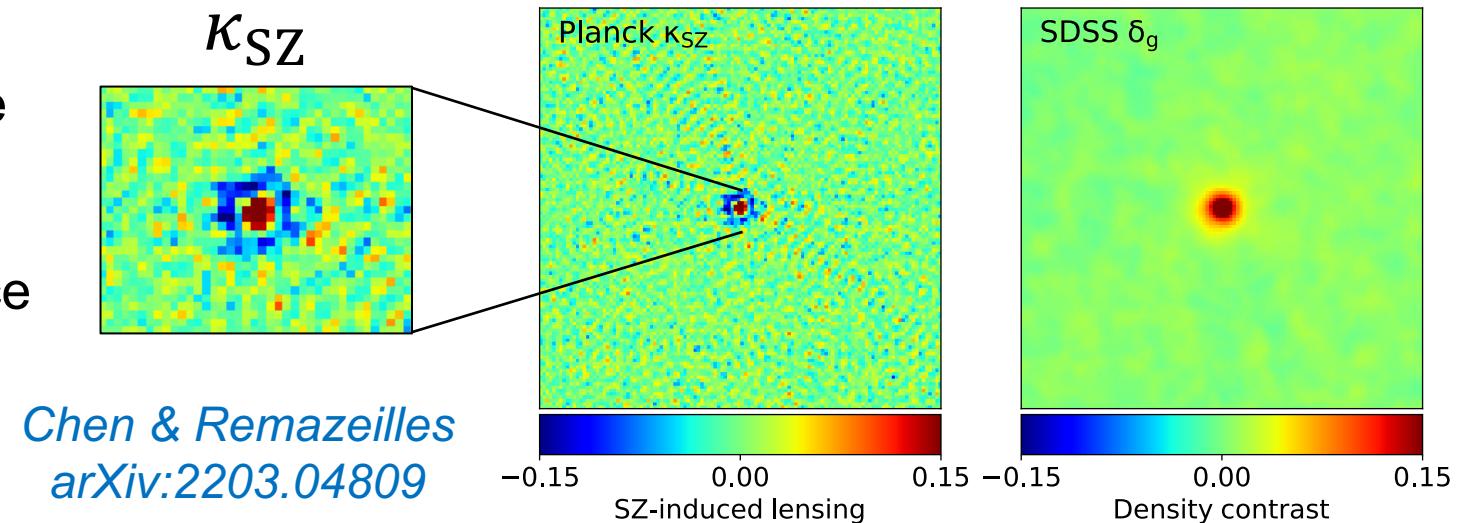
Decrement of lensing convergence  
in the outskirts of the clusters



# Planck residual SZ-induced lensing field

Increment of lensing convergence  
in the central part of the clusters

Decrement of lensing convergence  
in the outskirts of the clusters

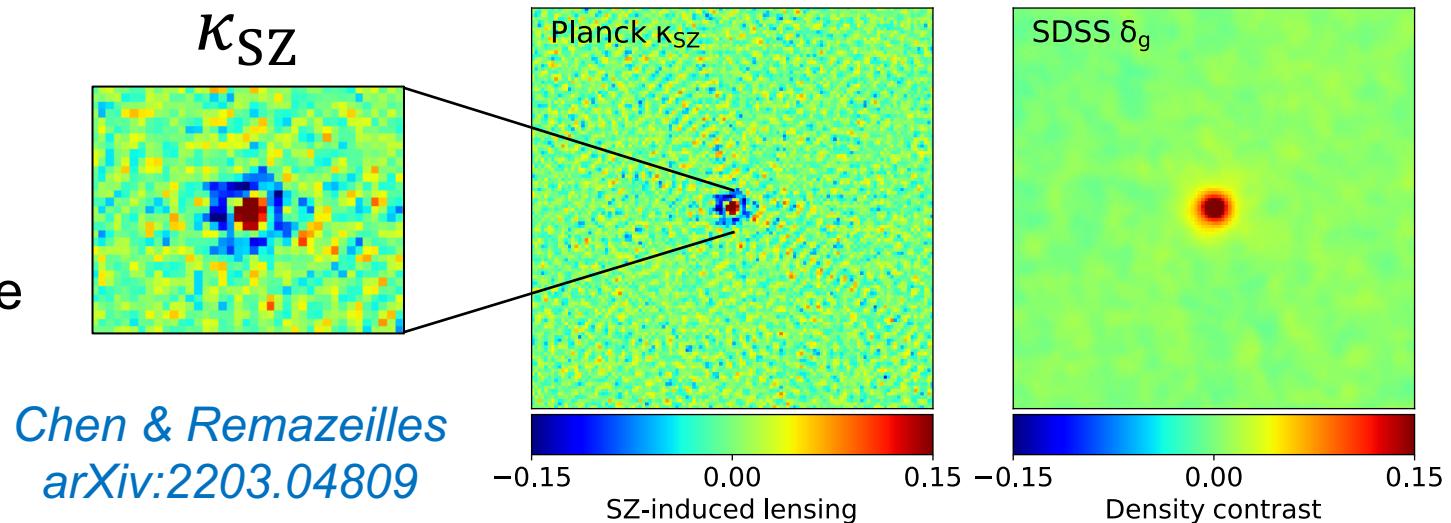


- correlated with LSS at small scale
- anticorrelated with LSS at large scale

# Planck residual SZ-induced lensing field

Increment of lensing convergence  
in the central part of the clusters

Decrement of lensing convergence  
in the outskirts of the clusters



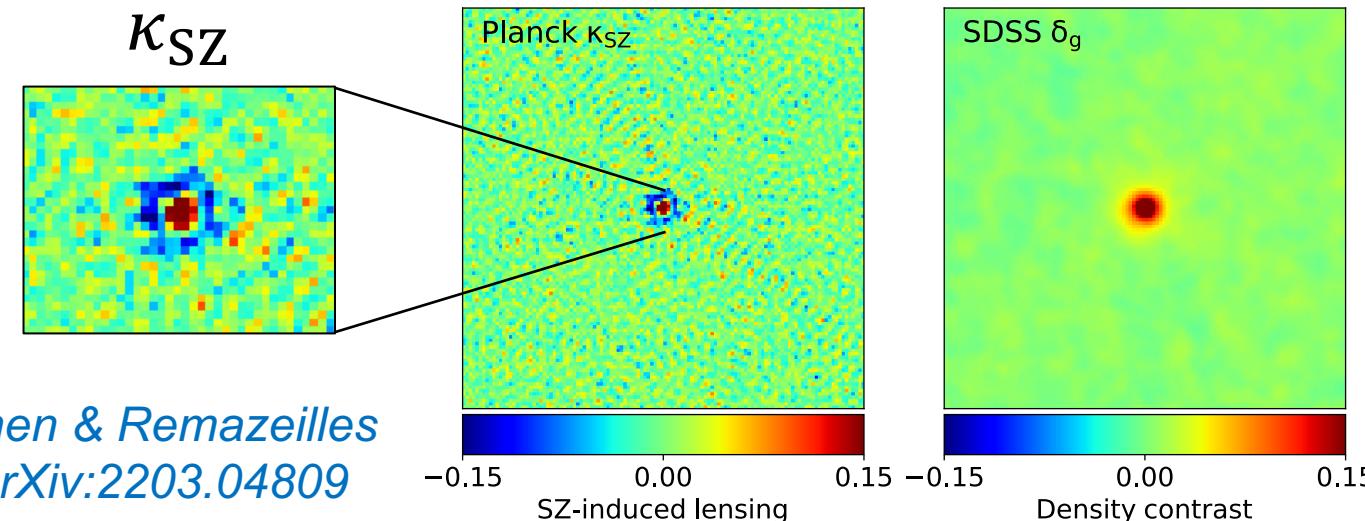
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scale-dependent bias  
on CMB lensing  $\times$  LSS  
cross-correlations

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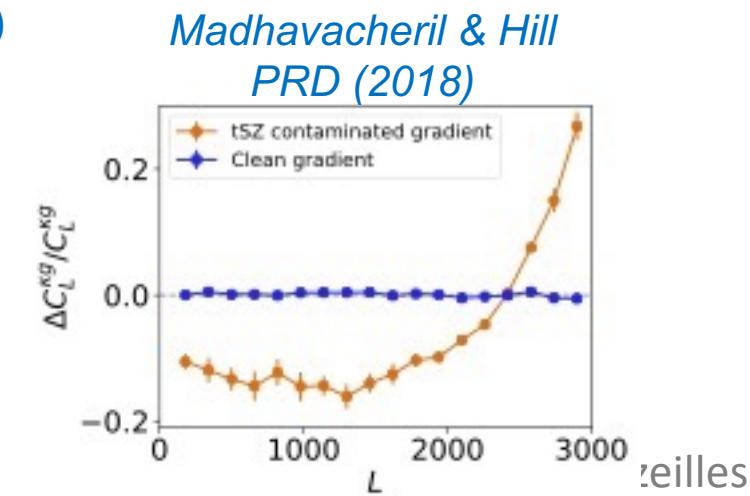
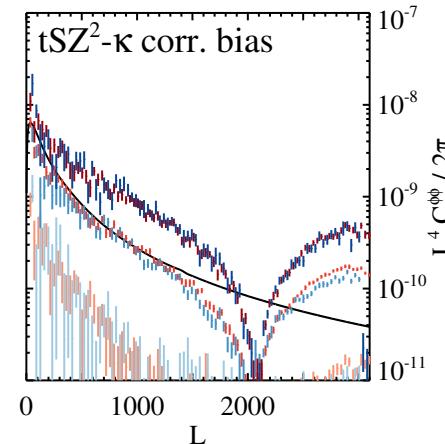
Decrement of lensing convergence  
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- correlated with LSS at small scale
- anticorrelated with LSS at large scale

scale-dependent bias  
on CMB lensing  $\times$  LSS  
cross-correlations

*consistent with  
earlier theoretical  
predictions*

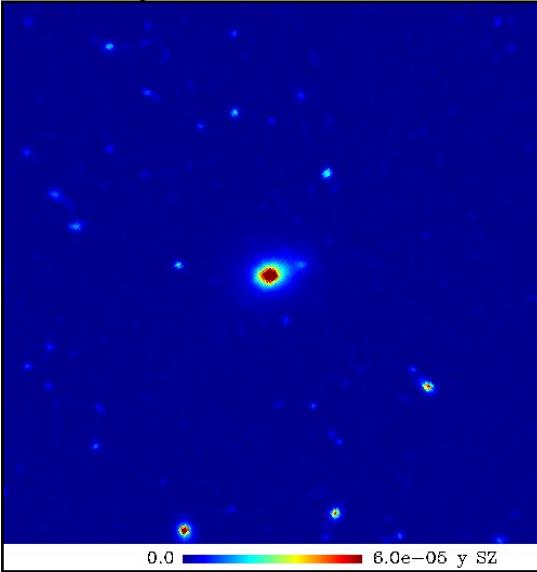


*Foreground correlated  
with the signal*

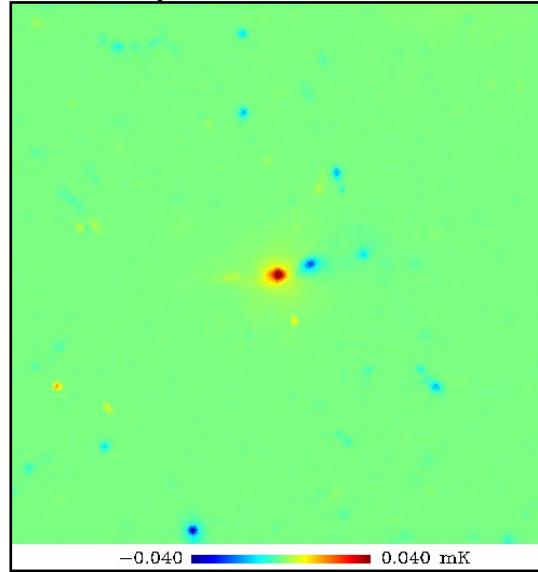
(e.g. SZ in CMB lensing  $\times$  LSS)

# Component separation: standard NILC

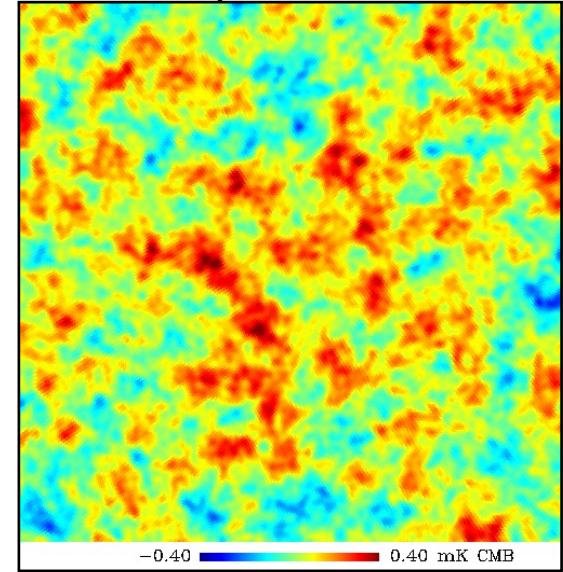
input thermal SZ



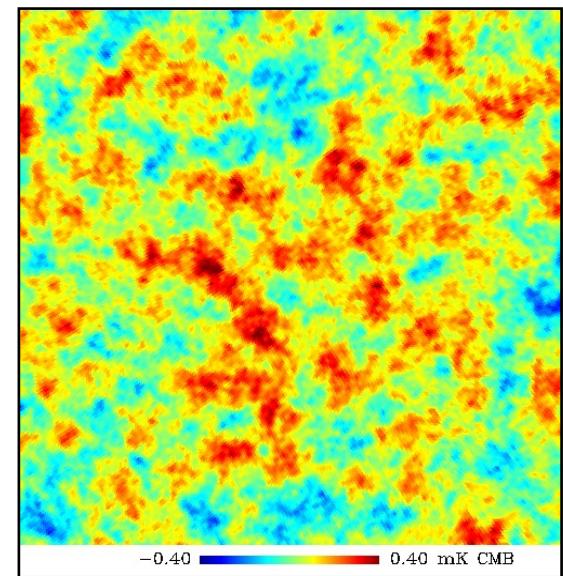
input kinetic SZ



input CMB



ILC



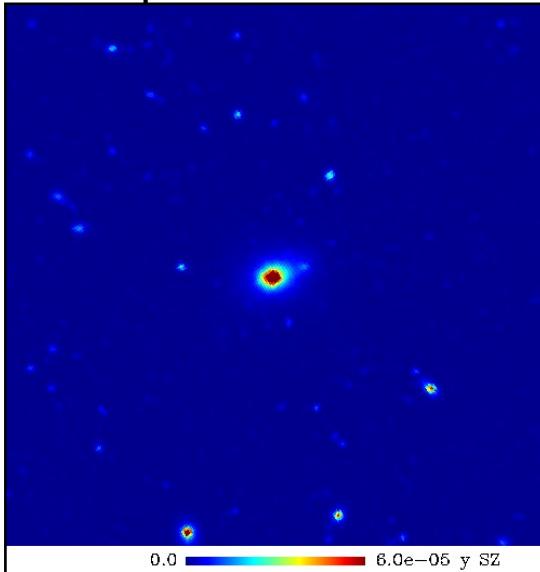
$$\mathbf{w} = \frac{\mathbf{a}^T \mathbf{C}^{-1}}{\mathbf{a}^T \mathbf{C}^{-1} \mathbf{a}}$$

Bennett et al 2003, Tegmark et al 2003  
Eriksen et al 2004, Delabrouille et al 2009

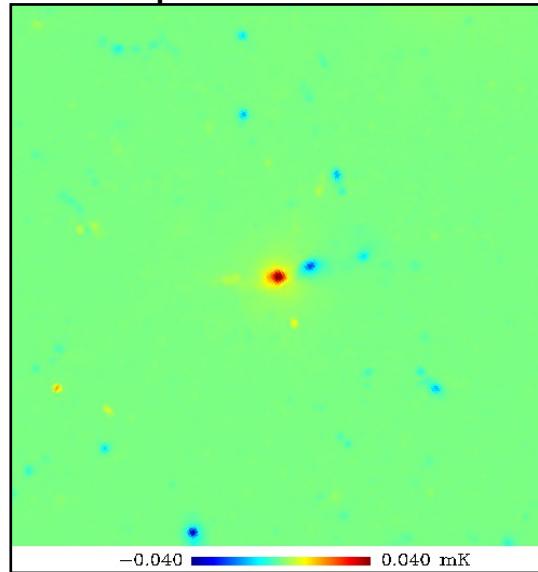
*Minimum-variance solution  
not adequate for certain  
scientific purposes!*

# Component separation: standard NILC

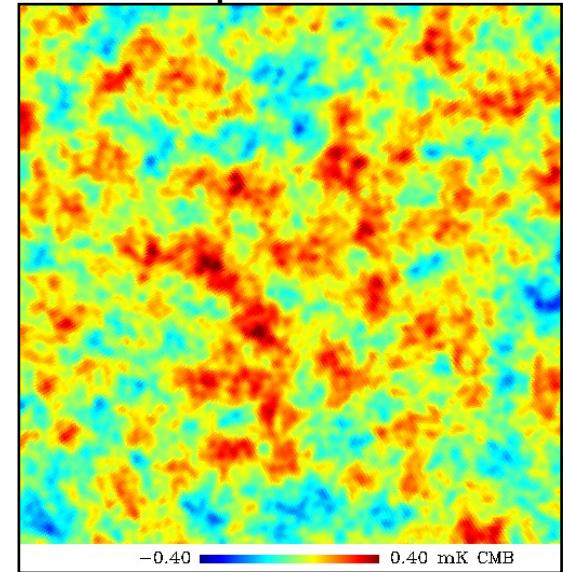
input thermal SZ



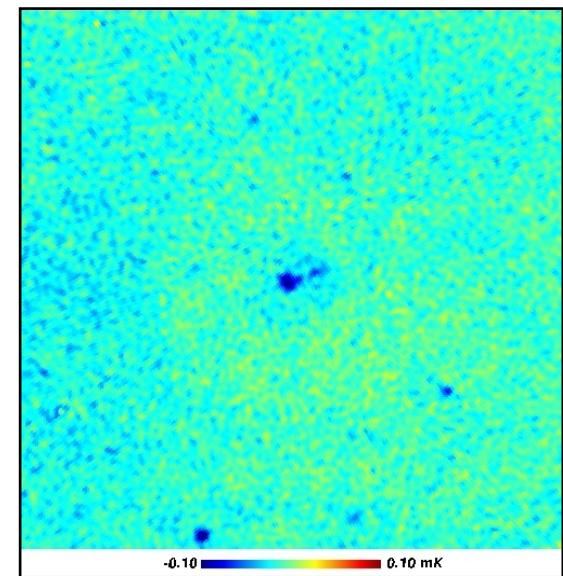
input kinetic SZ



input CMB



Error: ILC – CMB



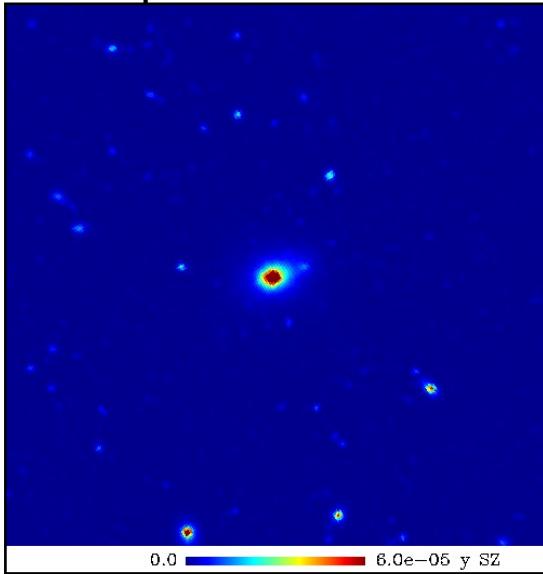
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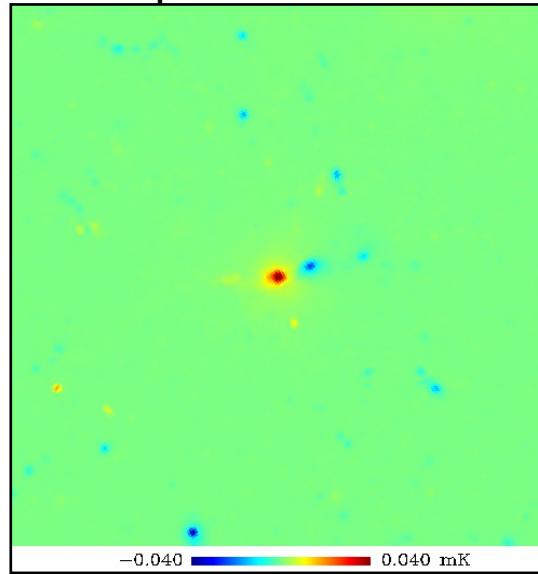
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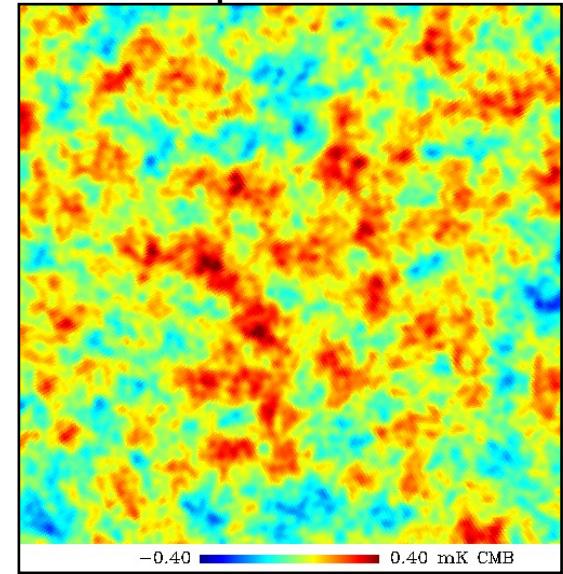
input thermal SZ



input kinetic SZ



input CMB



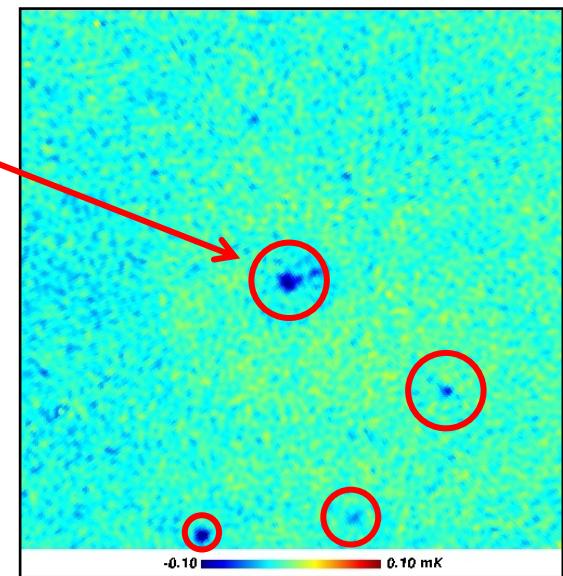
Error: ILC – CMB

**Thermal SZ residuals!**  
**(galaxy clusters in the CMB)**

*Minimum-variance solution  
not adequate for certain  
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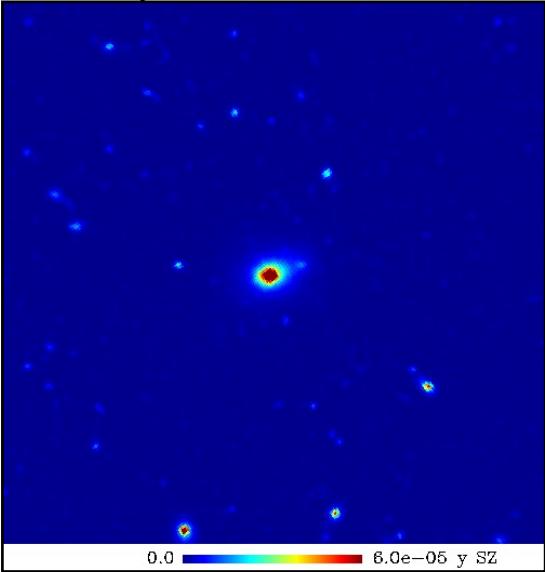
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Eriksen et al 2004, Delabrouille et al 2009

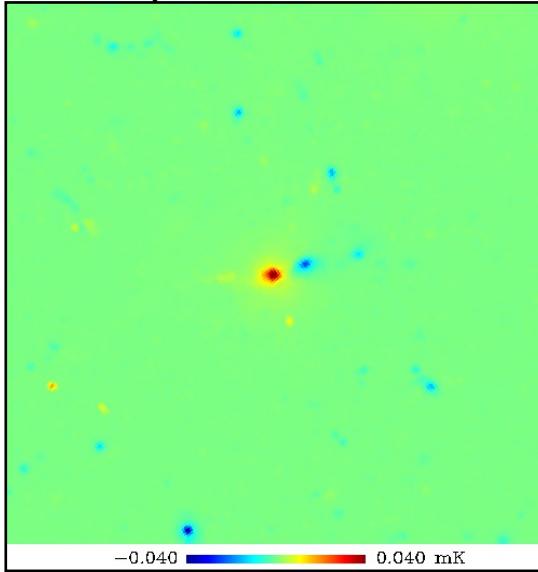


# Component separation: “Constrained ILC” (CILC)

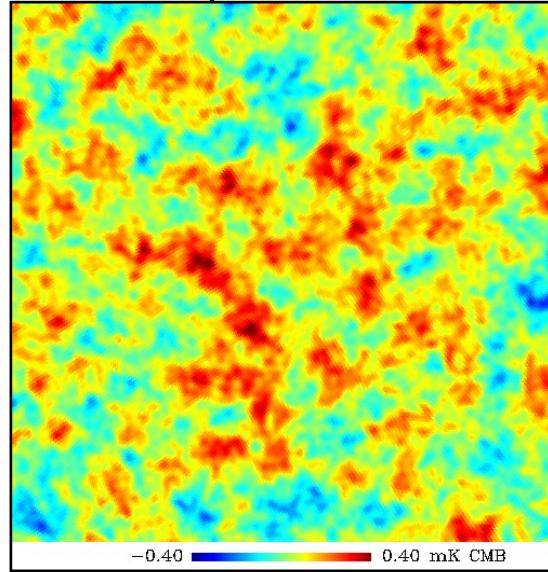
input thermal SZ



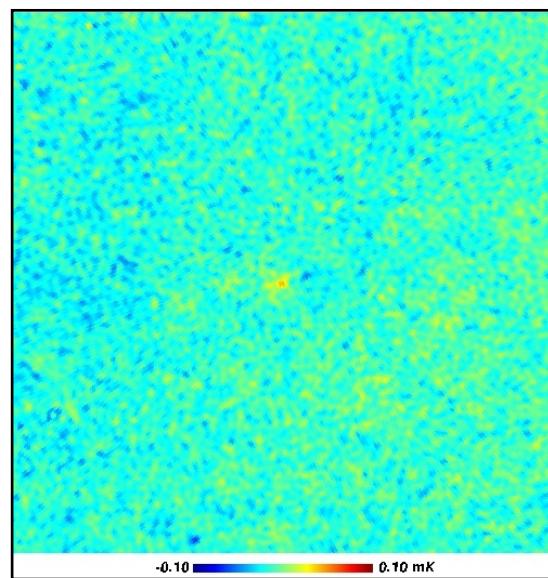
input kinetic SZ



input CMB



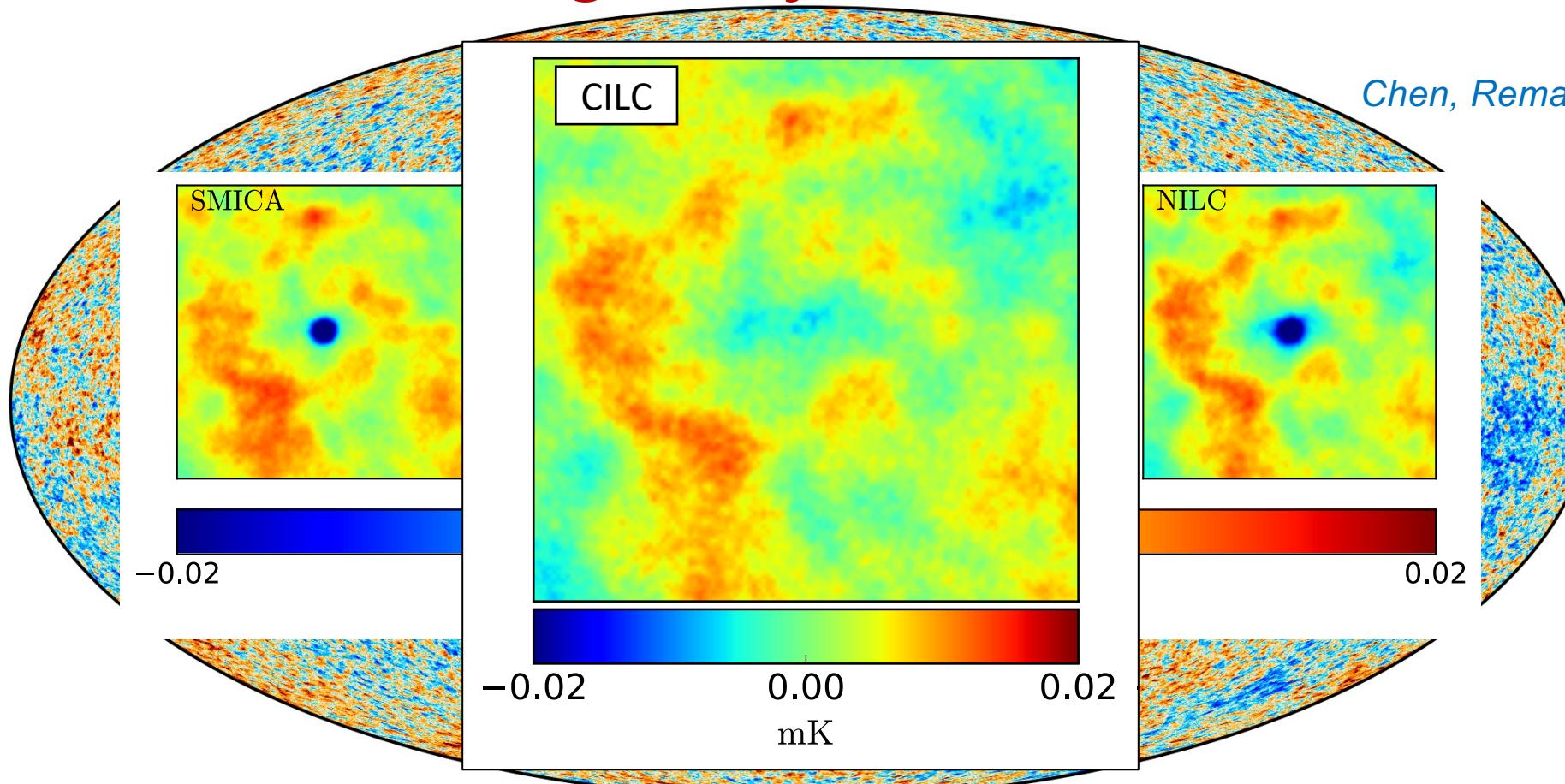
Error: CILC – CMB



$$\mathbf{w} = \frac{(\mathbf{b}^T \mathbf{C}^{-1} \mathbf{b}) \mathbf{a}^T \mathbf{C}^{-1} - (\mathbf{a}^T \mathbf{C}^{-1} \mathbf{b}) \mathbf{b}^T \mathbf{C}^{-1}}{(\mathbf{a}^T \mathbf{C}^{-1} \mathbf{a})(\mathbf{b}^T \mathbf{C}^{-1} \mathbf{b}) - (\mathbf{a}^T \mathbf{C}^{-1} \mathbf{b})^2}$$

Remazeilles, Delabrouille, Cardoso, MNRAS 2011a

# Four consistent Planck CMB maps stacked on galaxy cluster locations



*Chen, Remazeilles, Dickinson  
MNRAS 2018*

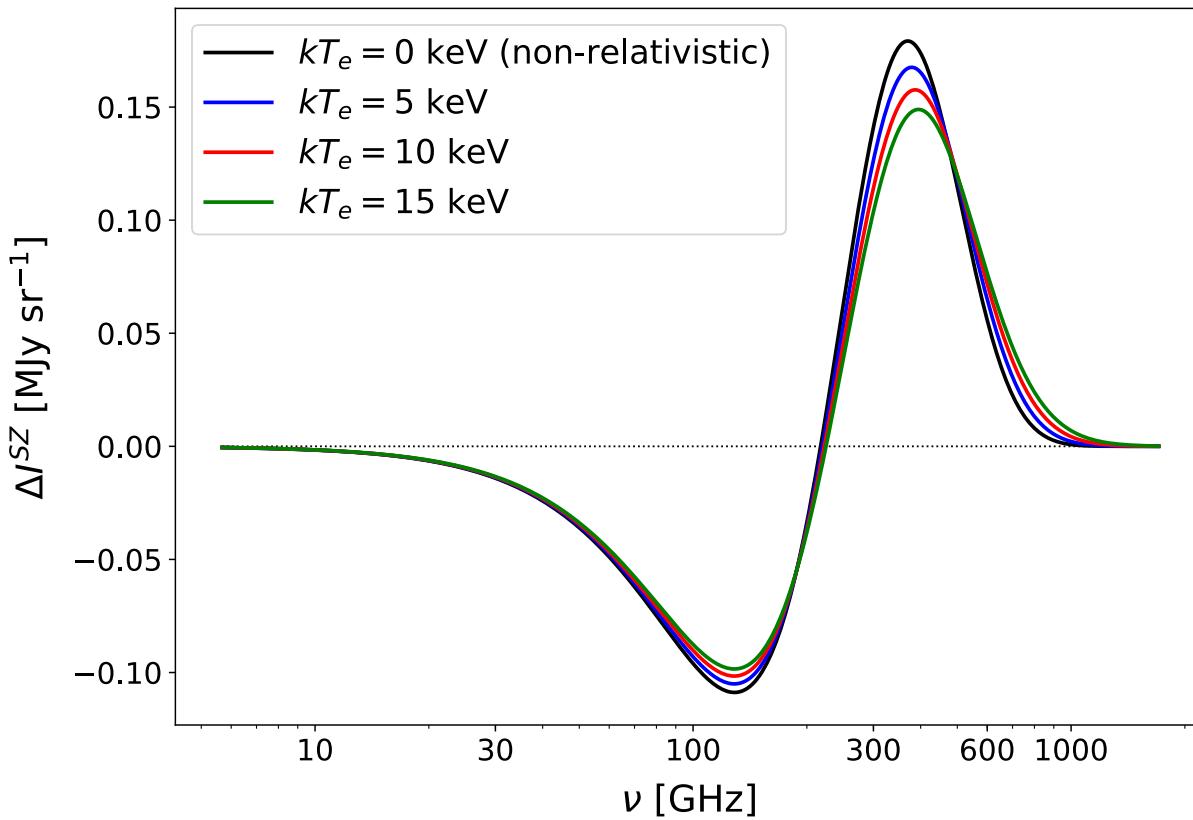
**CILC erases SZ clusters in primary CMB maps  
(at the cost of larger noise)**

*Foreground correlated  
with the signals*

(e.g. relativistic SZ)

# Relativistic SZ effect

$$I_{\nu}^{\text{SZ}} = f(\nu, \mathbf{T}_e(\vec{n})) y(\vec{n})$$



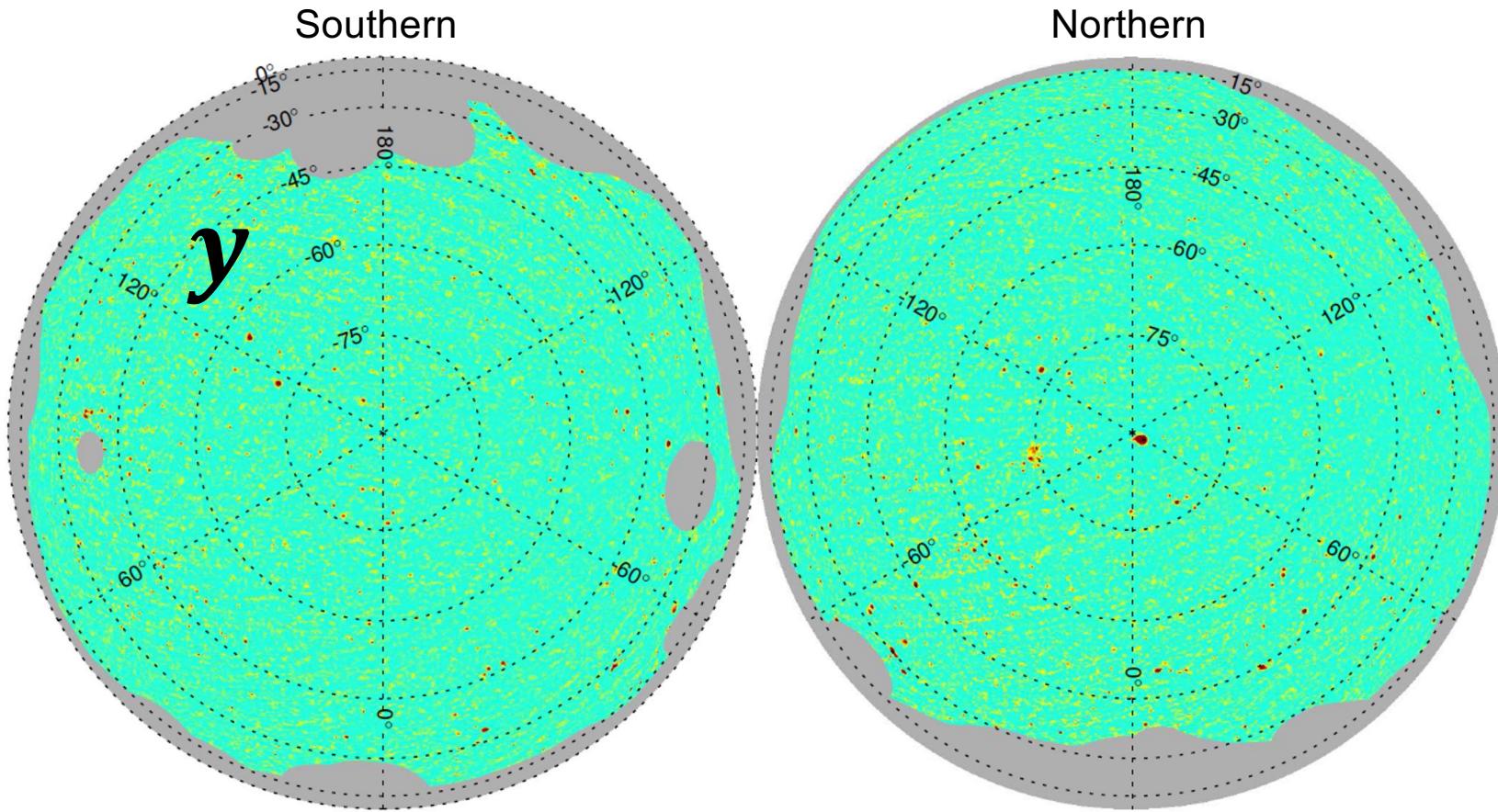
*Relativistic electron temperatures distort  
the shape of the SZ spectrum*

*Two complementary observables*

$$y(\vec{n}), \mathbf{T}_e(\vec{n})$$

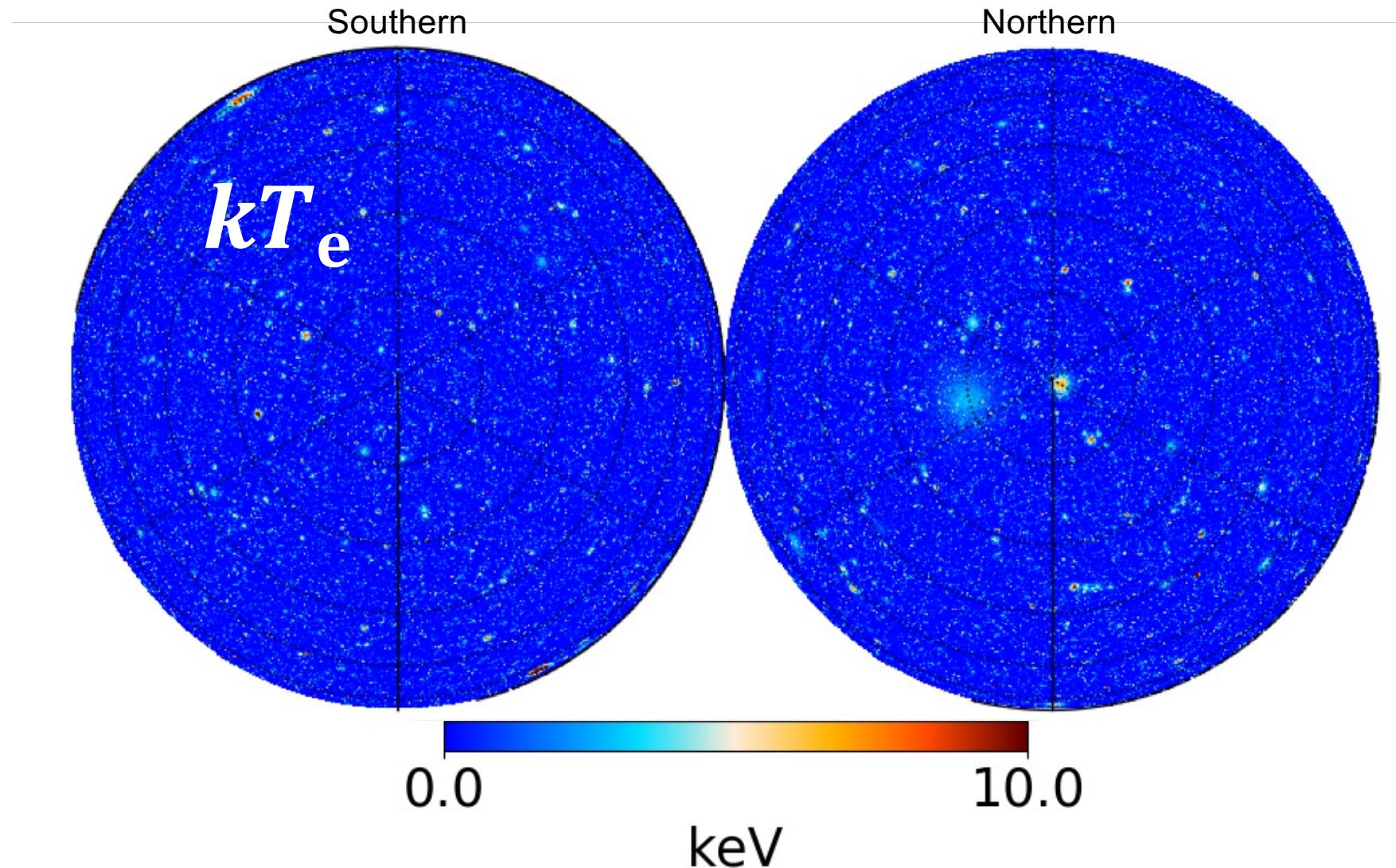
*whose statistics have different  
dependencies on cosmological parameters*

# “First SZ revolution”: The *Planck* Compton $y$ -map



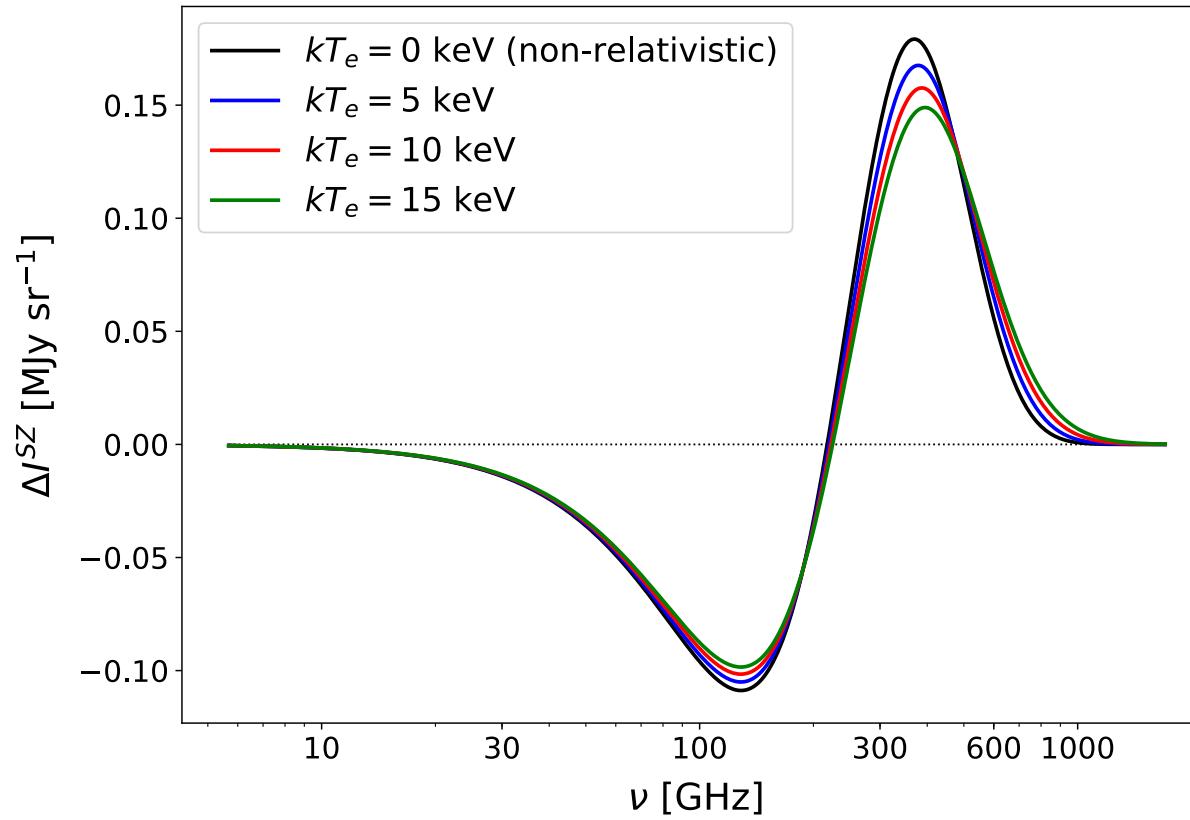
*Planck 2015 results XXII, A&A (2016)*

# “Second SZ revolution”: The electron temperature $T_e$ -map?



# Relativistic SZ temperature corrections

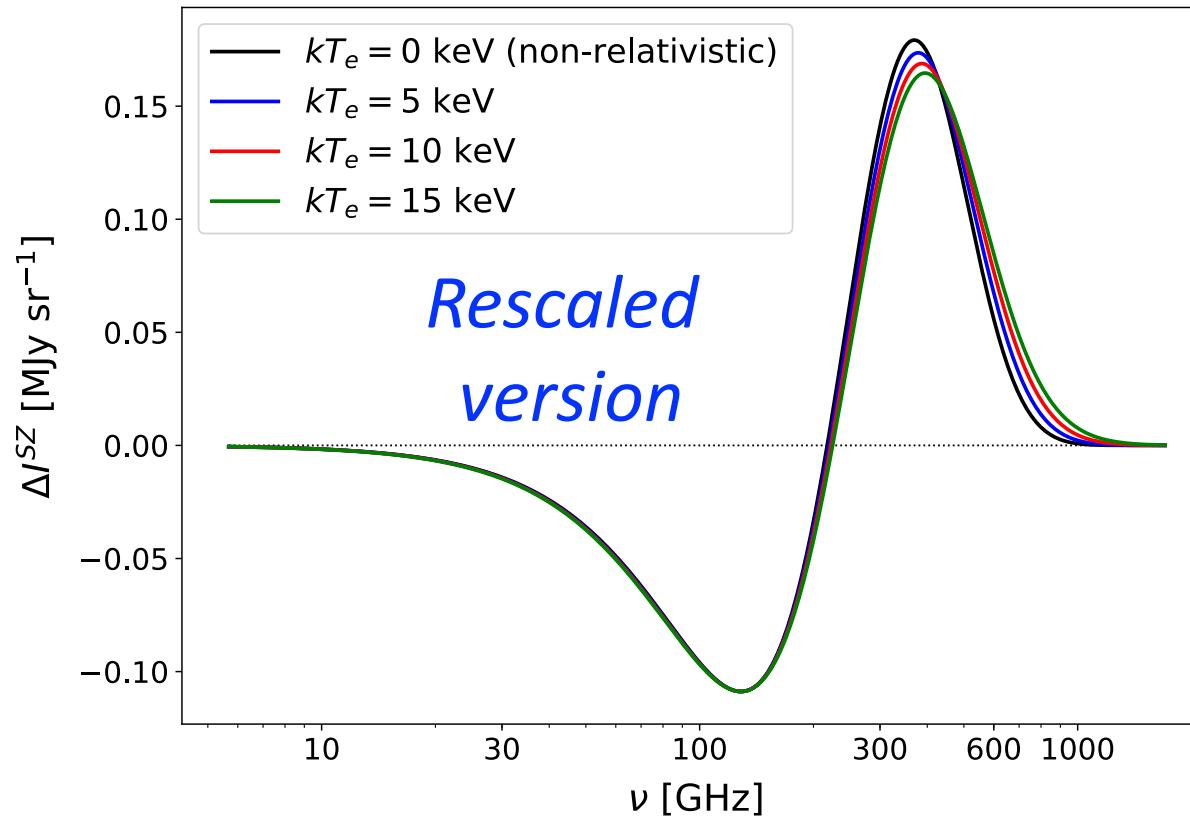
$$I_{\nu}^{\text{SZ}} = f(\nu, T_e(\vec{n})) y(\vec{n})$$



*The spectral signature of SZ emission from galaxy clusters changes with the local electron gas temperature*

# Relativistic SZ temperature corrections

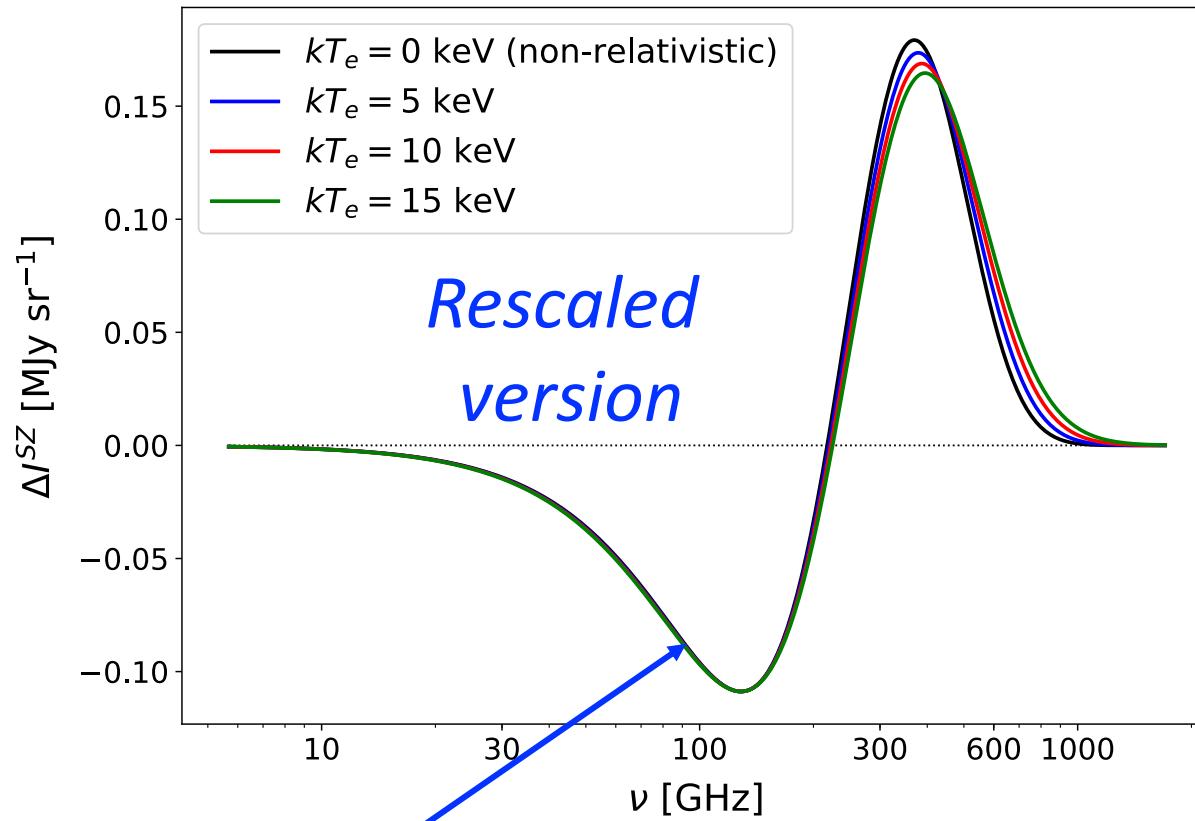
$$I_{\nu}^{\text{SZ}} = f(\nu, T_e(\vec{n})) y(\vec{n})$$



*The spectral signature of SZ emission from galaxy clusters changes with the local electron gas temperature*

# The $y$ - $T_e$ degeneracy at low frequency

$$I_\nu^{\text{SZ}} = f(\nu, T_e(\vec{n})) y(\vec{n})$$

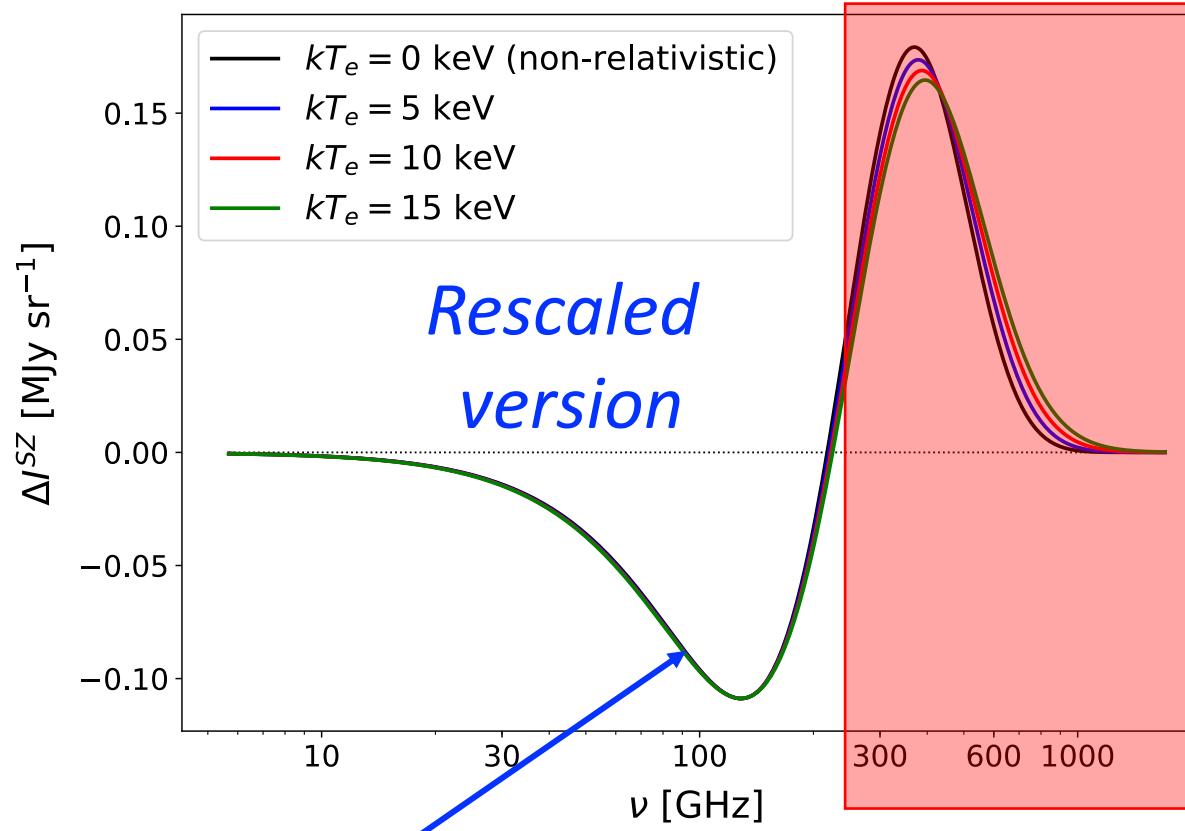


*Spectral shapes are degenerate  
at low frequencies*

*(impossible to disentangle  $y$  and  $T_e$ )*

# The $y$ - $T_e$ degeneracy at low frequency

$$I_\nu^{\text{SZ}} = f(\nu, T_e(\vec{n})) y(\vec{n})$$



*High frequencies are essential to extract rSZ*

*Spectral shapes are degenerate at low frequencies*

*(impossible to disentangle  $y$  and  $T_e$ )*

# rSZ component separation

How to disentangle the  $y$  and  $T_e$  observables  
of the rSZ effect in sky observations?

# SZ temperature moment expansion

$$I_{\nu}^{\text{SZ}}(\vec{n}) = f(\nu, \textcolor{red}{T}_{\mathbf{e}}(\vec{n})) y(\vec{n})$$

# SZ temperature moment expansion

$$I_{\nu}^{\text{SZ}}(\vec{n}) = \mathbf{f}(\nu, \mathbf{T}_e(\vec{n})) \mathbf{y}(\vec{n})$$

Taylor expansion around some pivot temperature  $\bar{T}_e$

$$I_{\nu}^{\text{SZ}}(\vec{n}) = \mathbf{f}(\nu, \bar{T}_e) \mathbf{y}(\vec{n}) + \frac{\partial \mathbf{f}(\nu, \bar{T}_e)}{\partial T_e} (T_e(\vec{n}) - \bar{T}_e) \mathbf{y}(\vec{n}) + \mathcal{O}(T_e^2)$$

# SZ temperature moment expansion

$$I_{\nu}^{\text{SZ}}(\vec{n}) = f(\nu, \mathbf{T}_e(\vec{n})) y(\vec{n})$$

Taylor expansion around some pivot temperature  $\bar{T}_e$

$$I_{\nu}^{\text{SZ}}(\vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y} \underbrace{y(\vec{n})}_{\text{component}} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } yT_e} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e \text{ component}} + \mathcal{O}(T_e^2)$$

# SZ temperature moment expansion

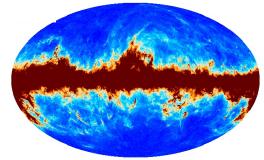
$$I_{\nu}^{\text{SZ}}(\vec{n}) = f(\nu, \mathbf{T}_e(\vec{n})) y(\vec{n})$$

Taylor expansion around some pivot temperature  $\bar{T}_e$

$$I_{\nu}^{\text{SZ}}(\vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y} \underbrace{y(\vec{n})}_{y \text{ component}} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } yT_e} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e \text{ component}} + \mathcal{O}(T_e^2)$$

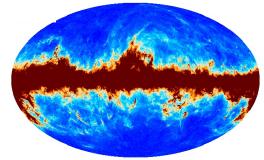
Two distinct components of emission,  $y$  and  $y\Delta T_e$ ,  
with different spectral signatures

# rSZ component separation



$$d(\nu, \vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y(\vec{n})} \underbrace{y(\vec{n})}_{y(\vec{n})} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } y\Delta T_e(\vec{n})} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e(\vec{n})} + \underbrace{N(\nu, \vec{n})}_{\text{foregrounds} + \text{noise}}$$

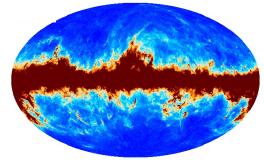
# rSZ component separation



$$d(\nu, \vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y(\vec{n})} \underbrace{y(\vec{n})}_{y(\vec{n})} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } y\Delta T_e(\vec{n})} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e(\vec{n})} + \underbrace{N(\nu, \vec{n})}_{\text{foregrounds} + \text{noise}}$$

*signal of interest  
(rSZ)*

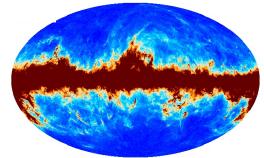
# rSZ component separation



$$d(\nu, \vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y(\vec{n})} \underbrace{y(\vec{n})}_{y(\vec{n})} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } y\Delta T_e(\vec{n})} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e(\vec{n})} + \underbrace{N(\nu, \vec{n})}_{\text{foregrounds} + \text{noise}}$$

*foreground correlated with the signal!*      *signal of interest (rSZ)*

# rSZ component separation

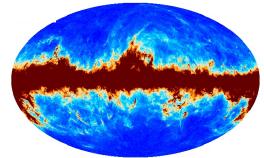


$$d(\nu, \vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y(\vec{n})} \underbrace{y(\vec{n})}_{y(\vec{n})} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } y\Delta T_e(\vec{n})} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e(\vec{n})} + \underbrace{N(\nu, \vec{n})}_{\text{foregrounds} + \text{noise}}$$

Component separation with the CILC method ([Remazeilles et al MNRAS 2011a](#))

$$\widehat{y\Delta T_e}(\vec{n}) = \sum_{\nu} w(\nu) d(\nu, \vec{n}) \quad \text{such that} \quad \left\{ \begin{array}{l} \langle (\widehat{y\Delta T_e})^2 \rangle \text{ of minimum variance} \\ \sum_{\nu} w(\nu) \frac{\partial f(\nu, \bar{T}_e)}{\partial T_e} = 1 \\ \sum_{\nu} w(\nu) f(\nu, \bar{T}_e) = 0 \end{array} \right.$$

# rSZ component separation



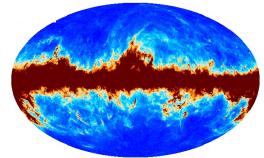
$$d(\nu, \vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y(\vec{n})} \underbrace{y(\vec{n})}_{y(\vec{n})} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } y\Delta T_e(\vec{n})} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e(\vec{n})} + \underbrace{N(\nu, \vec{n})}_{\text{foregrounds + noise}}$$

Component separation with the CILC method ([Remazeilles et al MNRAS 2011a](#))

$$\widehat{y\Delta T_e}(\vec{n}) = \sum_{\nu} w(\nu) d(\nu, \vec{n}) \quad \text{such that} \quad \left\{ \begin{array}{l} \langle (\widehat{y\Delta T_e})^2 \rangle \text{ of minimum variance} \\ \boxed{\sum_{\nu} w(\nu) \frac{\partial f(\nu, \bar{T}_e)}{\partial T_e} = 1} \\ \sum_{\nu} w(\nu) f(\nu, \bar{T}_e) = 0 \end{array} \right.$$

Guarantees the conservation of the signal of interest  $y\Delta T_e$

# rSZ component separation



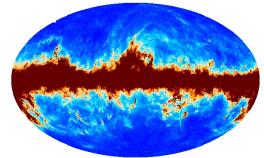
$$d(\nu, \vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y(\vec{n})} \underbrace{y(\vec{n})}_{y(\vec{n})} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } y\Delta T_e(\vec{n})} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e(\vec{n})} + \underbrace{N(\nu, \vec{n})}_{\text{foregrounds} + \text{noise}}$$

Component separation with the CILC method ([Remazeilles et al MNRAS 2011a](#))

$$\widehat{y\Delta T_e}(\vec{n}) = \sum_{\nu} w(\nu) d(\nu, \vec{n}) \quad \text{such that} \quad \left\{ \begin{array}{l} \langle (\widehat{y\Delta T_e})^2 \rangle \text{ of minimum variance} \\ \sum_{\nu} w(\nu) \frac{\partial f(\nu, \bar{T}_e)}{\partial T_e} = 1 \\ \boxed{\sum_{\nu} w(\nu) f(\nu, \bar{T}_e) = 0} \end{array} \right.$$

Guarantees the cancellation of  $y$  residuals in the  $y\Delta T_e$  map

# rSZ component separation



$$d(\nu, \vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y(\vec{n})} \underbrace{y(\vec{n})}_{y(\vec{n})} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } y\Delta T_e(\vec{n})} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e(\vec{n})} + \underbrace{N(\nu, \vec{n})}_{\text{foregrounds} + \text{noise}}$$

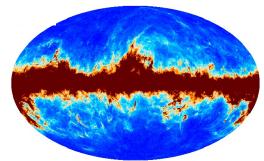
Component separation with the CILC method ([Remazeilles et al MNRAS 2011a](#))

$$\widehat{y\Delta T_e}(\vec{n}) = \sum_{\nu} w(\nu) d(\nu, \vec{n}) \quad \text{such that}$$

$$\left\{ \begin{array}{l} \langle (\widehat{y\Delta T_e})^2 \rangle \text{ of minimum variance} \\ \sum_{\nu} w(\nu) \frac{\partial f(\nu, \bar{T}_e)}{\partial T_e} = 1 \\ \sum_{\nu} w(\nu) f(\nu, \bar{T}_e) = 0 \end{array} \right.$$

Guarantees the mitigation of foregrounds and noise

# rSZ component separation



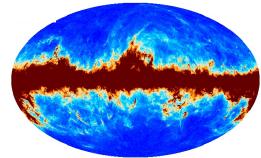
$$d(\nu, \vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y(\vec{n})} \underbrace{y(\vec{n})}_{y(\vec{n})} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } y\Delta T_e(\vec{n})} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e(\vec{n})} + \underbrace{N(\nu, \vec{n})}_{\text{foregrounds} + \text{noise}}$$

Component separation with the CILC method ([Remazeilles et al MNRAS 2011a](#))

$$\widehat{y\Delta T_e}(\vec{n}) = \sum_{\nu} w(\nu) d(\nu, \vec{n}) \quad \text{such that} \quad \begin{cases} \langle (\widehat{y\Delta T_e})^2 \rangle \text{ of minimum variance} \\ \sum_{\nu} w(\nu) \frac{\partial f(\nu, \bar{T}_e)}{\partial T_e} = 1 \\ \sum_{\nu} w(\nu) f(\nu, \bar{T}_e) = 0 \end{cases}$$

$$\Rightarrow \widehat{y\Delta T_e}(\vec{n}) = \underbrace{(w \cdot f)}_{=0} y(\vec{n}) + \underbrace{(w \cdot \partial_{T_e} f)}_{=1} (T_e(\vec{n}) - \bar{T}_e) y(\vec{n}) + \underbrace{w \cdot N}_{\text{minimized}}$$

# rSZ component separation



$$d(\nu, \vec{n}) = \underbrace{f(\nu, \bar{T}_e)}_{\text{spectrum of } y(\vec{n})} \underbrace{y(\vec{n})}_{y(\vec{n})} + \underbrace{\frac{\partial f(\nu, \bar{T}_e)}{\partial T_e}}_{\text{spectrum of } y\Delta T_e(\vec{n})} \underbrace{(T_e(\vec{n}) - \bar{T}_e)y(\vec{n})}_{y\Delta T_e(\vec{n})} + \underbrace{N(\nu, \vec{n})}_{\text{foregrounds} + \text{noise}}$$

Component separation with the CILC method ([Remazeilles et al MNRAS 2011a](#))

$$\widehat{y\Delta T_e}(\vec{n}) = \sum_{\nu} w(\nu) d(\nu, \vec{n}) \quad \text{such that} \quad \left\{ \begin{array}{l} \langle (\widehat{y\Delta T_e})^2 \rangle \text{ of minimum variance} \\ \sum_{\nu} w(\nu) \frac{\partial f(\nu, \bar{T}_e)}{\partial T_e} = 1 \\ \sum_{\nu} w(\nu) f(\nu, \bar{T}_e) = 0 \end{array} \right.$$

$$\Rightarrow \widehat{y\Delta T_e}(\vec{n}) = \underline{(T_e(\vec{n}) - \bar{T}_e) y(\vec{n})} + w \cdot N \quad \textbf{\color{red}{$T_e$-modulated $y$-map!}}$$

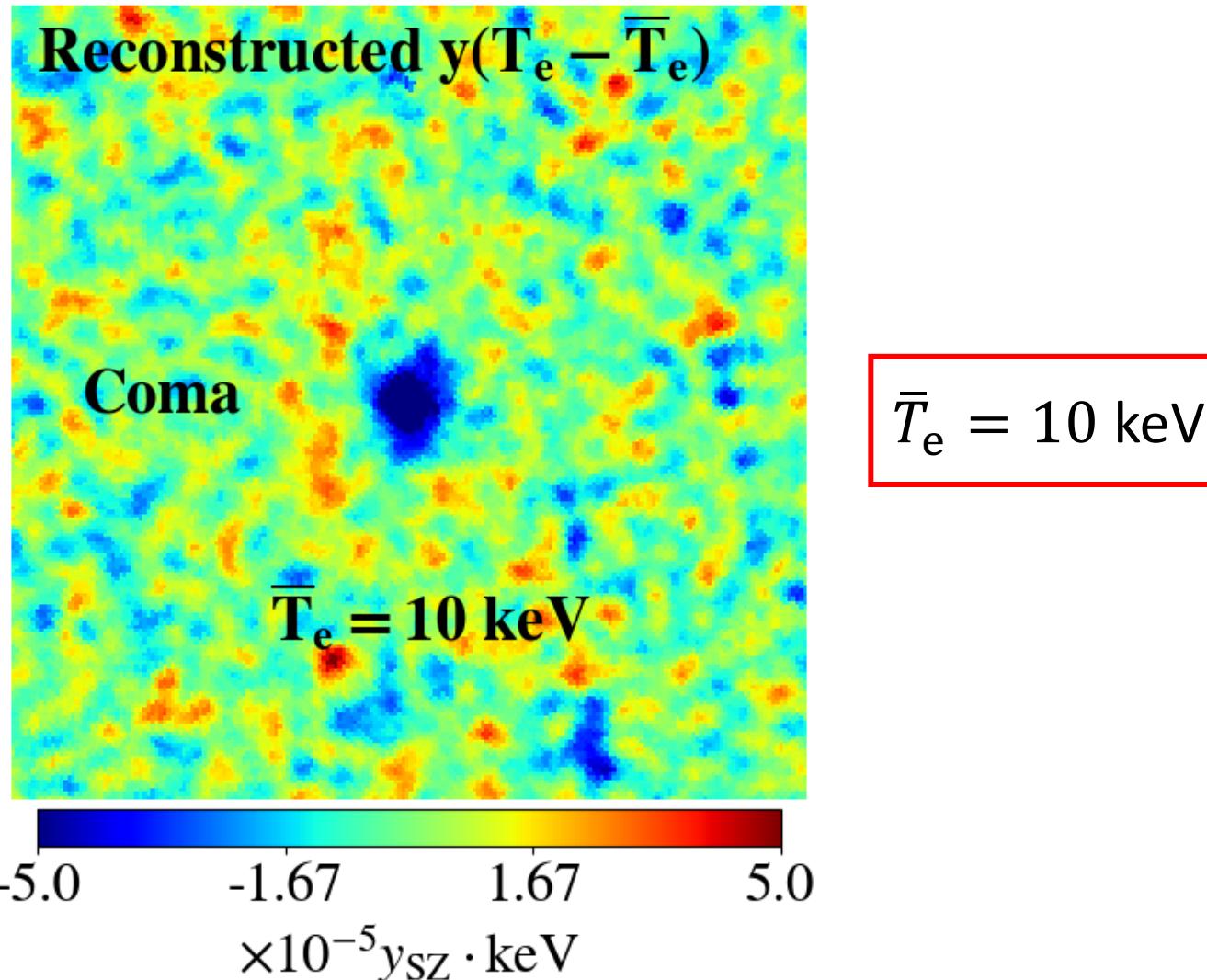
# Why is this new SZ observable so interesting?

$$y\Delta T_e(\vec{n}) \equiv y(\vec{n})(T_e(\vec{n}) - \bar{T}_e)$$

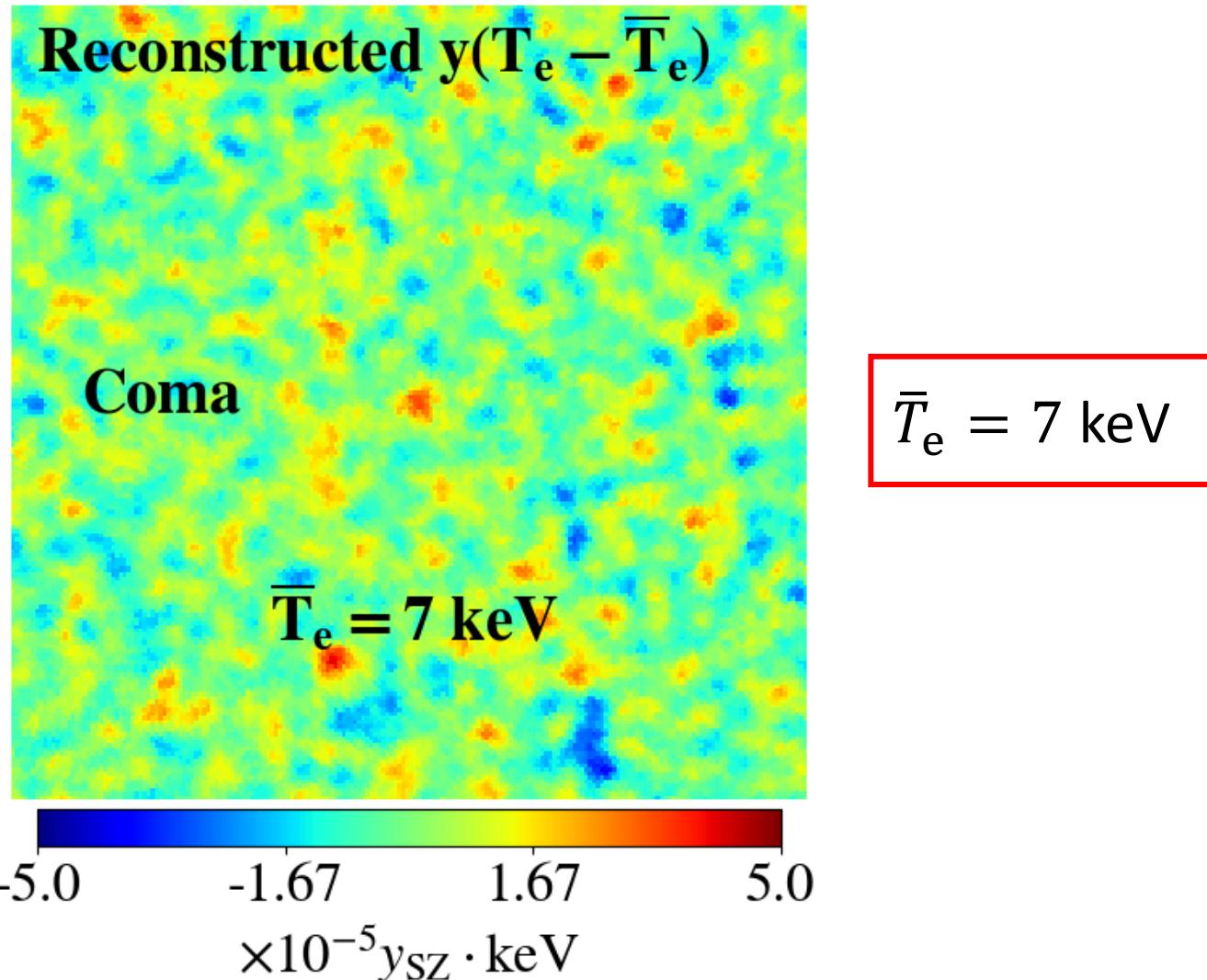
Varying the pivot temperature  $\bar{T}_e$  in the analysis allows us to perform a real **temperature spectroscopy** of the cluster:

- Decrement if actual temperature  $T_e(\vec{n}) < \bar{T}_e$
- Increment if actual temperature  $T_e(\vec{n}) > \bar{T}_e$
- Null if actual temperature  $T_e(\vec{n}) \simeq \bar{T}_e$

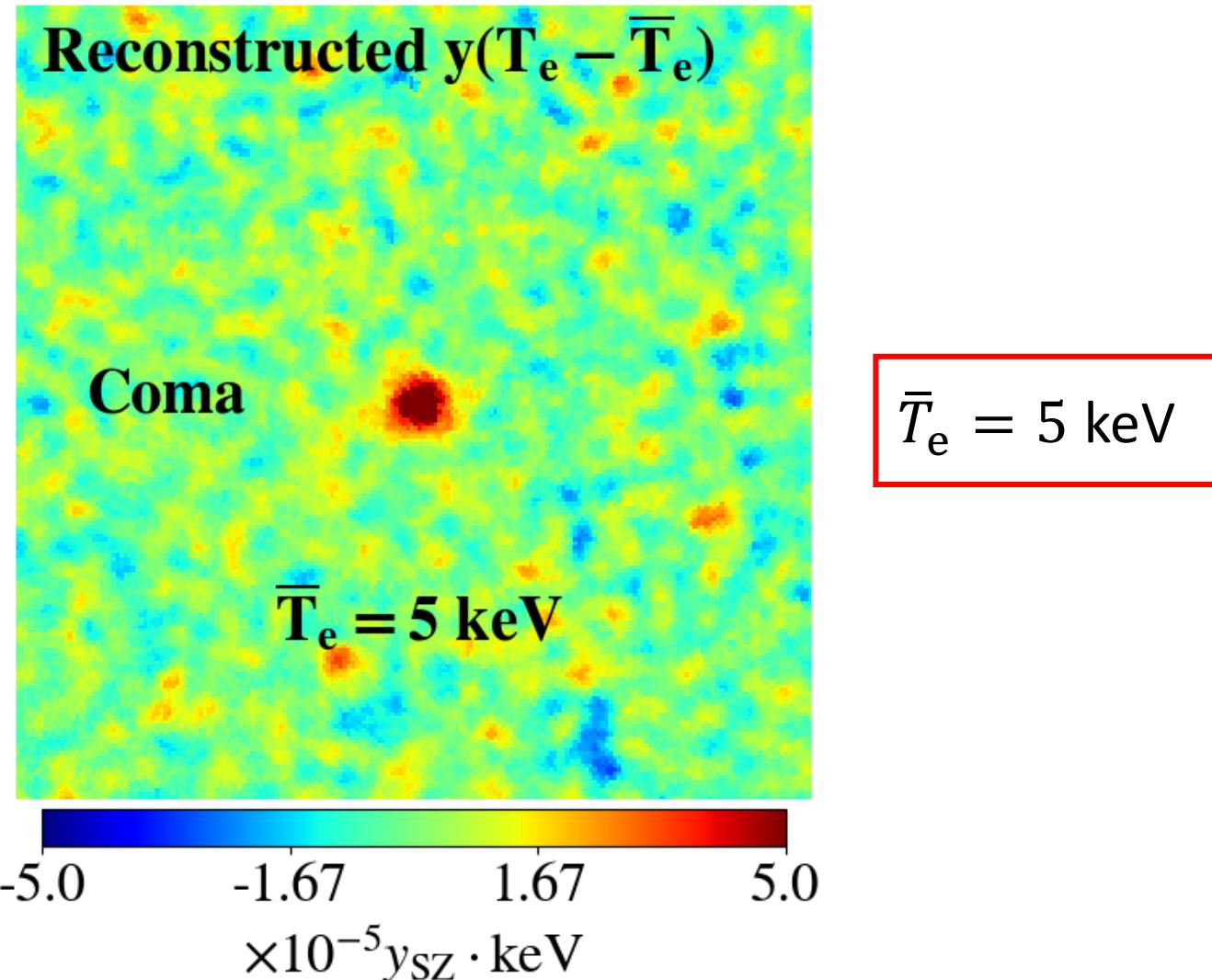
# Cluster spectroscopy across temperature



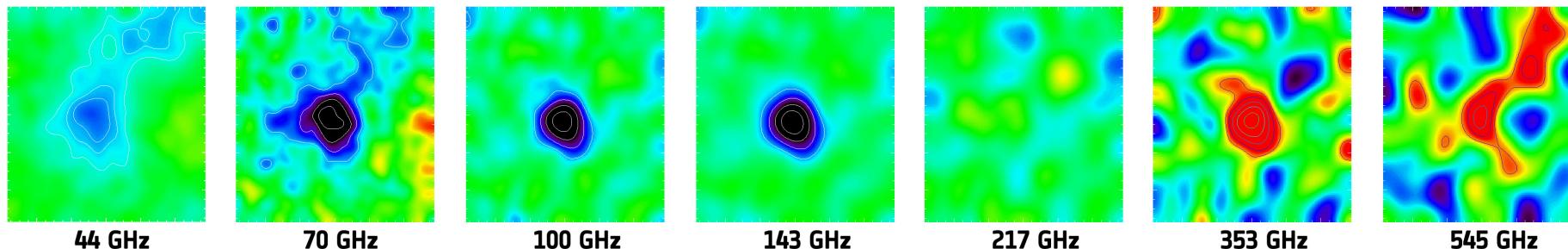
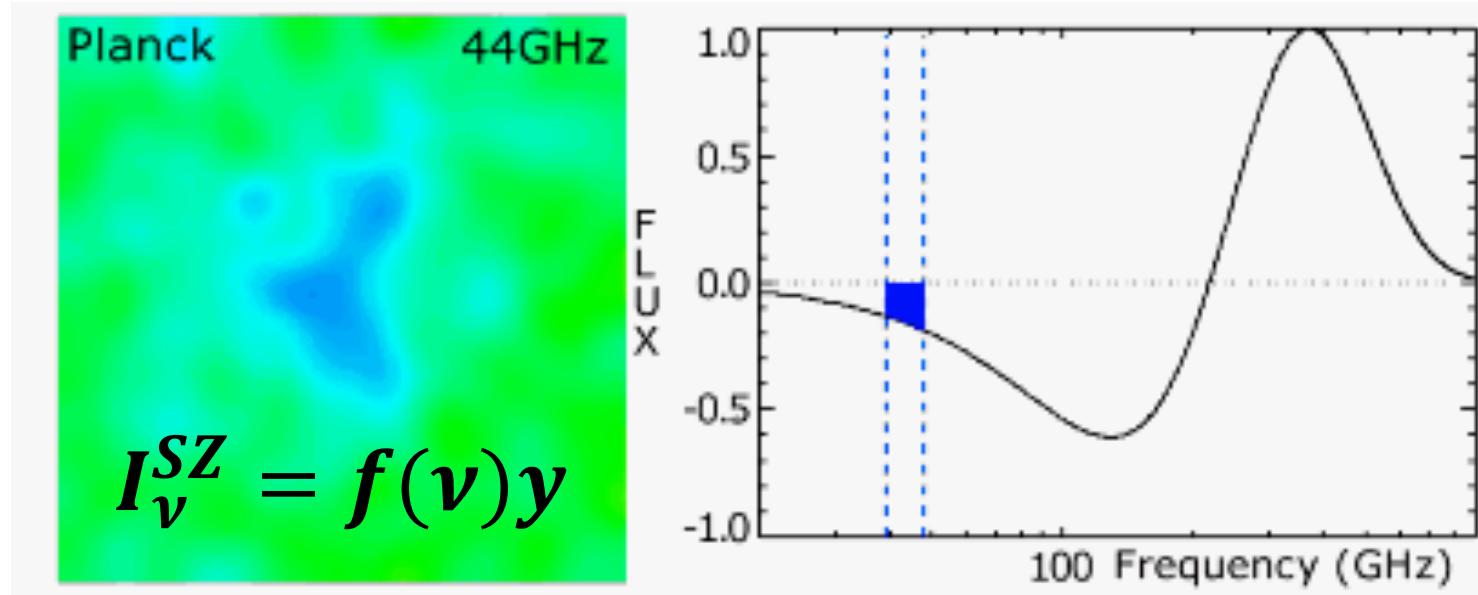
# Cluster spectroscopy across temperature



# Cluster spectroscopy across temperature



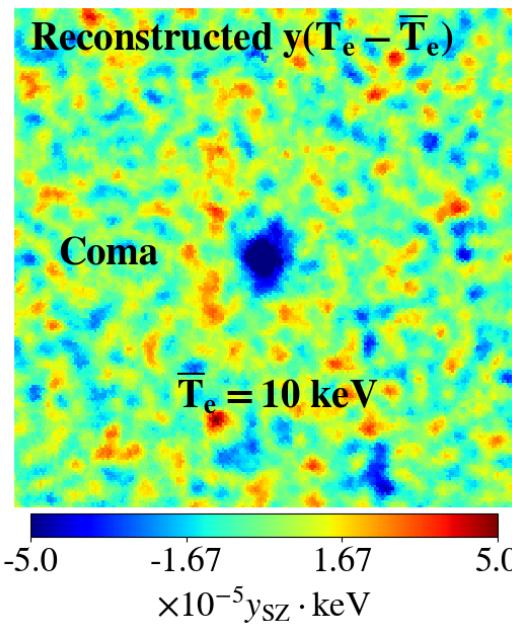
# “First SZ revolution”: cluster spectroscopy across frequencies



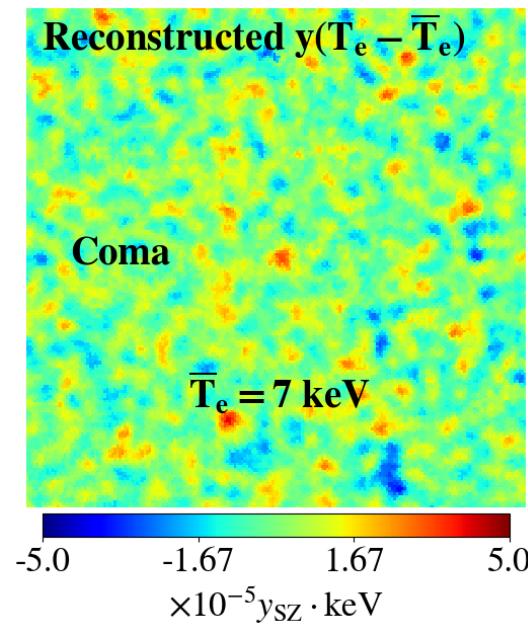
Credit: ESA/Planck Collaboration

# “Second SZ revolution”: cluster spectroscopy across temperatures

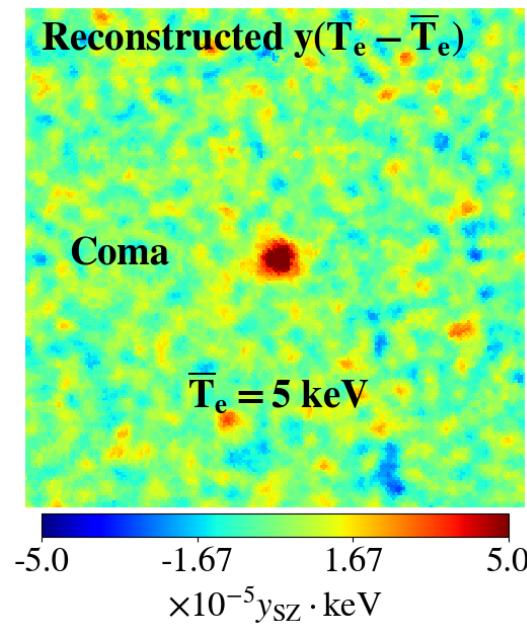
Recovered  $y(T_e - \bar{T}_e)$ -map for different pivots



colder than 10 keV  
(decrement)



closer to 7 keV  
(null)

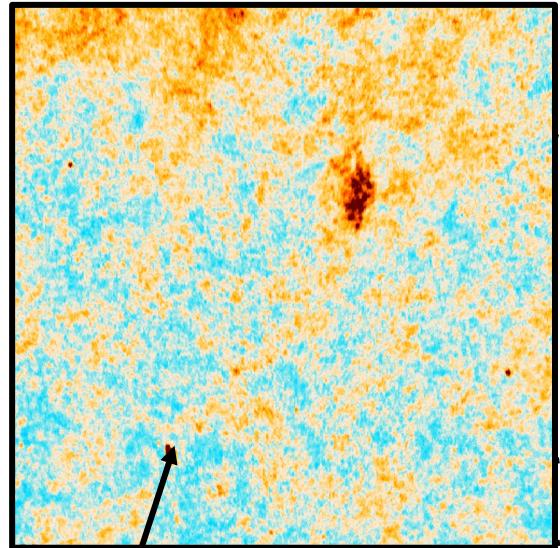


hotter than 5 keV  
(increment)

# *Spectral degeneracies*

(e.g. dust and CIB)

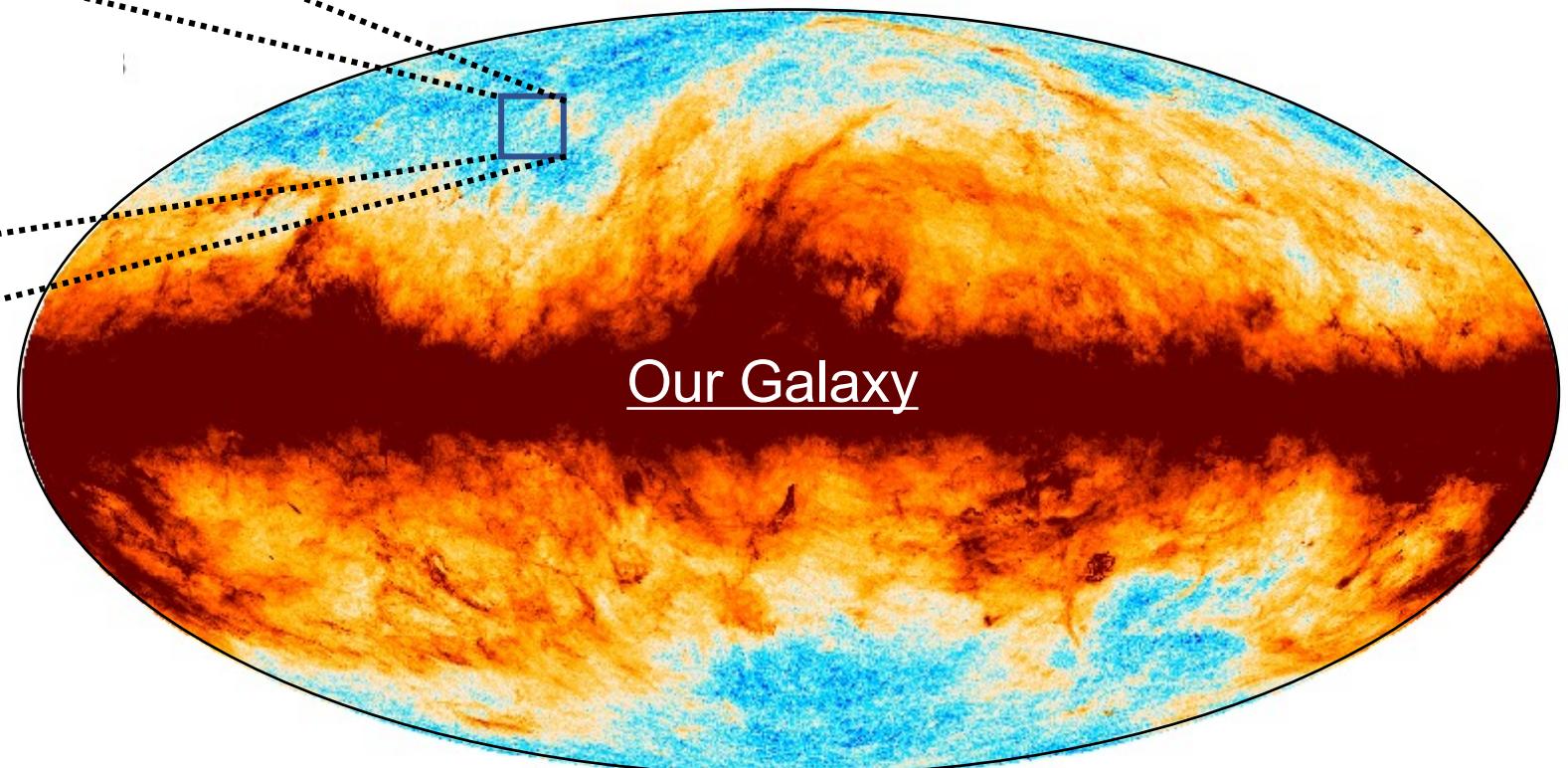
# Planck 2013 map of Galactic dust



CIB contamination  
at small scales!

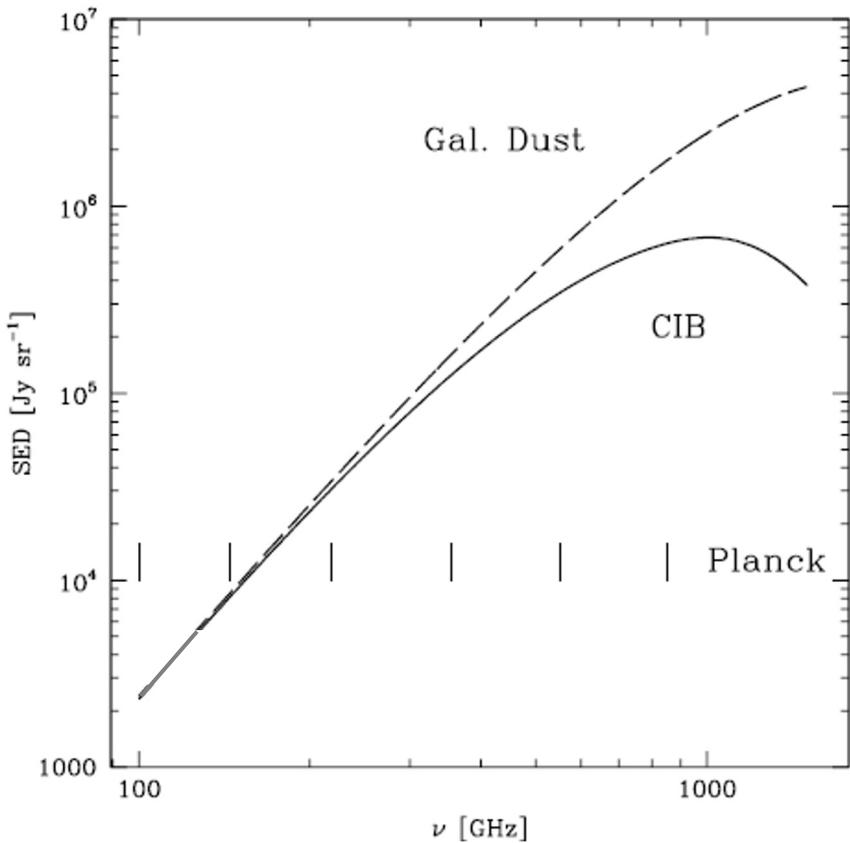
(background galaxies)

*Planck 2013 results XI, A&A 2014*



# Dust-CIB spectral degeneracy

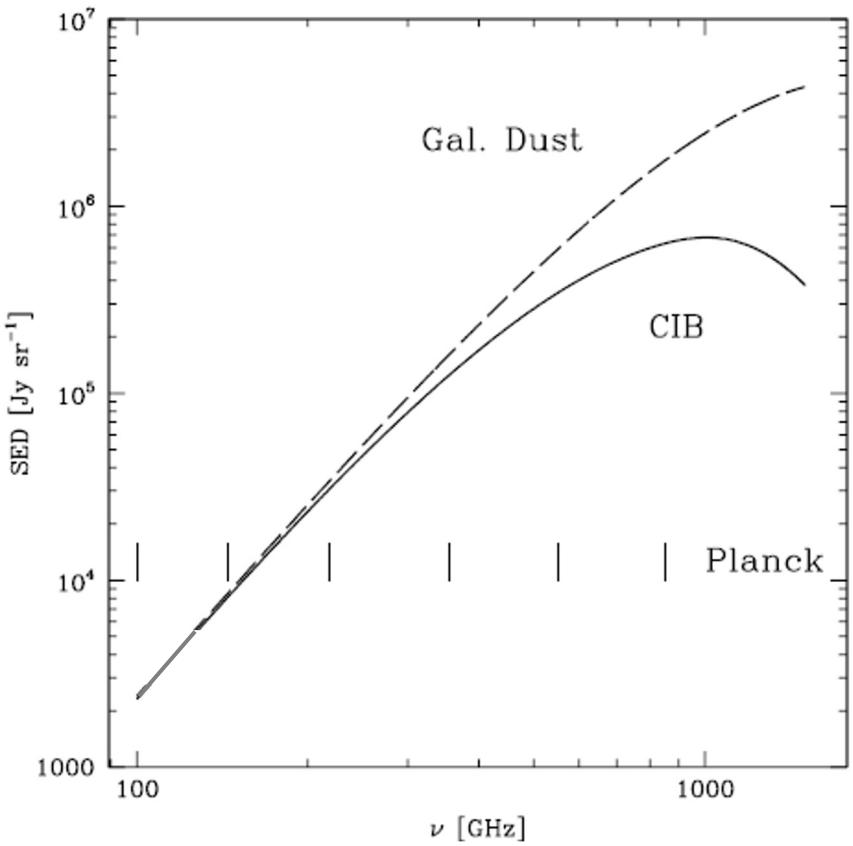
*CIB and thermal dust have similar spectral signatures (modified blackbody)*



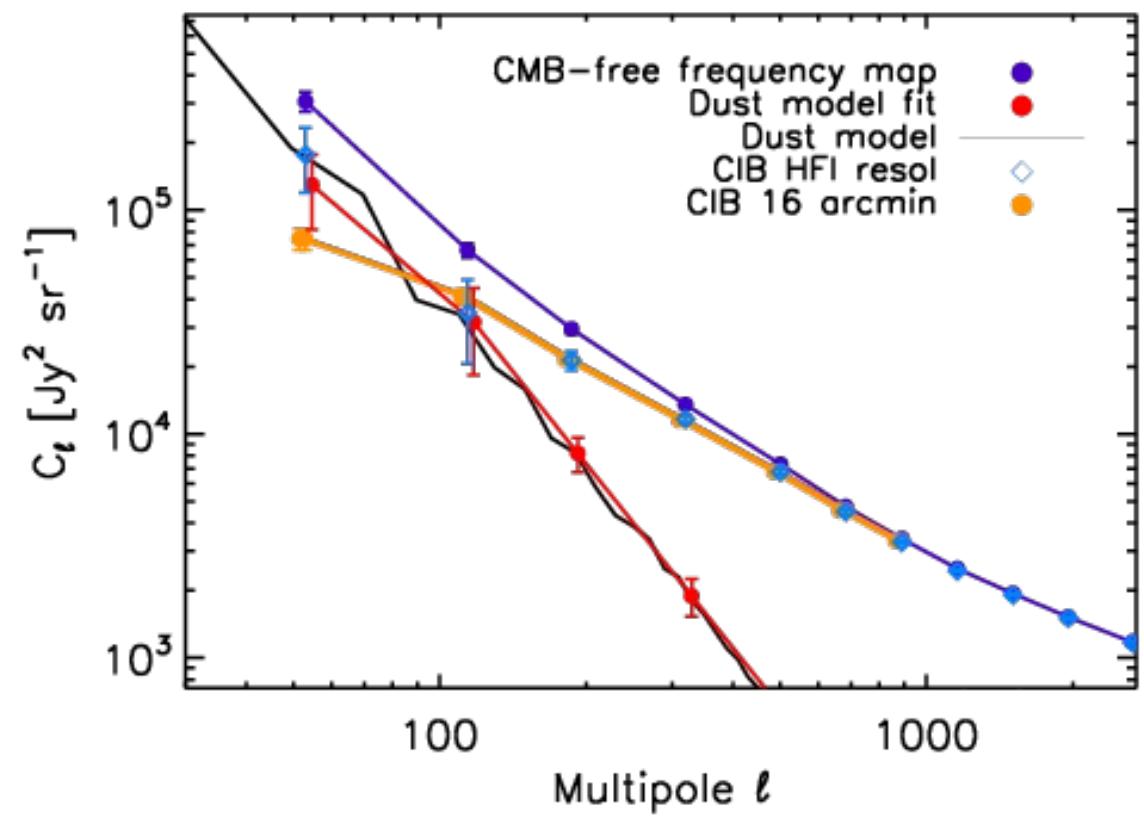
- Fitting a modified blackbody to *Planck* multi-frequency data cannot help disentangling thermal dust and CIB emissions
- Need to think beyond spectral modelling for component separation
- Instead of relying solely on spectral information, we can use statistical information to discriminate between dust and CIB

# Breaking the dust-CIB spectral degeneracy

*Thermal dust and CIB have similar spectral signatures (modified blackbody)*



*Thermal dust and CIB have different angular power spectra!*

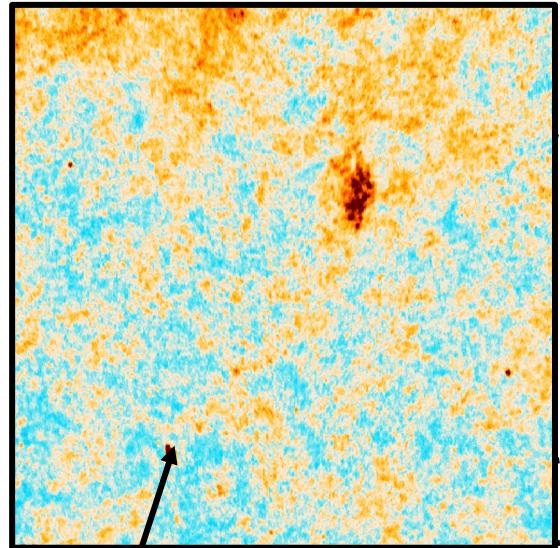


# GNILC

*Remazeilles, Delabrouille, Cardoso, MNRAS 2011b*

- ❑ **Use statistical / spatial information** (power spectrum) to break spectral degeneracies
- ❑ **Blind**, i.e. no assumption about astrophysical foregrounds
  - sole prior assumption: power spectrum of the cosmological signal
- ❑ **Wavelet-based**
  - Allow us to optimize component separation depending on the local variations of the foregrounds across the sky and across angular scales

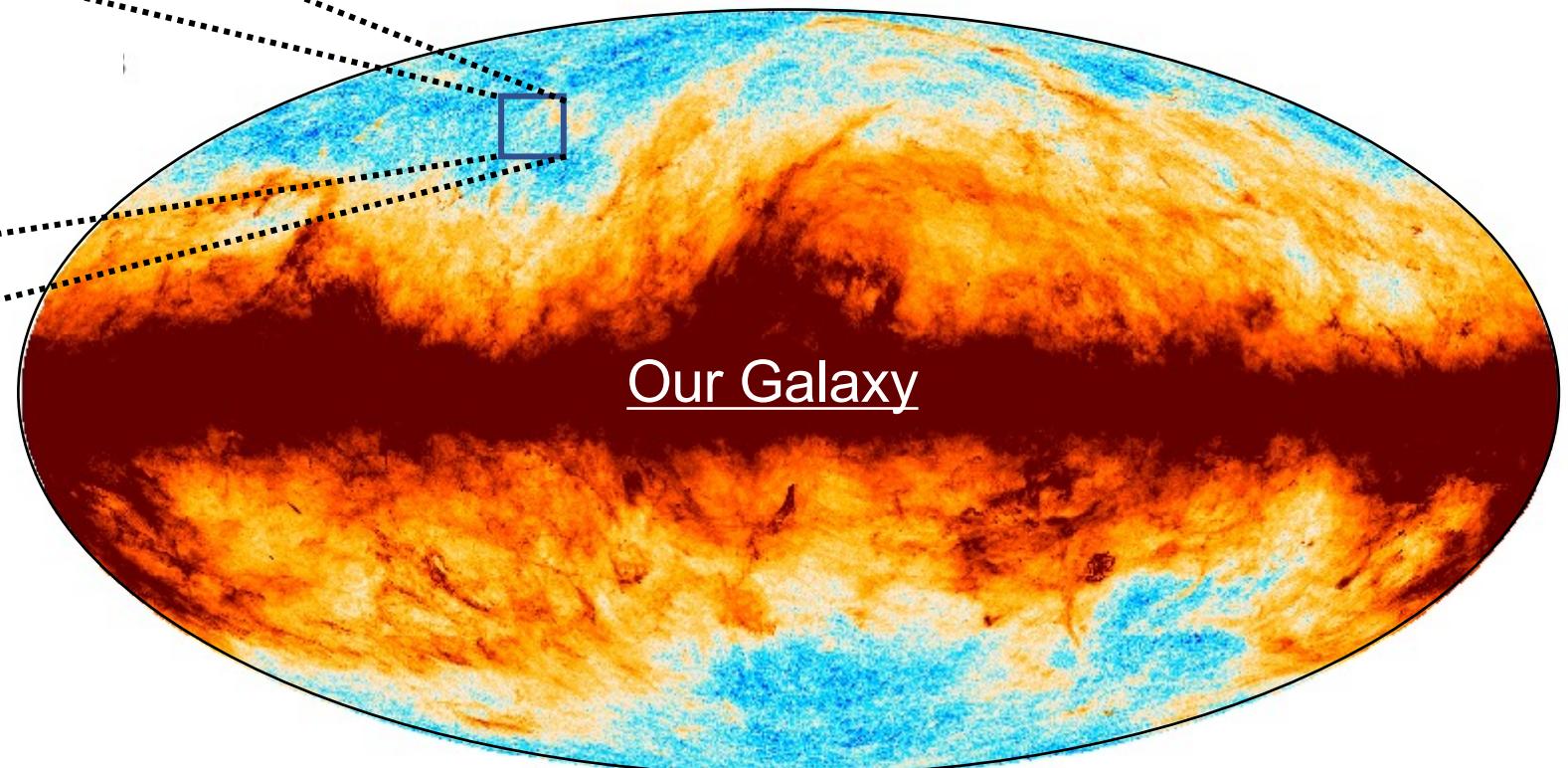
# Planck 2013 map of Galactic dust



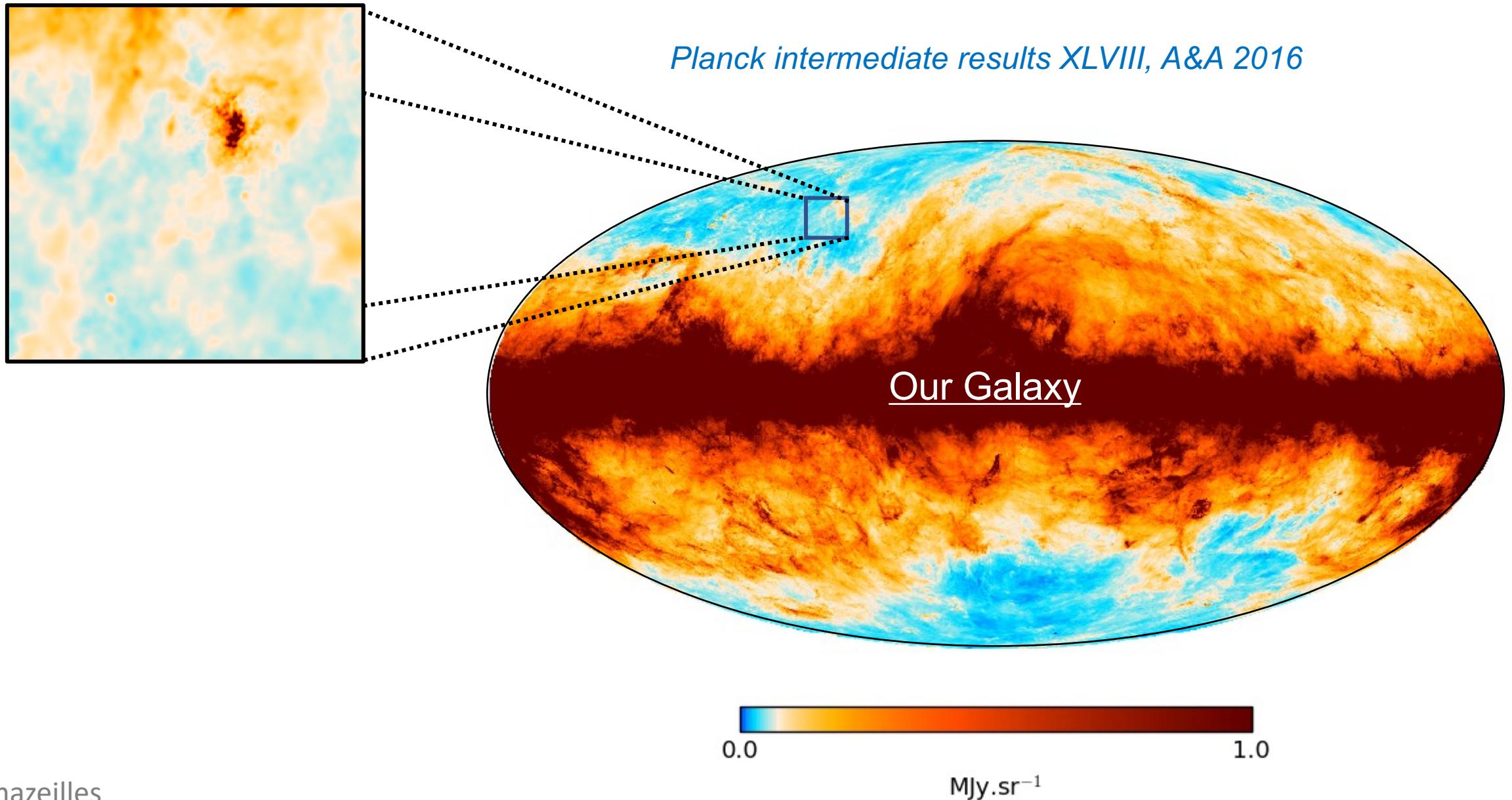
CIB contamination  
at small scales!

(background galaxies)

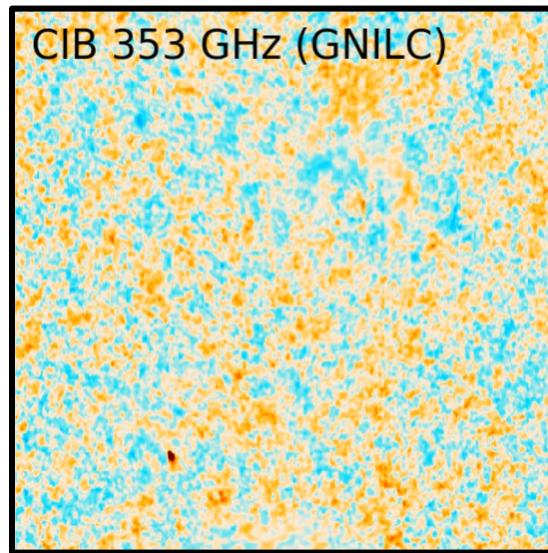
*Planck 2013 results XI, A&A 2014*



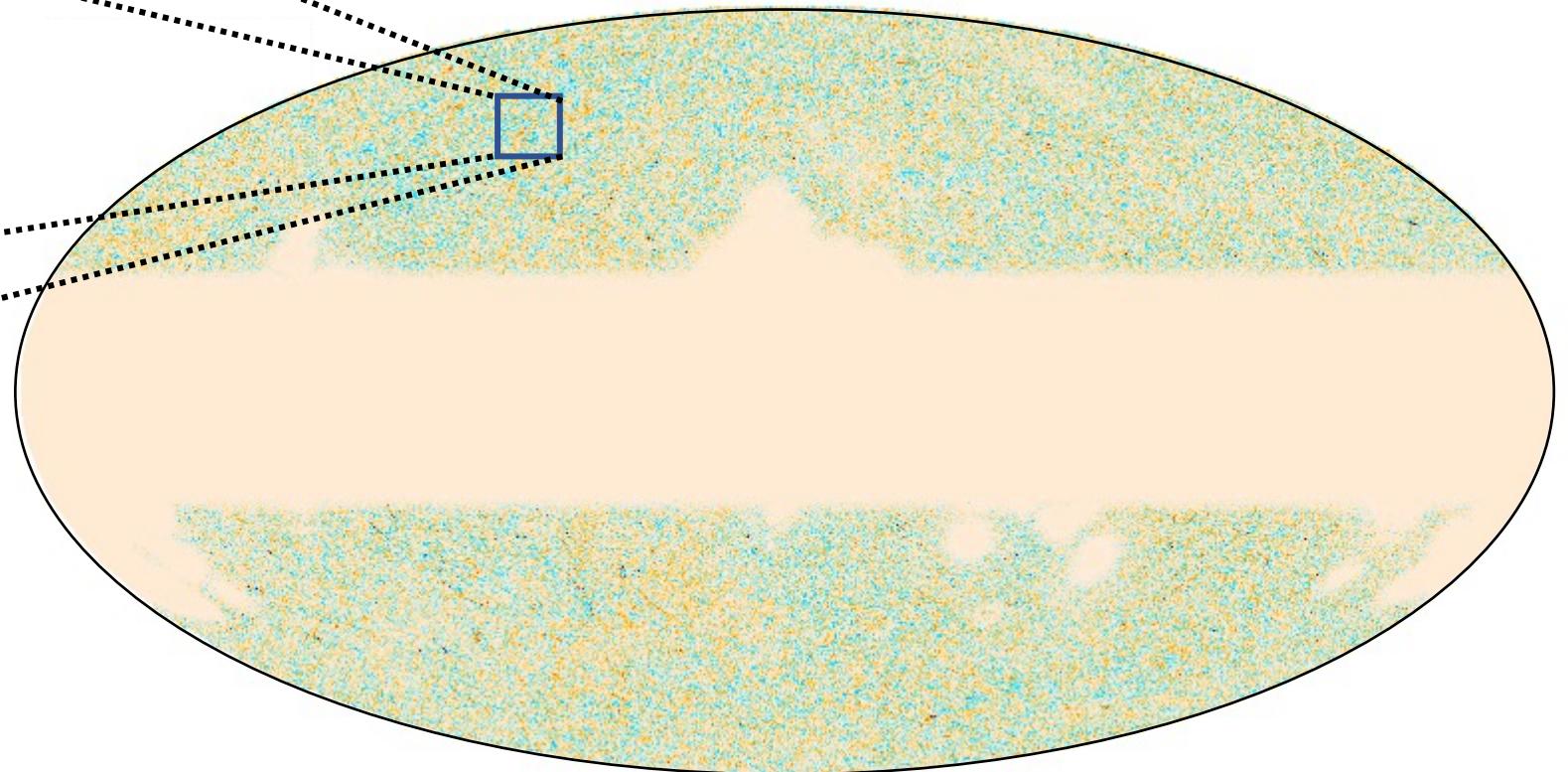
# Planck GNILC map of Galactic dust



# Planck GNILC map of CIB fluctuations

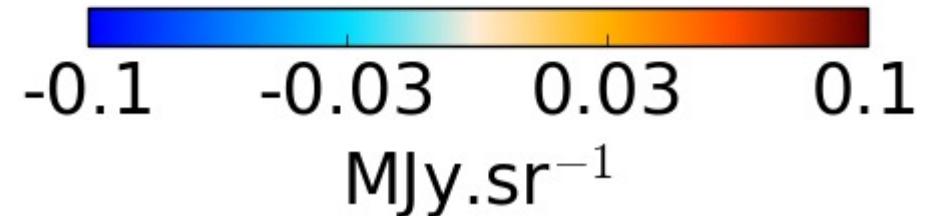


*Planck intermediate results XLVIII, A&A 2016*

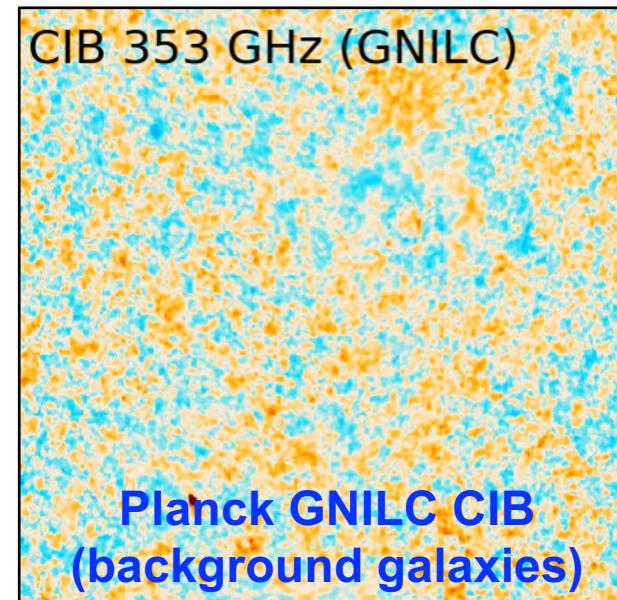
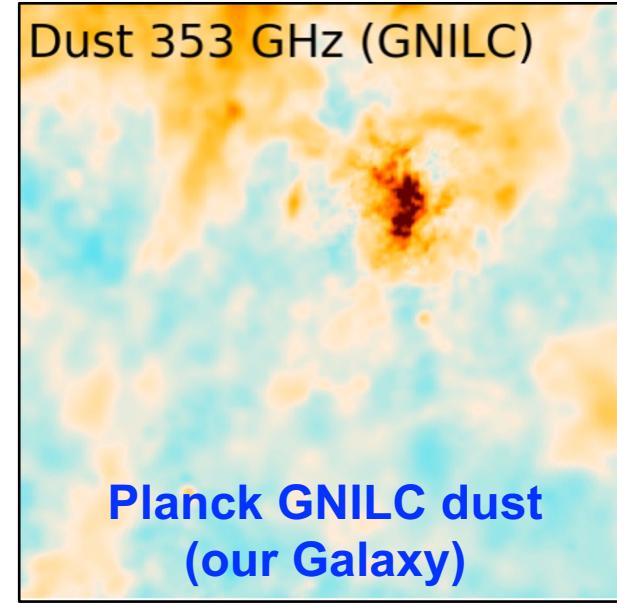
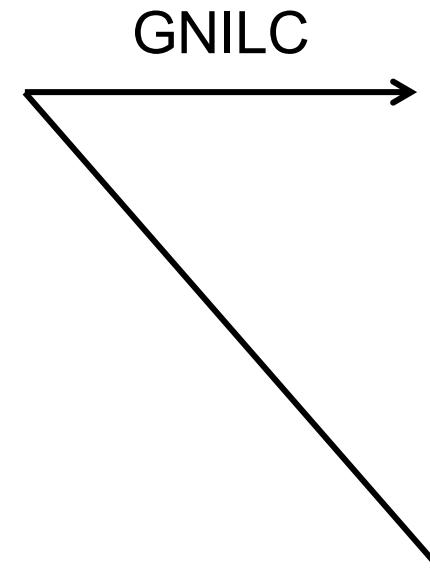
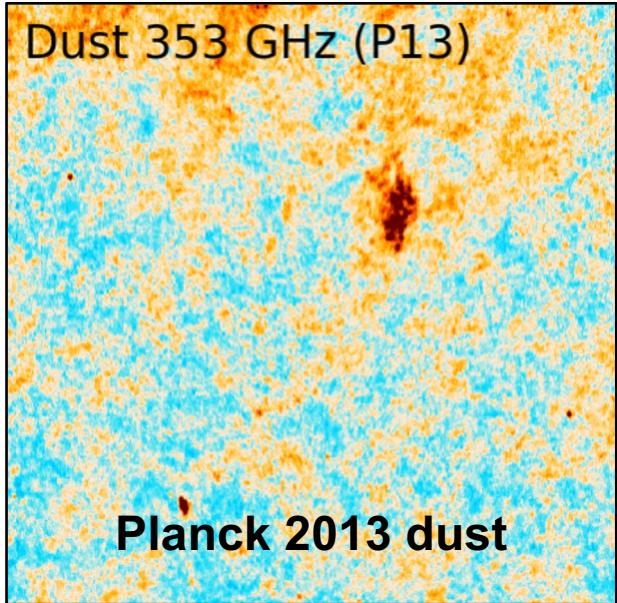


*Best template to date  
for CMB delensing*

(see A. Challinor's talk)



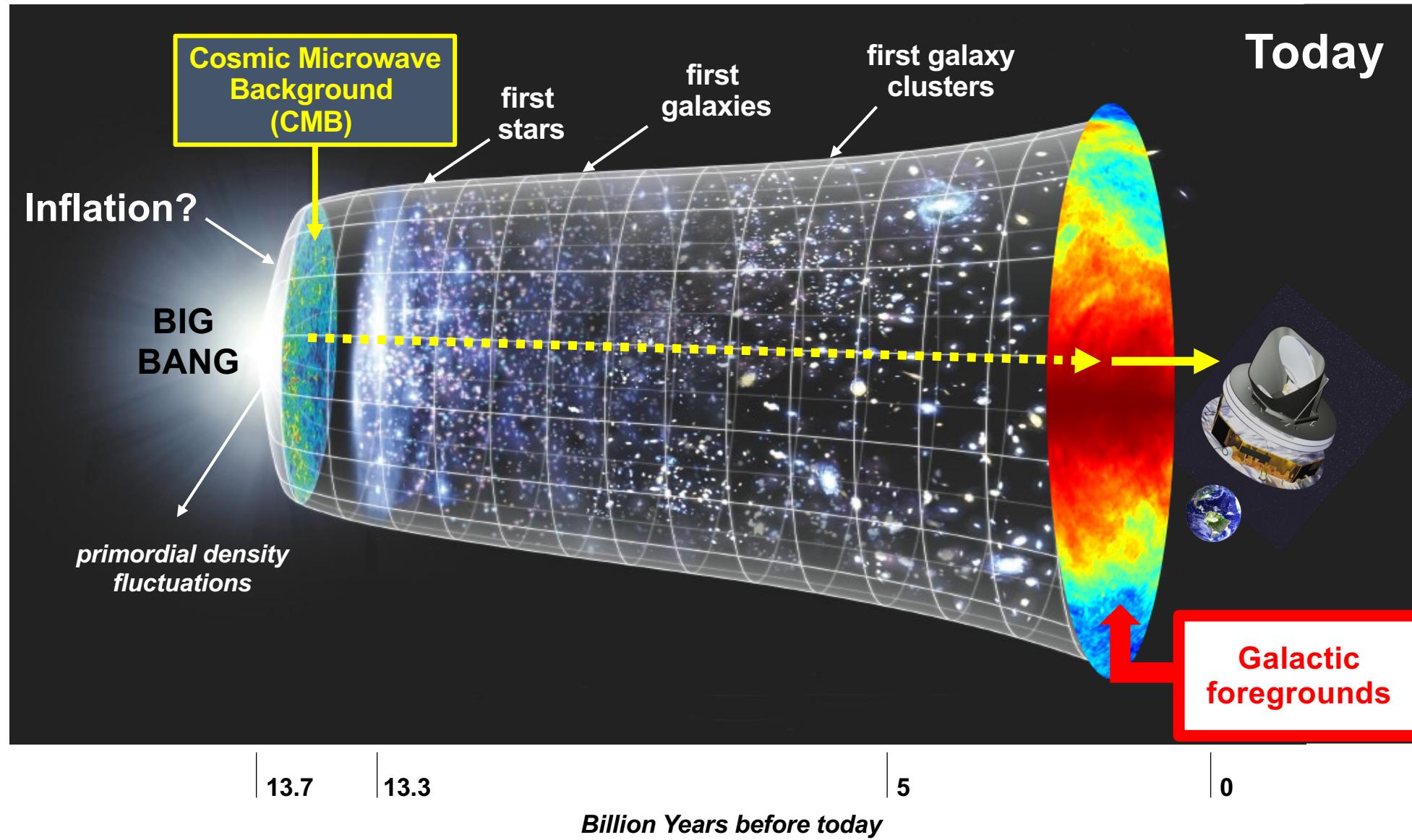
# GNILC disentangles Galactic dust and CIB



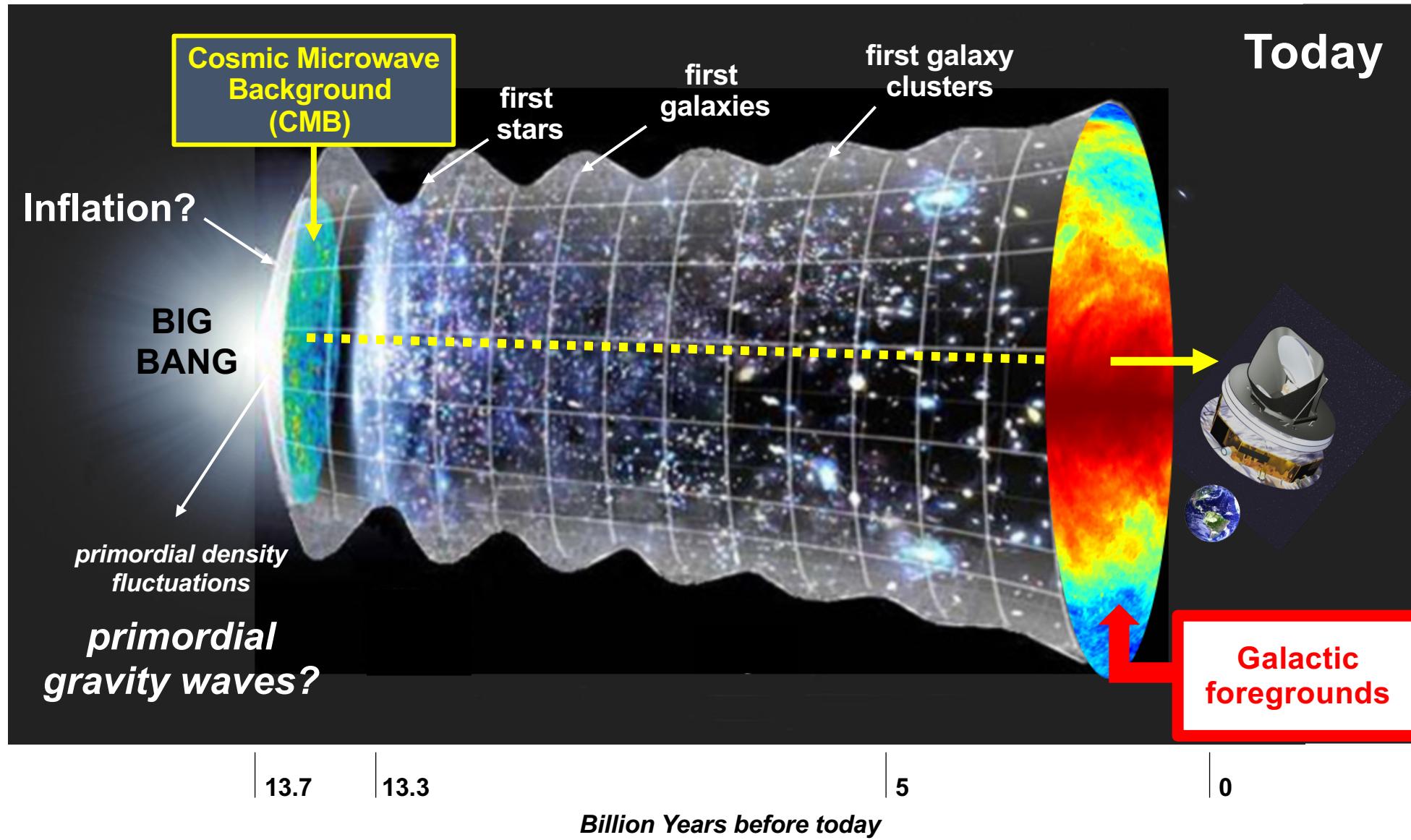
*Planck intermediate results XLVIII, A&A (2016)*

*Remazeilles, Delabrouille, Cardoso, MNRAS (2011b)*

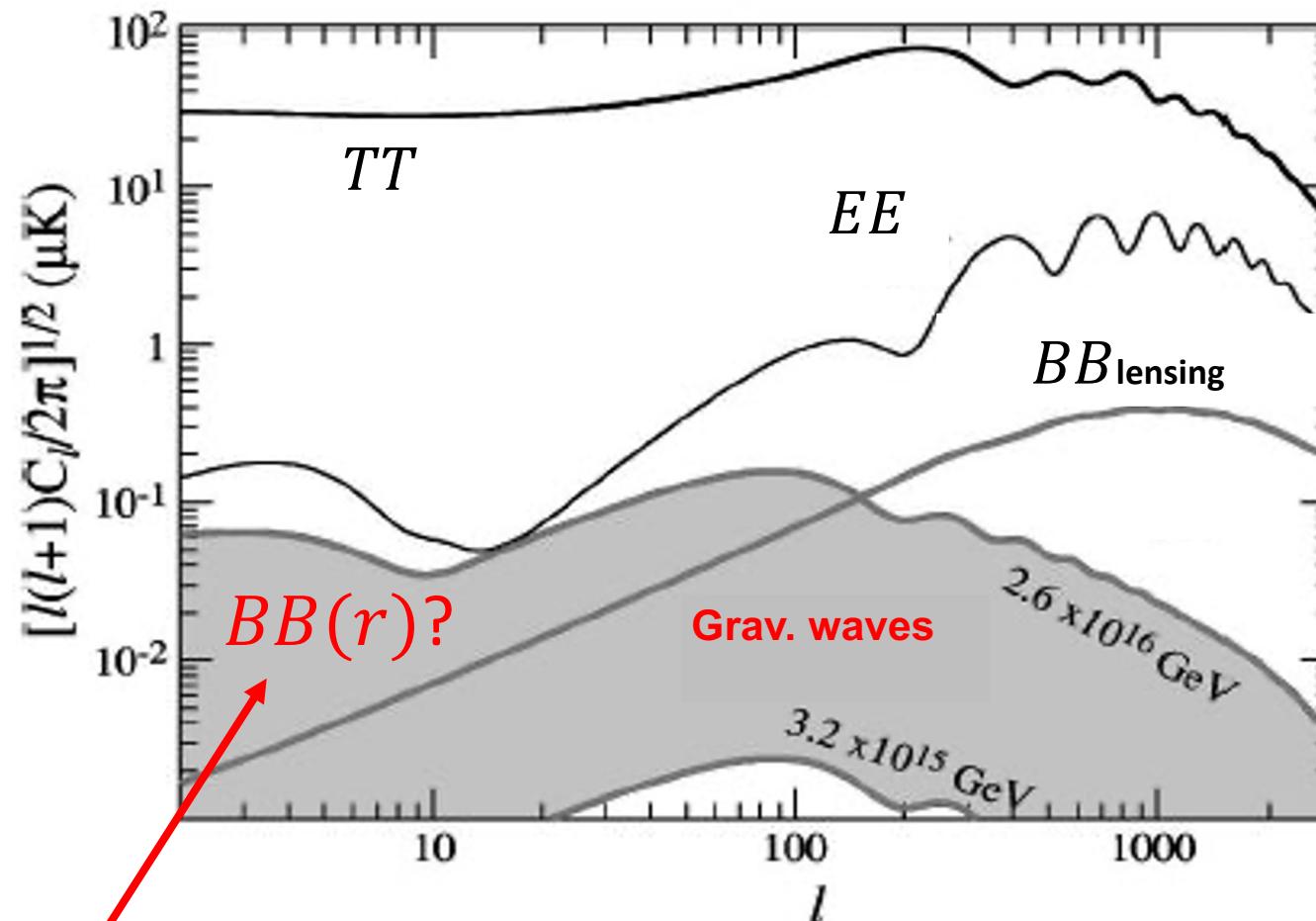
# Search for primordial B-modes



# Search for primordial B-modes

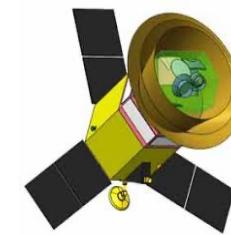


# Future CMB experiments aim at detecting $r \lesssim 0.001$

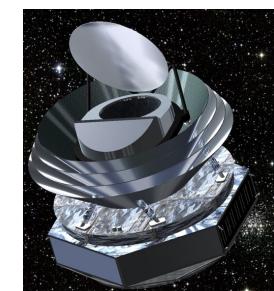


footprint of  
primordial gravitational waves  
predicted by inflation

$$E_{inf} \simeq (r/0.01)^{1/4} \times 10^{16} \text{ GeV}$$



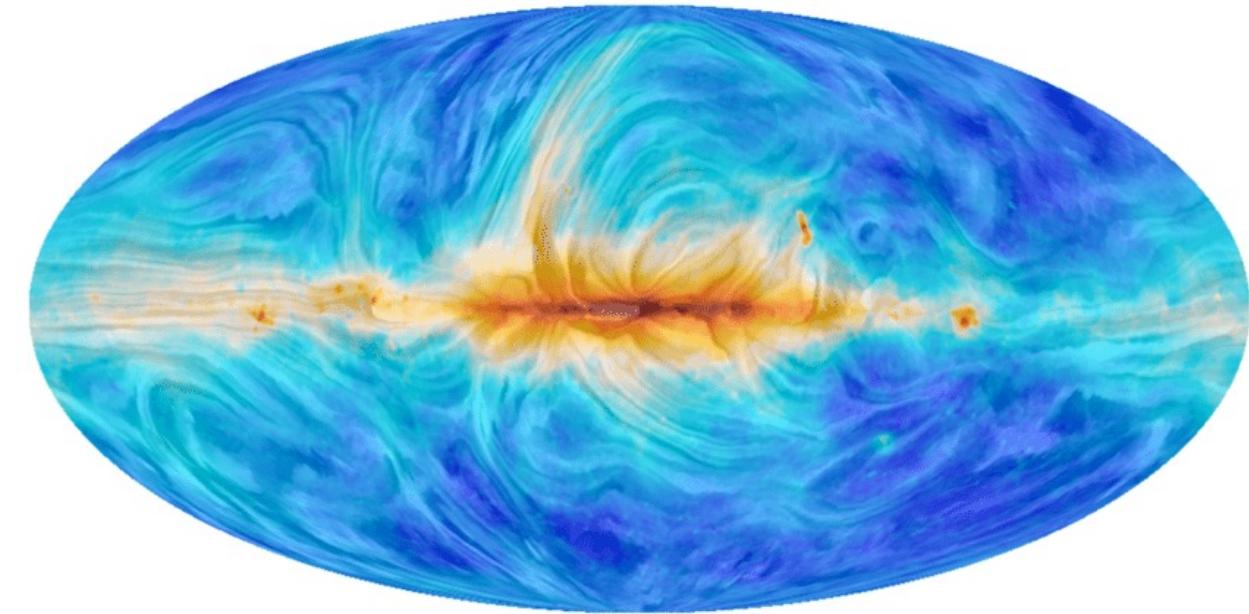
**TT**



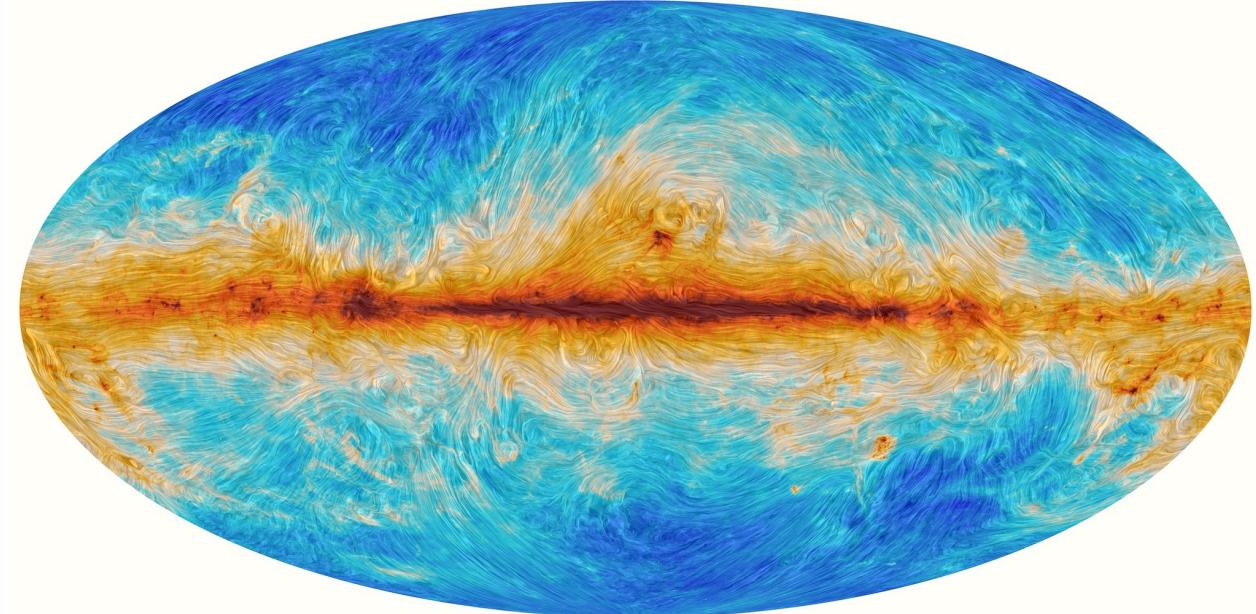
**PICO**

*Poor knowledge of the foregrounds  
at sensitivity levels of  $r \sim 0.001$*

# A variety of plausible foreground skies compatible with current polarization data



Synchrotron

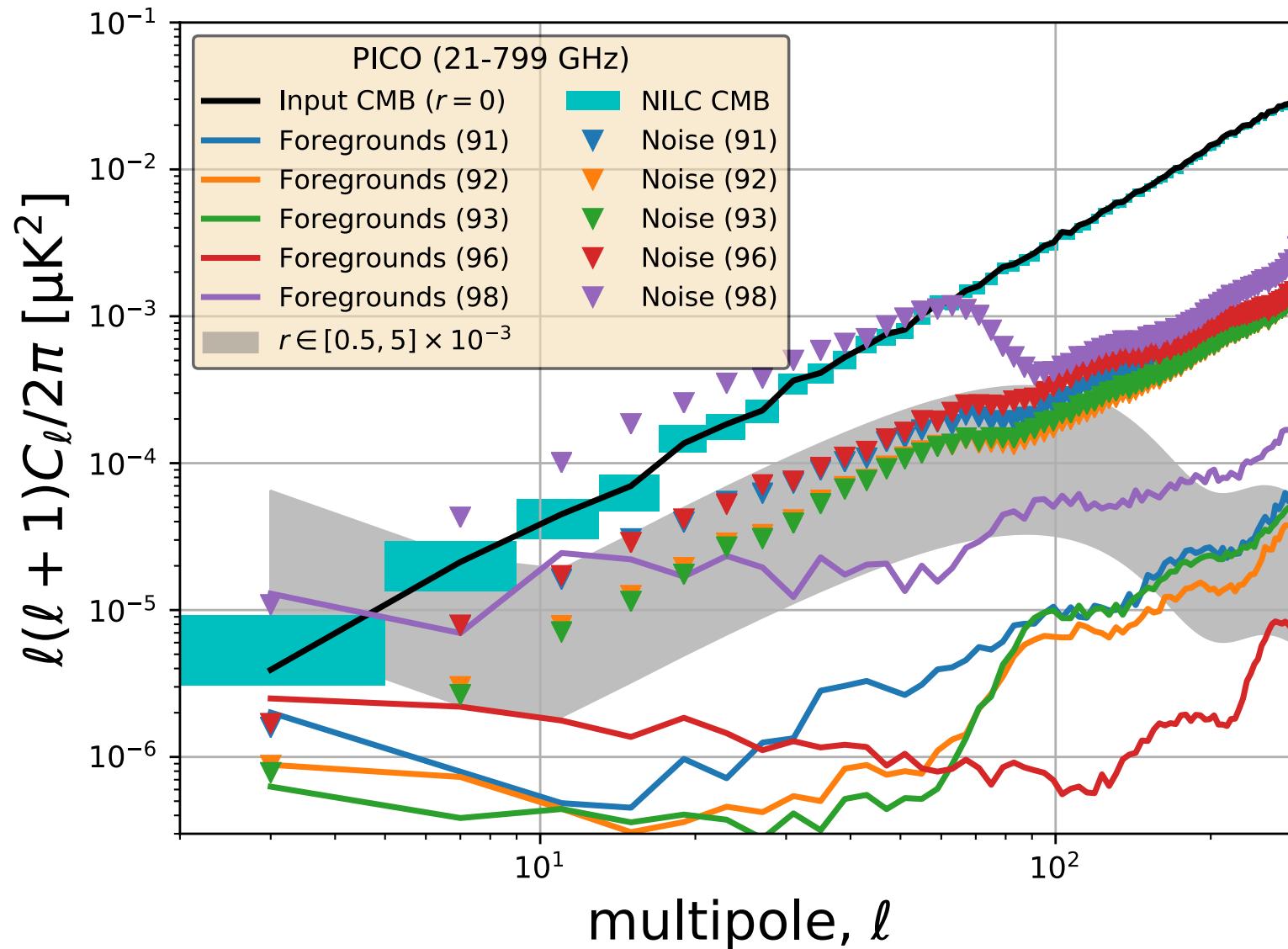


Thermal dust

# A variety of plausible foreground skies

- ❖ **Model 91 (d1s1):** *Planck* dust MBB with  $\beta, T$  variations, synchrotron power-law with  $\beta$  variations  
*Planck Collaboration X (2016)*
- ❖ **Model 92 (d4s3a2):** Two dust MBBs with  $T_1, T_2$  variations, synchrotron curvature, AME 2% polarization  
*Meissner & Finkbeiner (2015)*
- ❖ **Model 93 (d7s3a2):** Physical dust model (not MBB), synchrotron curvature, AME 2% polarization  
*Hensley (2015)*
- ❖ **Model 96 (MHD):** dust and synchrotron derived from MHD, multiple MBBs along the line-of-sight  
*Kim et al (2019)*
- ❖ **Model 98 (MKD):** Multi-layer 3D dust model (decorrelation), MBB layers along the line-of-sight  
*Martínez-Solaeche et al (2018)*

# Some foreground skies are more challenging



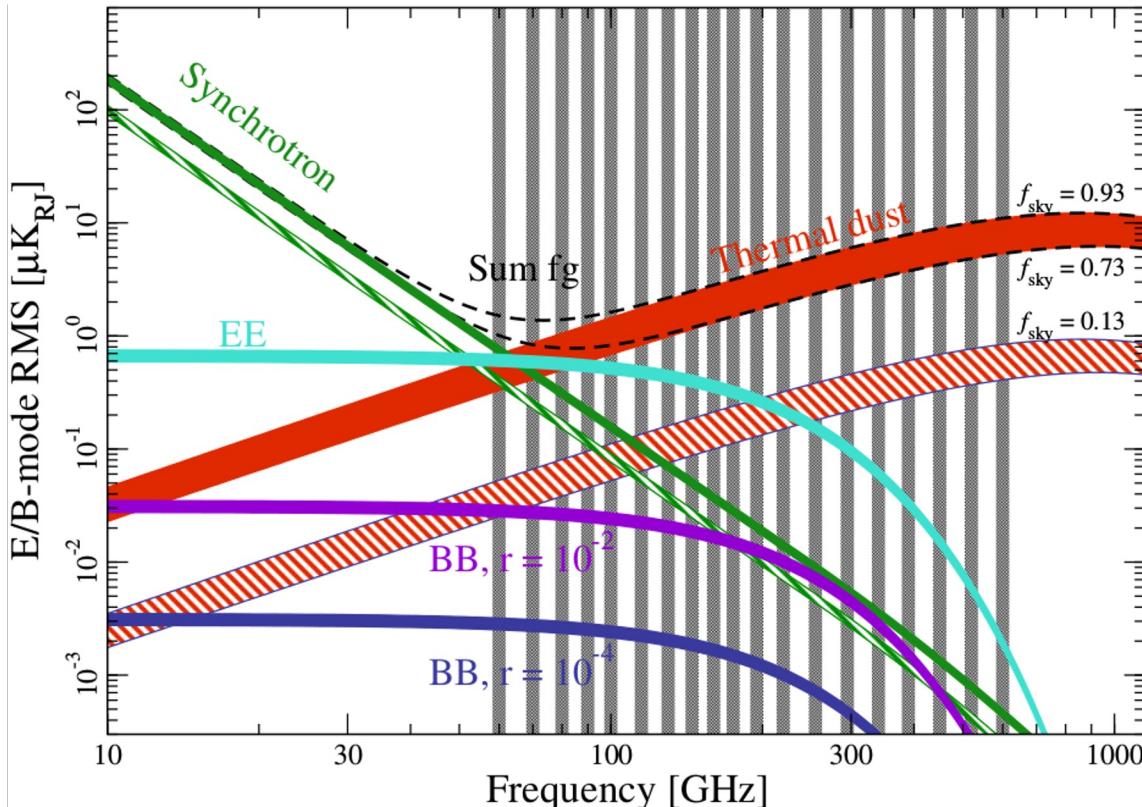
Aurlien, Belkner, Carron,  
Delabrouille, Eriksen, Fuskeland,  
Galloway, Górski, Hanany,  
Hensley, Lawrence, Pryke,  
Remazeilles, and Wehus  
(in preparation)



PICO

# CMB B-mode versus foregrounds

Remazeilles et al, JCAP 2018



- **Polarization less complex than intensity (fewer foregrounds) but more challenging (weaker signal)**
  - Huge dynamic range in amplitude between CMB B-modes and foregrounds
  - Component separation much more sensitive to any slight mismodeling of the foregrounds
- **Foregrounds cannot be avoided by narrowing the frequency range of observations**
  - At ~300 GHz, synchrotron and CMB B-modes with  $r = 0.01$  have similar amplitude and color!
  - A broad frequency coverage of the instrument is essential to break degeneracies

# Impact on $r$ of foreground mismodeling

Mismatch between the foreground model and the data

Model

$$\begin{aligned} \text{Model}(\nu, \hat{n}) = & a(\nu) s_{\text{CMB}}(\hat{n}) \\ & + \nu^{\beta_s} s_{\text{sync}}(\hat{n}) \\ & + \nu^{\beta_d} B_\nu(T_d) s_{\text{dust}}(\hat{n}) \\ & + n(\nu, \hat{n}) \end{aligned}$$

Data

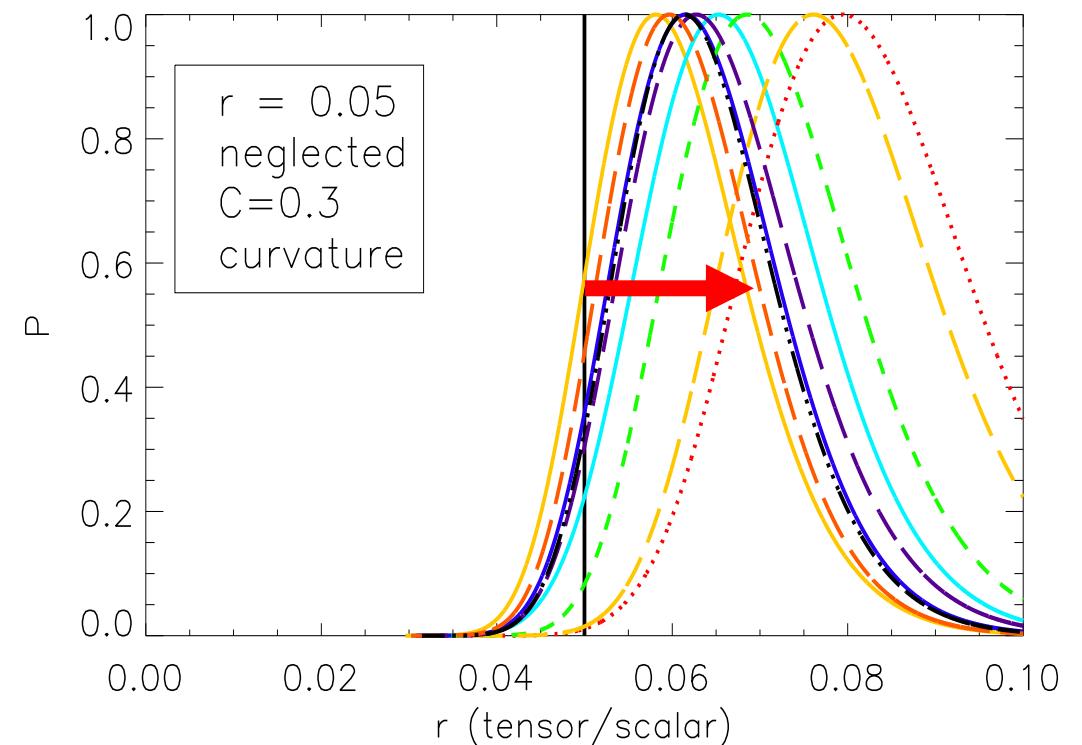
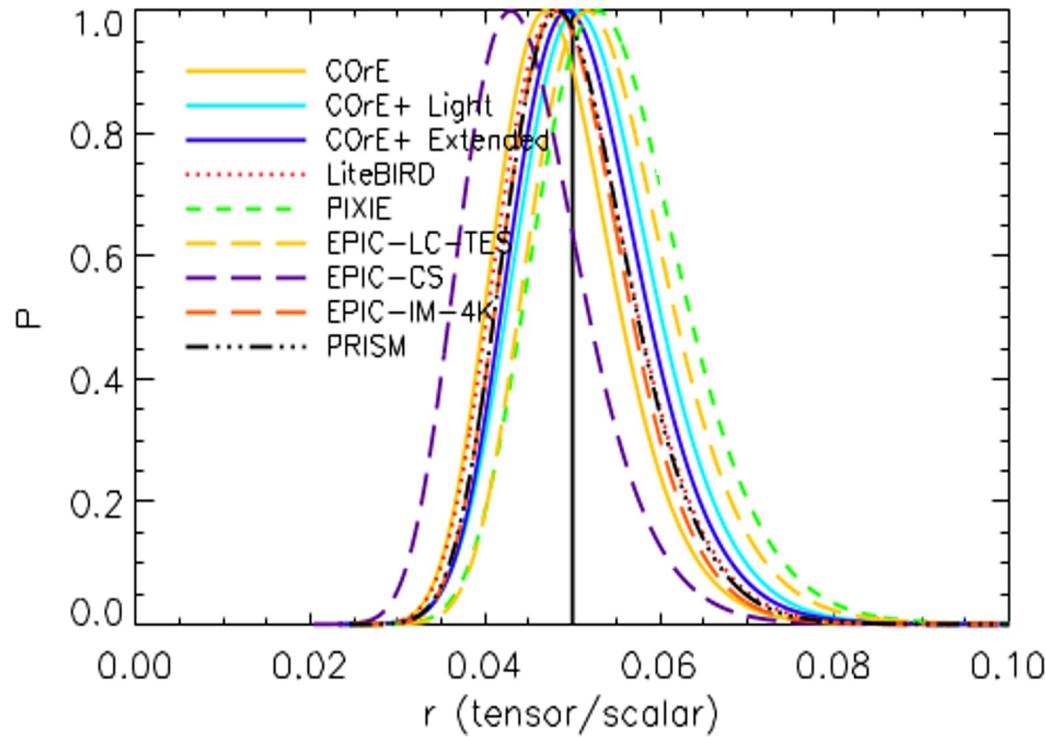
$$\begin{aligned} \text{Data}(\nu, \hat{n}) = & a(\nu) s_{\text{CMB}}(\hat{n}) \quad \text{Synchrotron curvature} \\ & + \nu^{(\beta_s + C \ln \nu)} s_{\text{sync}}(\hat{n}) \\ & + \left( f_1 \nu^{\beta_1} B_\nu(T_1) + f_2 \nu^{\beta_2} B_\nu(T_2) \right) s_{\text{dust}}(\hat{n}) \\ & + \varepsilon(\nu) s_{\text{AME}}(\hat{n}) \quad \text{Extra dust MBB} \\ & + n(\nu, \hat{n}) \quad \text{AME polarization} \end{aligned}$$

Remazeilles, Dickinson, Eriksen, Wehus, MNRAS (2016)

M. Remazeilles

# Impact on $r$ of foreground mismodeling

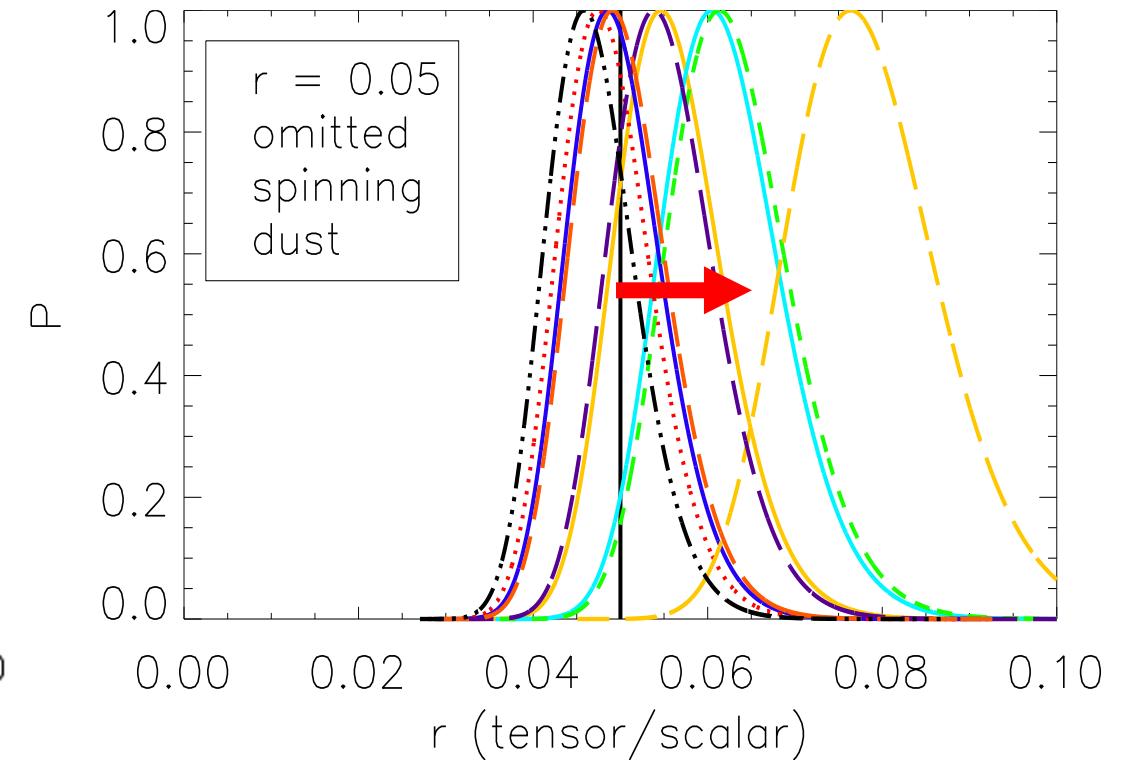
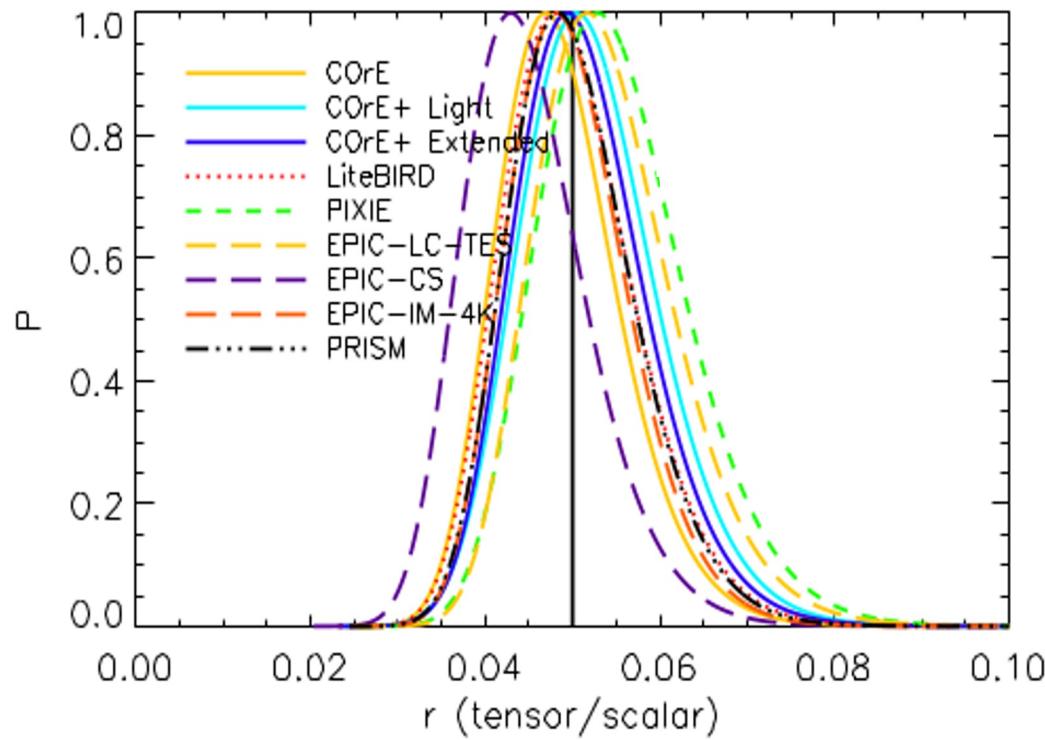
## Omitting synchrotron curvature



Remazeilles, Dickinson, Eriksen, Wehus, MNRAS (2016)

# Impact on $r$ of foreground mismodeling

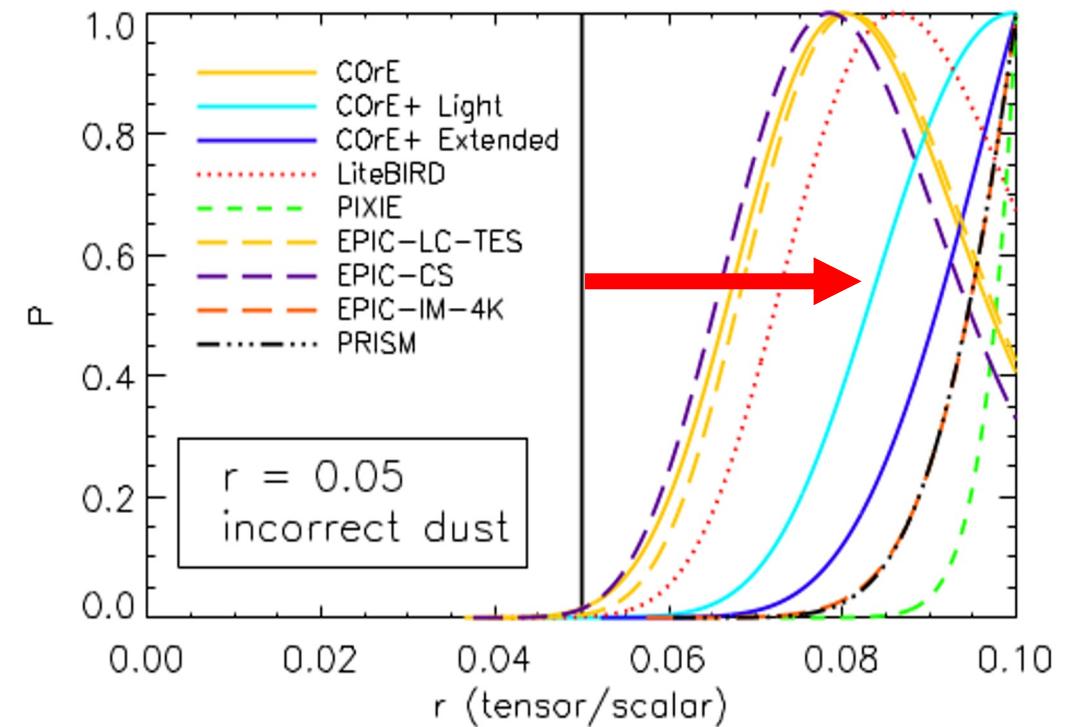
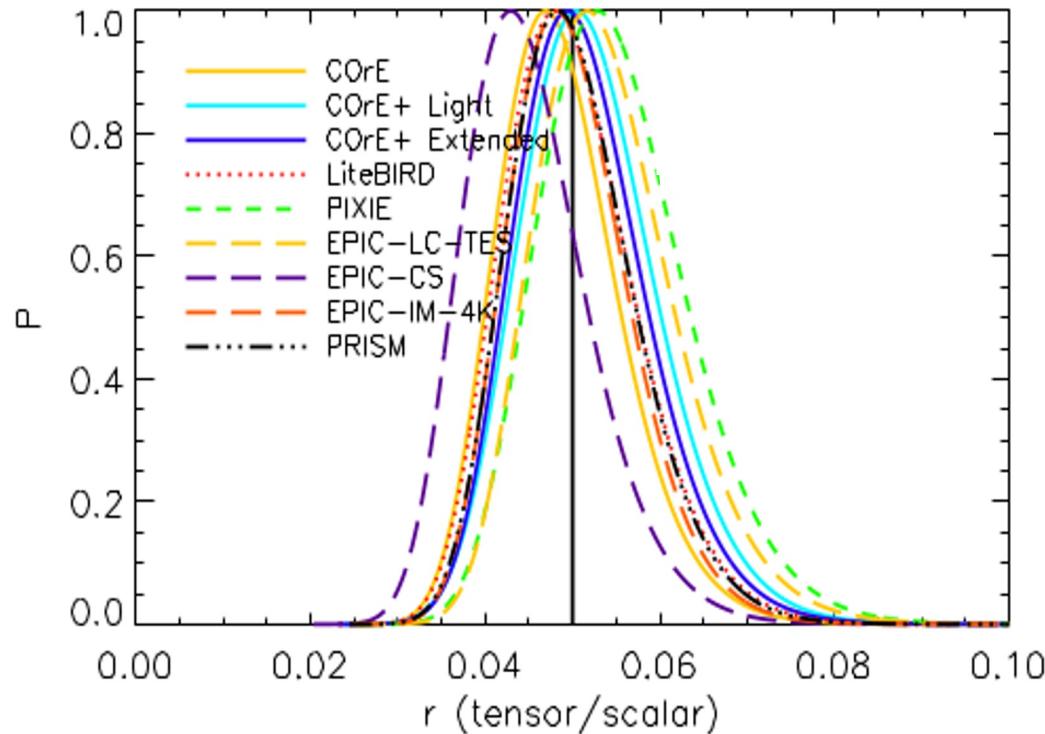
## Neglecting AME polarization



Remazeilles, Dickinson, Eriksen, Wehus, MNRAS (2016)

# Impact on $r$ of foreground mismodeling

Mismodeling two dust graybodies as a single graybody



Remazeilles, Dickinson, Eriksen, Wehus, MNRAS (2016)

# Getting evidence of foreground mismodeling?

CMB experiments with  $\nu < 300$  GHz get  $\chi^2 \simeq 1$  despite biased  $r$

correct dust model			incorrect dust model		
Mean $\chi^2/N_{\text{ch}}$	Recovered $r$	Experiment	Mean $\chi^2/N_{\text{ch}}$	Recovered $r$	Experiment
0.99	$0.05271 \pm 0.00595$	<i>COrE+ Light</i>	1.07	$0.08929 \pm 0.00766$	<i>COrE+ Light</i>
0.99	$0.05202 \pm 0.00585$	<i>COrE+ Extended</i>	1.20	$0.09218 \pm 0.00624$	<i>COrE+ Extended</i>
0.98	$0.05107 \pm 0.00575$	<i>COrE</i>	1.12	$0.08023 \pm 0.01045$	<i>COrE</i>
0.96	$0.05132 \pm 0.00578$	<i>LiteBIRD</i>	1.10	$0.08428 \pm 0.00935$	<i>LiteBIRD</i>
0.99	$0.05145 \pm 0.00616$	<i>PIXIE</i>	1.08	$0.09711 \pm 0.00265$	<i>PIXIE</i>
0.97	$0.05074 \pm 0.00572$	<i>EPIC-LC-TES</i>	1.32	$0.08113 \pm 0.00999$	<i>EPIC-LC-TES</i>
0.96	$0.05086 \pm 0.00583$	<i>EPIC-CS</i>	1.08	$0.07911 \pm 0.01048$	<i>EPIC-CS</i>
0.97	$0.05096 \pm 0.00572$	<i>EPIC-IM-4K</i>	2.88	$0.09434 \pm 0.00485$	<i>EPIC-IM-4K</i>
0.99	$0.05140 \pm 0.00578$	<i>PRISM</i>	1.58	$0.09446 \pm 0.00467$	<i>PRISM</i>

$$\chi^2(\hat{n}) = \sum_{\nu} \frac{(\text{Data}(\nu, \hat{n}) - \text{Model}(\nu, \hat{n}))^2}{\sigma_{\text{noise}}^2(\nu, \hat{n})}$$

*Remazeilles, Dickinson, Eriksen, Wehus, MNRAS (2016)*

# Getting evidence of foreground mismodeling?

CMB experiments with  $\nu < 300$  GHz get  $\chi^2 \simeq 1$  despite biased  $r$

## How to fool yourself!

*Good fit of total emission, not of individual components*

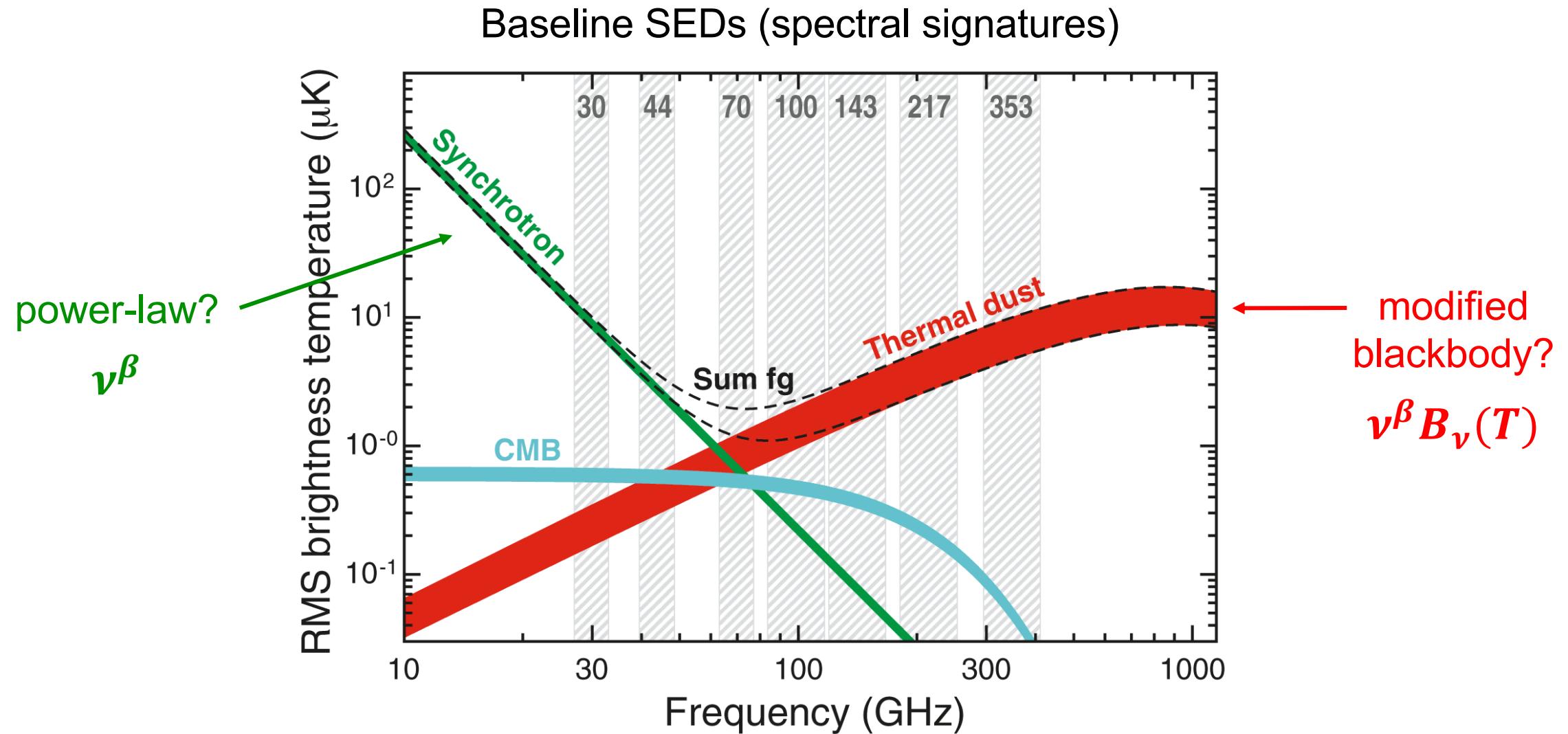
*Model degeneracies over narrow frequency ranges*

*Without high frequencies, no chi-square evidence for incorrect foreground modelling and biased detection of  $r$*

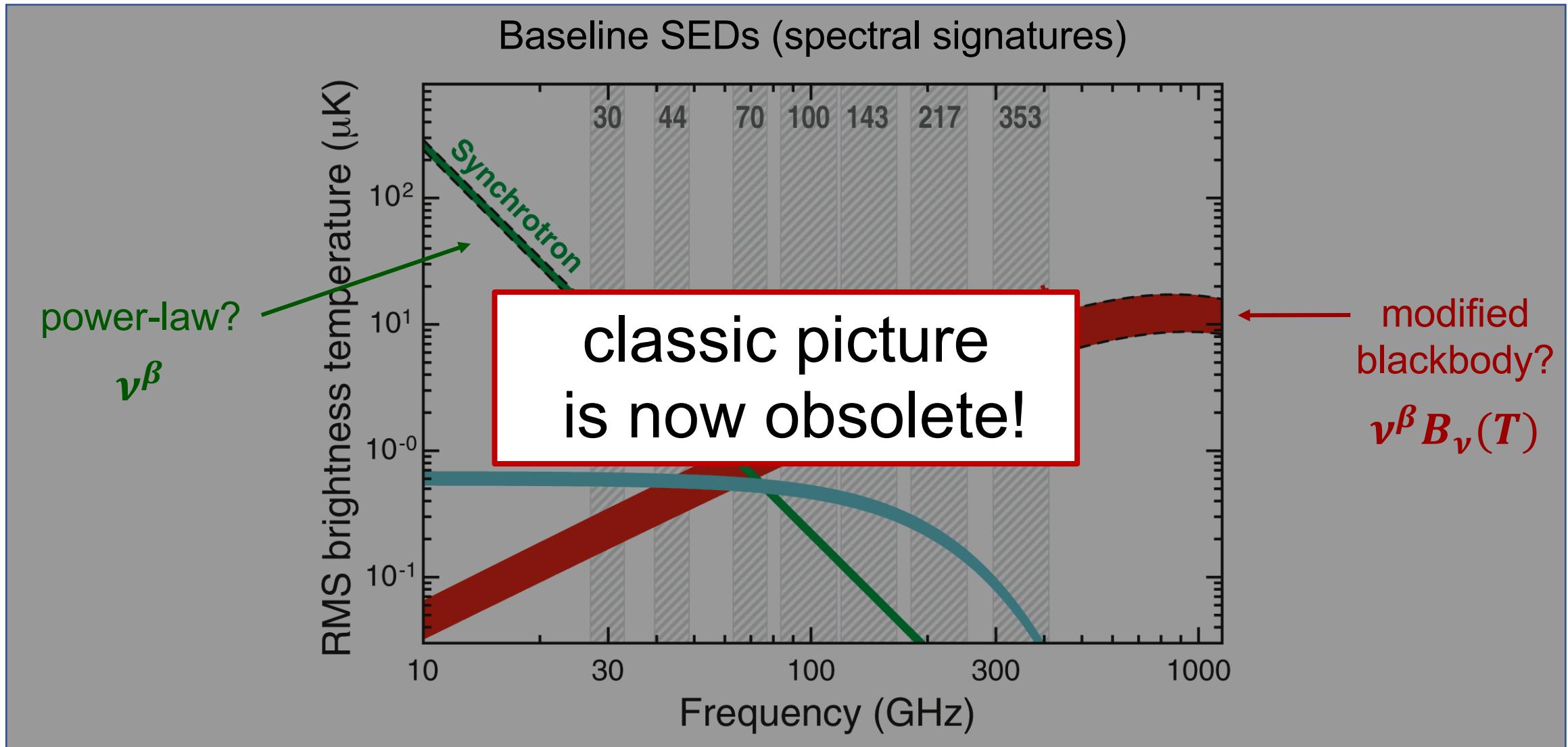
*Remazeilles, Dickinson, Eriksen, Wehus, MNRAS (2016)*

*Spectral distortions  
of the foregrounds*

# Rethinking current foregrounds parameterizations

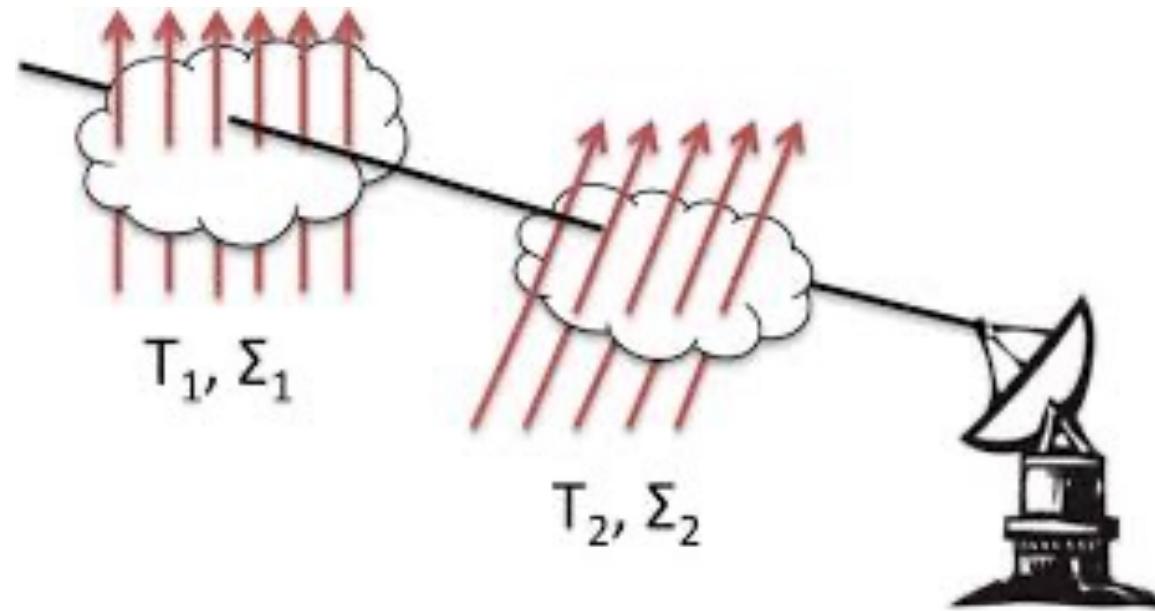


# Rethinking current foregrounds parameterizations



# Spectral distortions of the foreground SEDs

Line-of-sight integration of multiple contributions to thermal dust emission



*Tassis et al, MNRAS (2015)*  
*Pelgrims et al, A&A (2021)*

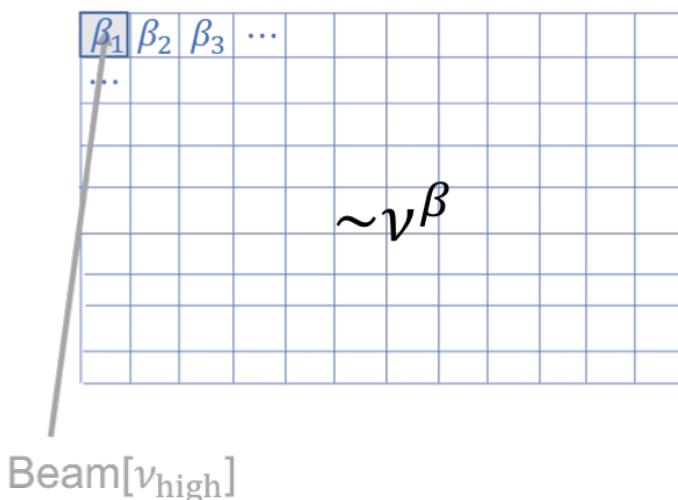
$$\nu^{\beta_1} B_\nu(T_1) + \nu^{\beta_2} B_\nu(T_2) \neq \nu^\beta B_\nu(T)$$

in each single pixel

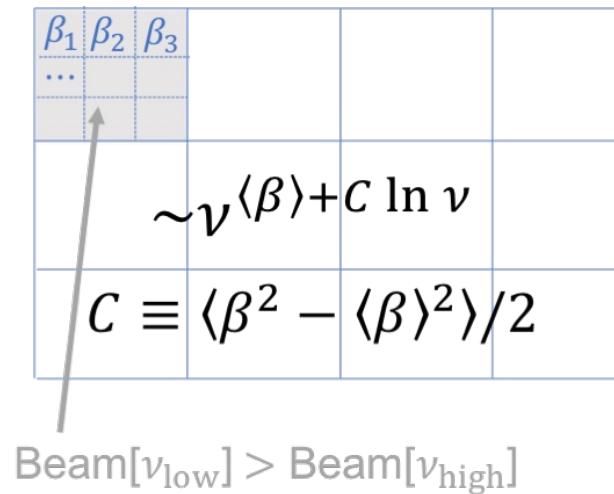
# Spectral distortions of the foreground SEDs

Foreground emission integrated over larger beams at low frequencies

*High-resolution high-frequency map*



*Low-resolution low-frequency map*



*Remazeilles et al  
MNRAS (2021)*

## Spectral distortion of the baseline foreground SED

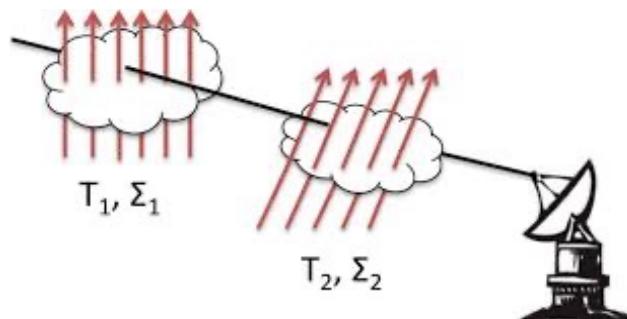
“power-law” at high frequency  $\leftrightarrow$  “curved power-law” at low frequency

effective curvature = variance (2<sup>nd</sup> moment) of the spectral index in the beam

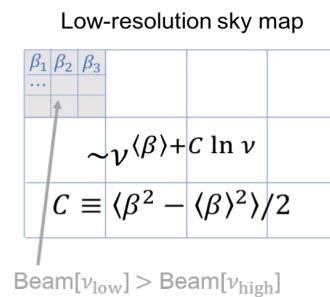
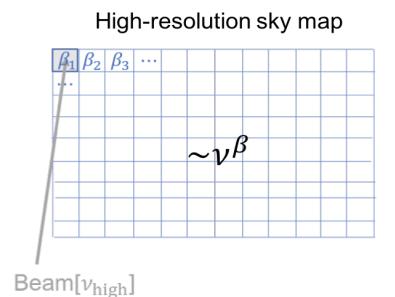
# Spectral distortions of the foreground SEDs

## Foreground SED distortions

### Averaging along the line of sight



### Averaging between lines of sight in the beam



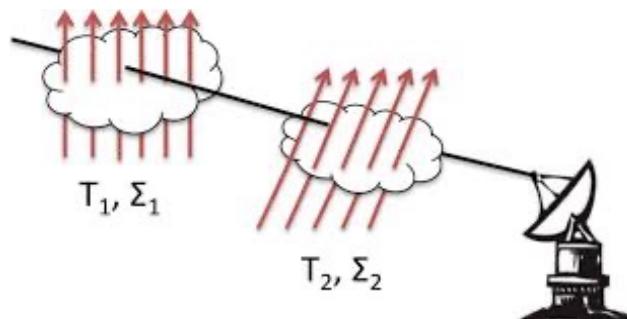
- Spectral distortions of the baseline foreground SEDs
- Decorrelation across frequencies
- Augmented list of expected foreground parameters

*Chluba et al MNRAS (2017)  
Remazeilles et al MNRAS (2021)  
Vacher et al A&A (2022)*

# Spectral distortions of the foreground SEDs

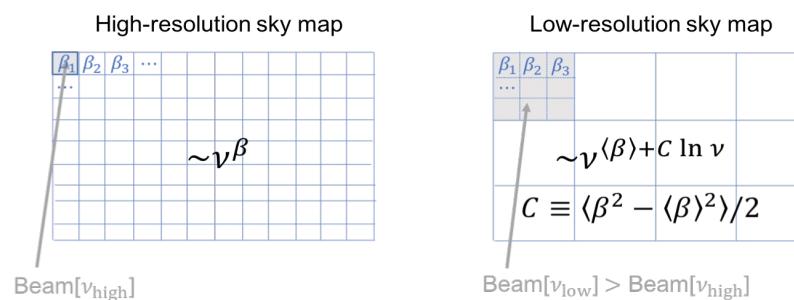
## Foreground SED distortions

### Averaging along the line of sight



$$\nu^{\beta_1} B_\nu(T_1) + \nu^{\beta_2} B_\nu(T_2) \neq \nu^\beta B_\nu(T)$$

### Averaging between lines of sight in the beam



- Spectral distortions of the baseline foreground SEDs
- Decorrelation across frequencies
- Augmented list of expected foreground parameters

**tiny corrections to foregrounds >> CMB B-mode !**

*Chluba et al MNRAS (2017)  
Remazeilles et al MNRAS (2021)  
Vacher et al A&A (2022)*

# Extending current foreground parameterizations

Moment expansion of the foreground emission beyond leading-order SED

$$I_\nu(\hat{n}) = A(\hat{n})f(\nu, \bar{\beta}) + A(\hat{n})(\beta(\hat{n}) - \bar{\beta})\partial_\beta f(\nu, \bar{\beta}) + \frac{1}{2}A(\hat{n})(\beta(\hat{n}) - \bar{\beta})^2\partial_\beta^2 f(\nu, \bar{\beta}) + \dots$$

$$\simeq \nu^\beta \quad \simeq \frac{2h\nu^3}{c^2} \frac{\nu^\beta}{e^{\frac{h\nu}{kT}} - 1}$$

synchrotron

thermal dust

(leading order)

*Chluba, Hill, Abitbol, MNRAS (2017)*  
*Remazeilles, Rotti, Chluba, MNRAS (2021)*

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Moment expansion of the foreground emission beyond leading-order SED

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$$\begin{aligned} & \simeq \nu^\beta & \simeq \frac{2h\nu^3}{c^2} \frac{\nu^\beta}{e^{\frac{h\nu}{kT}} - 1} \\ \text{synchrotron} & & \text{thermal dust} \\ & \text{(leading order)} \end{aligned}$$

Extra components (moments)

*Chluba, Hill, Abitbol, MNRAS (2017)*  
*Remazeilles, Rotti, Chluba, MNRAS (2021)*

# Extending current foreground parameterizations

Moment expansion of the foreground emission beyond leading-order SED

$$I_\nu(\hat{n}) = A(\hat{n})f(\nu, \bar{\beta}) + \underbrace{A(\hat{n})(\beta(\hat{n}) - \bar{\beta})\partial_\beta f(\nu, \bar{\beta})}_{\text{Extra components (moments)}} + \underbrace{\frac{1}{2}A(\hat{n})(\beta(\hat{n}) - \bar{\beta})^2\partial_\beta^2 f(\nu, \bar{\beta})}_{\text{New SEDs}} + \dots$$

$$\begin{aligned} & \simeq \nu^\beta & \simeq \frac{2h\nu^3}{c^2} \frac{\nu^\beta}{e^{\frac{h\nu}{kT}} - 1} \\ \text{synchrotron} & & \text{thermal dust} \\ & \text{(leading order)} \end{aligned}$$

Extra components (moments)

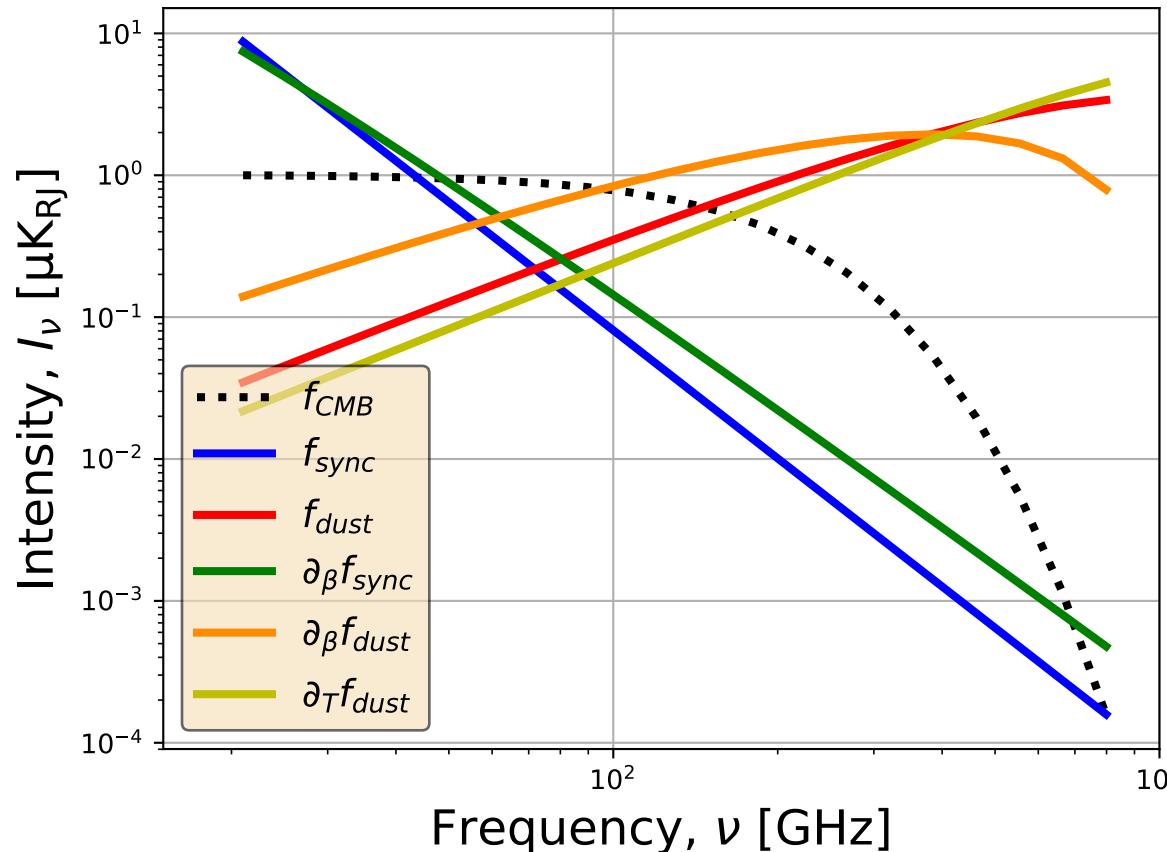
New SEDs

*Chluba, Hill, Abitbol, MNRAS (2017)*  
*Remazeilles, Rotti, Chluba, MNRAS (2021)*

# Extending current foreground parameterizations

Moment expansion of the foreground emission beyond leading-order SED

$$I_\nu(\hat{n}) = A(\hat{n})f(\nu, \bar{\beta}) + \underbrace{A(\hat{n})(\beta(\hat{n}) - \bar{\beta})\partial_\beta f(\nu, \bar{\beta})}_{\text{red}} + \underbrace{\frac{1}{2}A(\hat{n})(\beta(\hat{n}) - \bar{\beta})^2\partial_\beta^2 f(\nu, \bar{\beta})}_{\text{blue}} + \dots$$



Extra components (moments)

New SEDs

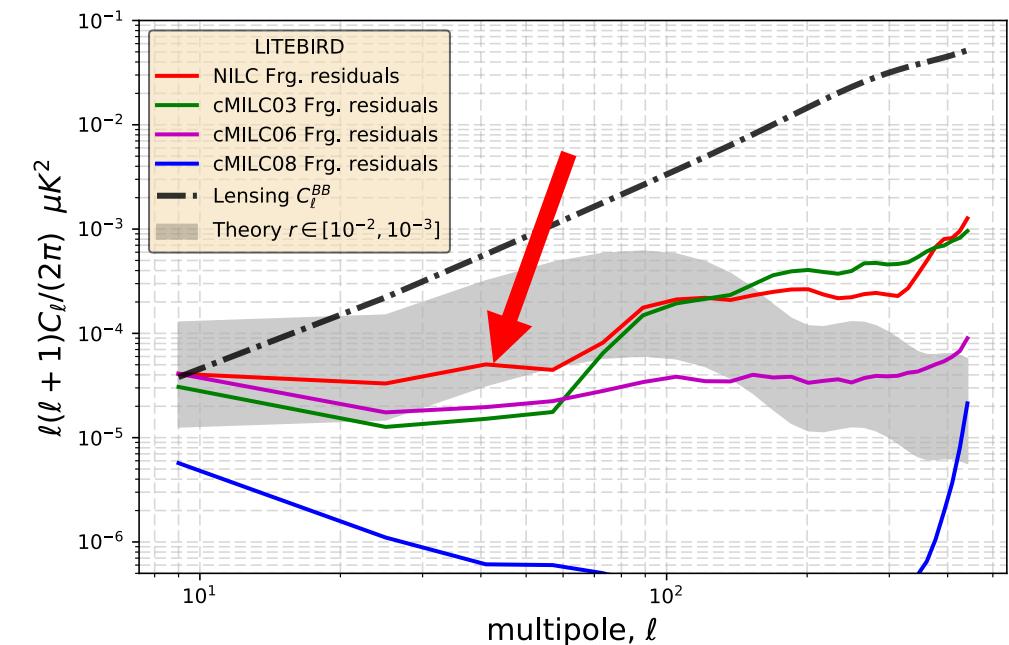
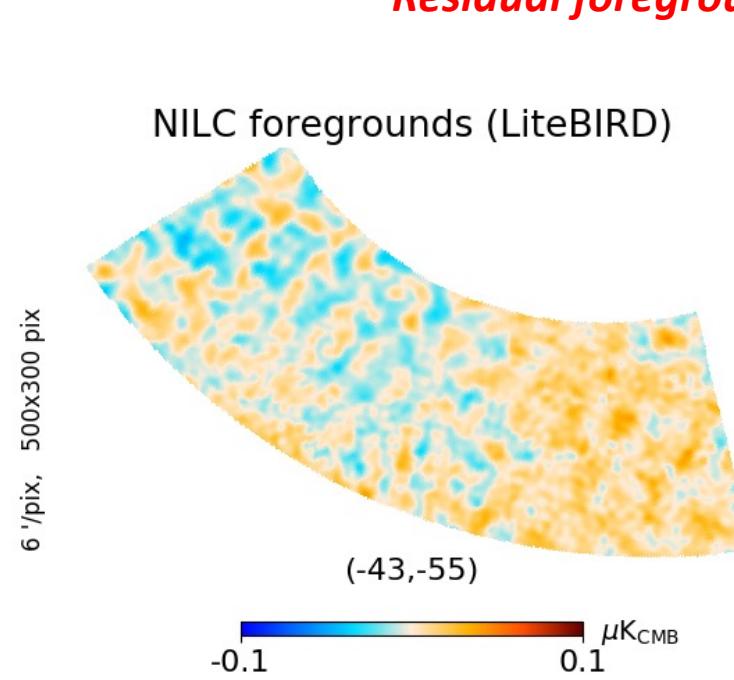
*Chluba, Hill, Abitbol, MNRAS (2017)*  
*Remazeilles, Rotti, Chluba, MNRAS (2021)*

# Deprojecting the foreground moments

Can we augment the routine NILC method for B-modes?

$$\hat{s}_{\text{CMB}}(\vec{n}) = \sum_{\nu} w(\nu) \cdot d(\nu, \vec{n})$$

$$\sum_{\nu} w(\nu) \cdot f_{\text{CMB}}(\nu) = 1$$

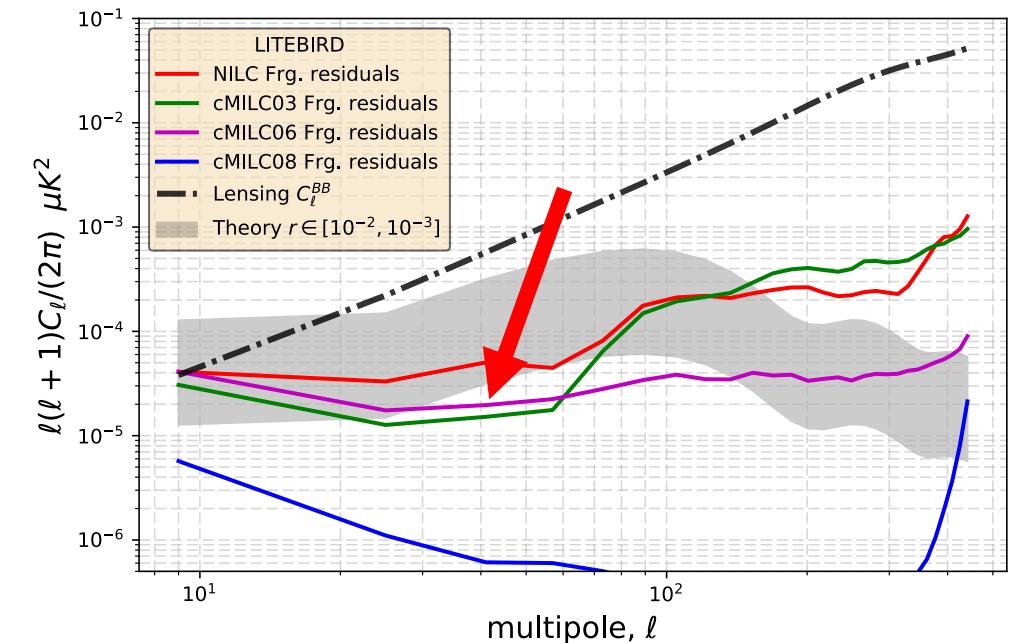
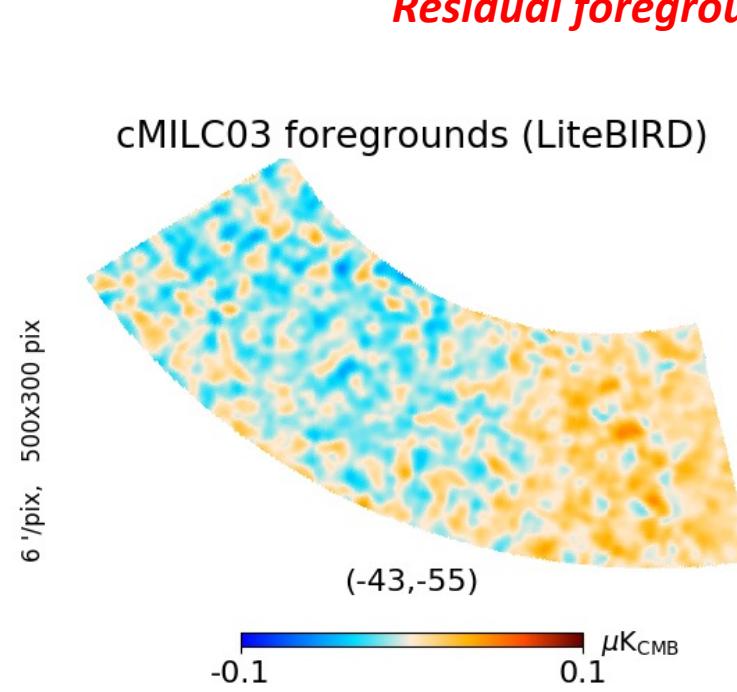


# Deprojecting the foreground moments

**cMILC:** Constrained version of NILC to deproject the spectral moments of the foregrounds arising from line-of-sight averaging

$$\hat{s}_{\text{CMB}}(\vec{n}) = \sum_{\nu} w(\nu) \cdot d(\nu, \vec{n})$$

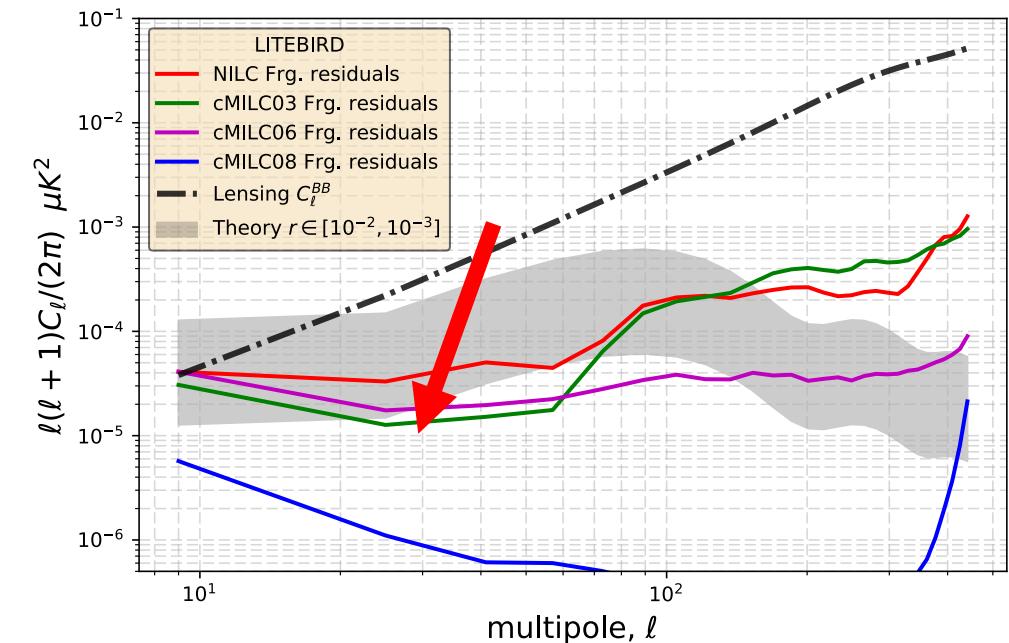
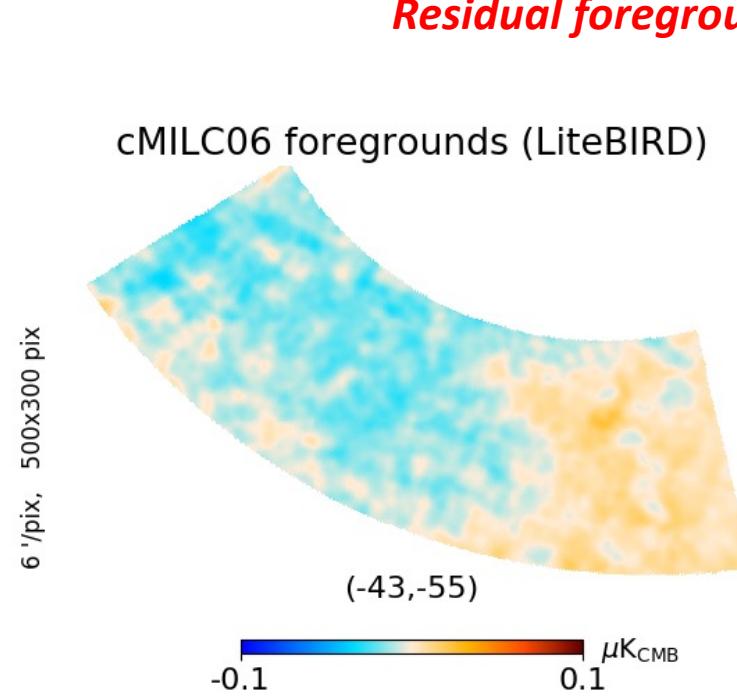
$$\begin{cases} \sum_{\nu} w(\nu) \cdot f_{\text{CMB}}(\nu) = 1 \\ \sum_{\nu} w(\nu) \cdot f_{\text{dust}}(\nu) = 0 \\ \sum_{\nu} w(\nu) \cdot f_{\text{sync}}(\nu) = 0 \end{cases}$$



# Deprojecting the foreground moments

**cMILC:** Constrained version of NILC to deproject the spectral moments of the foregrounds arising from line-of-sight averaging

$$\begin{aligned}\hat{s}_{\text{CMB}}(\vec{n}) &= \sum_{\nu} w(\nu) \cdot d(\nu, \vec{n}) \\ \left\{ \begin{array}{l} \sum_{\nu} w(\nu) \cdot f_{\text{CMB}}(\nu) = 1 \\ \sum_{\nu} w(\nu) \cdot f_{\text{dust}}(\nu) = 0 \\ \sum_{\nu} w(\nu) \cdot f_{\text{sync}}(\nu) = 0 \\ \sum_{\nu} w(\nu) \cdot \frac{\partial f_{\text{dust}}}{\partial \bar{\beta}_d}(\nu) = 0 \end{array} \right.\end{aligned}$$



# Deprojecting the foreground moments

**cMILC:** Constrained version of NILC to deproject the spectral moments of the foregrounds arising from line-of-sight averaging

$$\hat{s}_{\text{CMB}}(\vec{n}) = \sum_{\nu} w(\nu) \cdot d(\nu, \vec{n})$$

$$\sum_{\nu} w(\nu) \cdot f_{\text{CMB}}(\nu) = 1$$

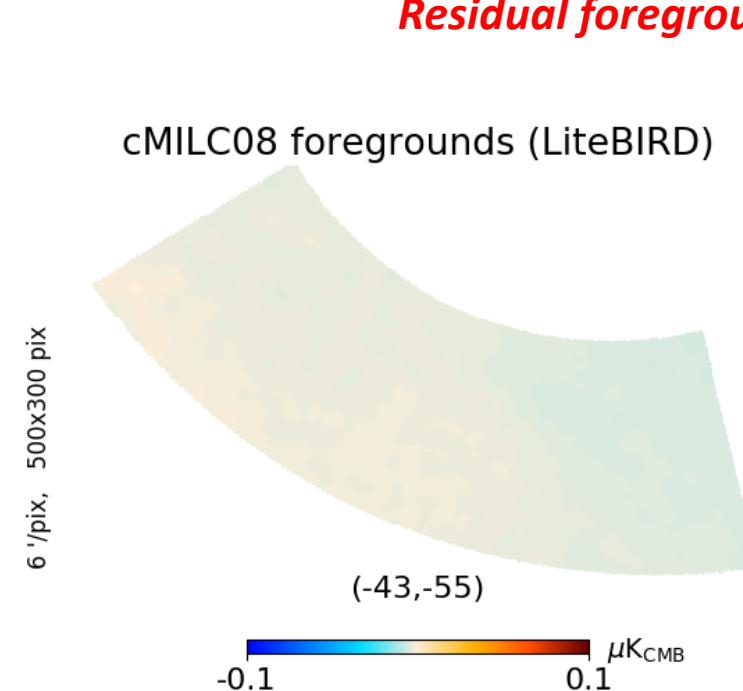
$$\sum_{\nu} w(\nu) \cdot f_{\text{dust}}(\nu) = 0$$

$$\sum_{\nu} w(\nu) \cdot f_{\text{sync}}(\nu) = 0$$

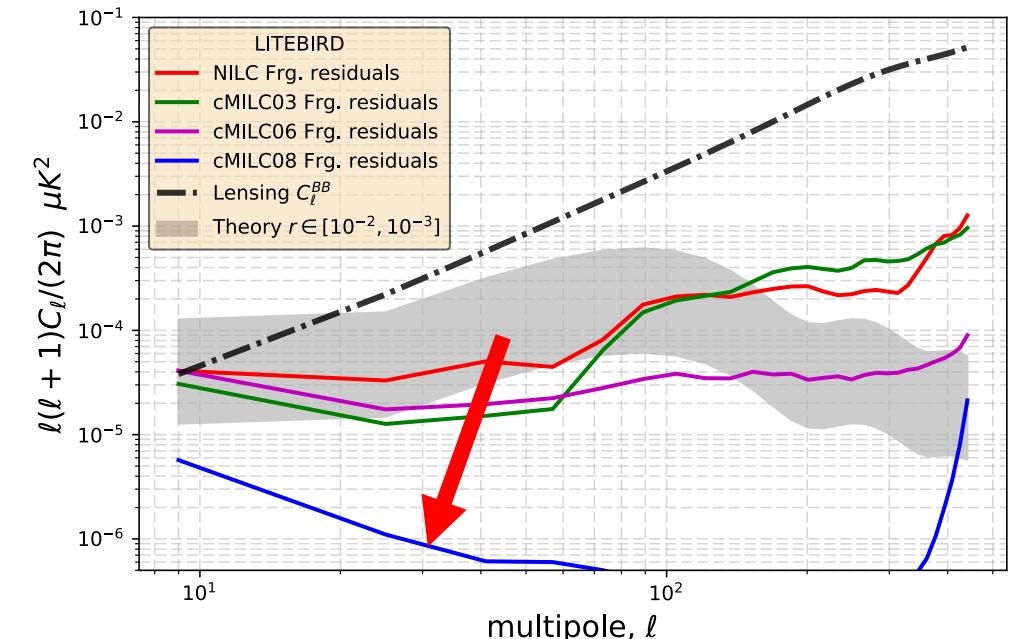
$$\sum_{\nu} w(\nu) \cdot \frac{\partial f_{\text{dust}}}{\partial \bar{\beta}_d}(\nu) = 0$$

$$\sum_{\nu} w(\nu) \cdot \frac{\partial f_{\text{sync}}}{\partial \bar{\beta}_s}(\nu) = 0$$

$$\sum_{\nu} w(\nu) \cdot \frac{\partial f_{\text{dust}}}{\partial \bar{T}_d}(\nu) = 0$$

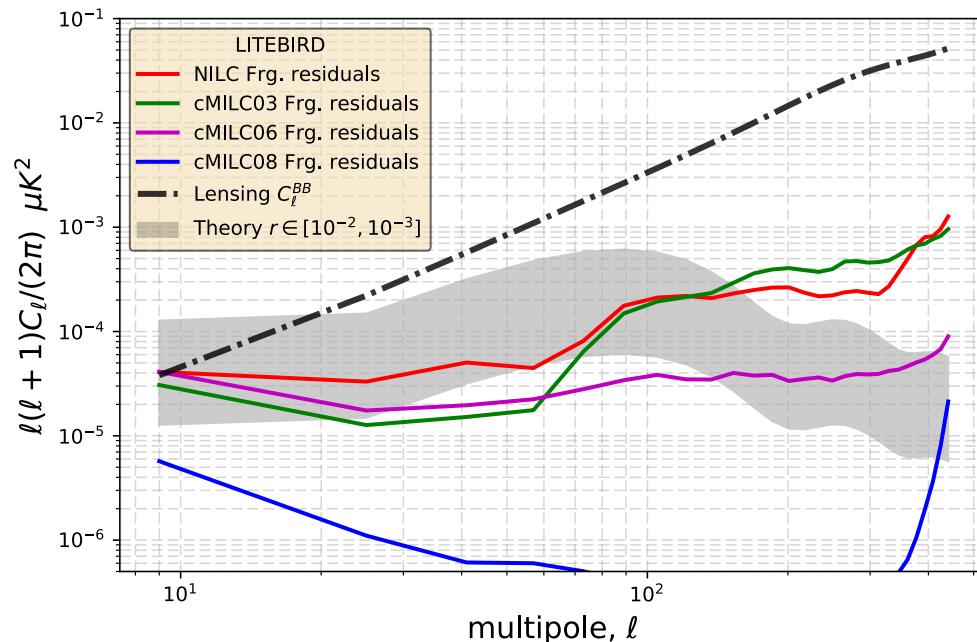


Remazeilles, Rotti, Chluba, MNRAS (2021)

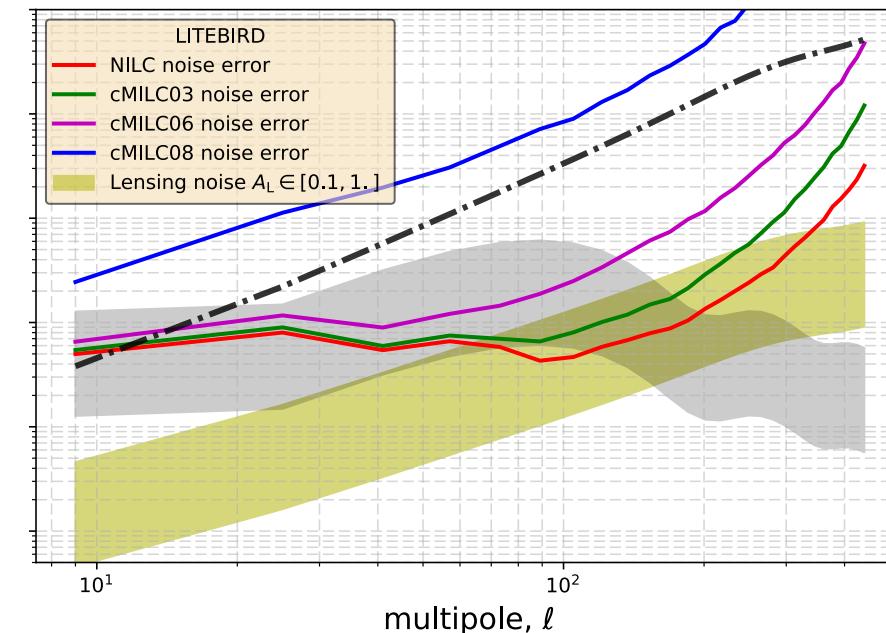


# Deprojecting the foreground moments (LiteBIRD simulation)

Foreground residuals



Noise



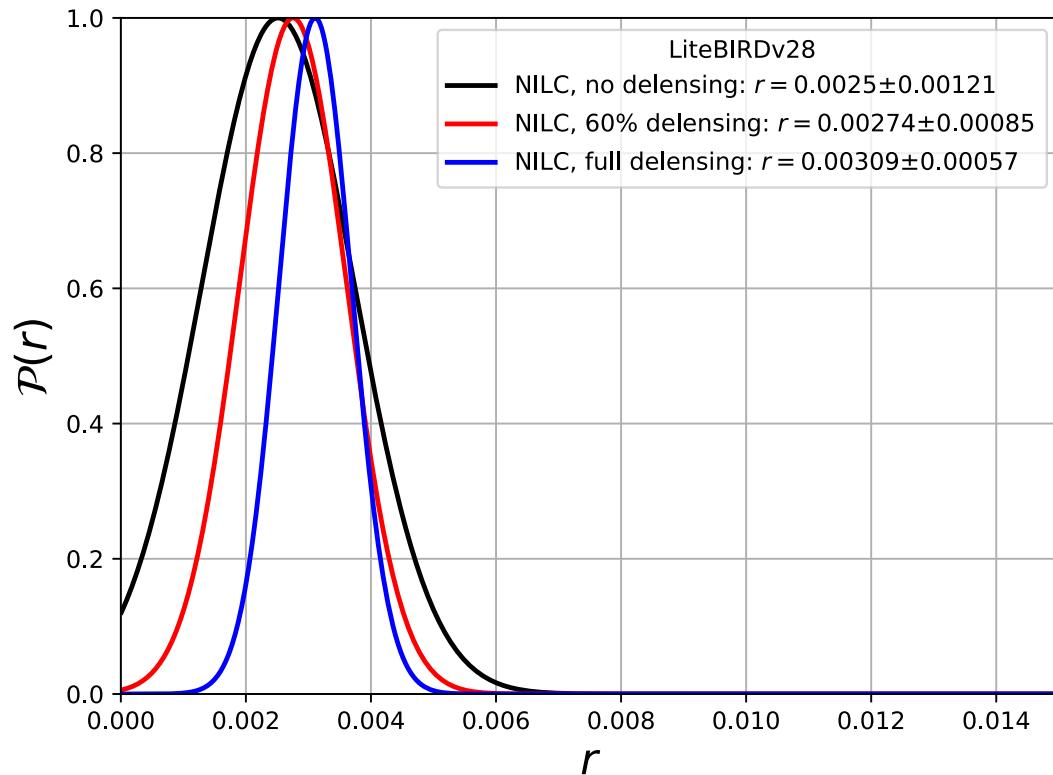
*Significant reduction of residual foreground contamination with cMILC!*

*Increase of noise as a trade-off*

# Likelihood constraints on $r = 0$

## (LiteBIRD simulation)

NILC

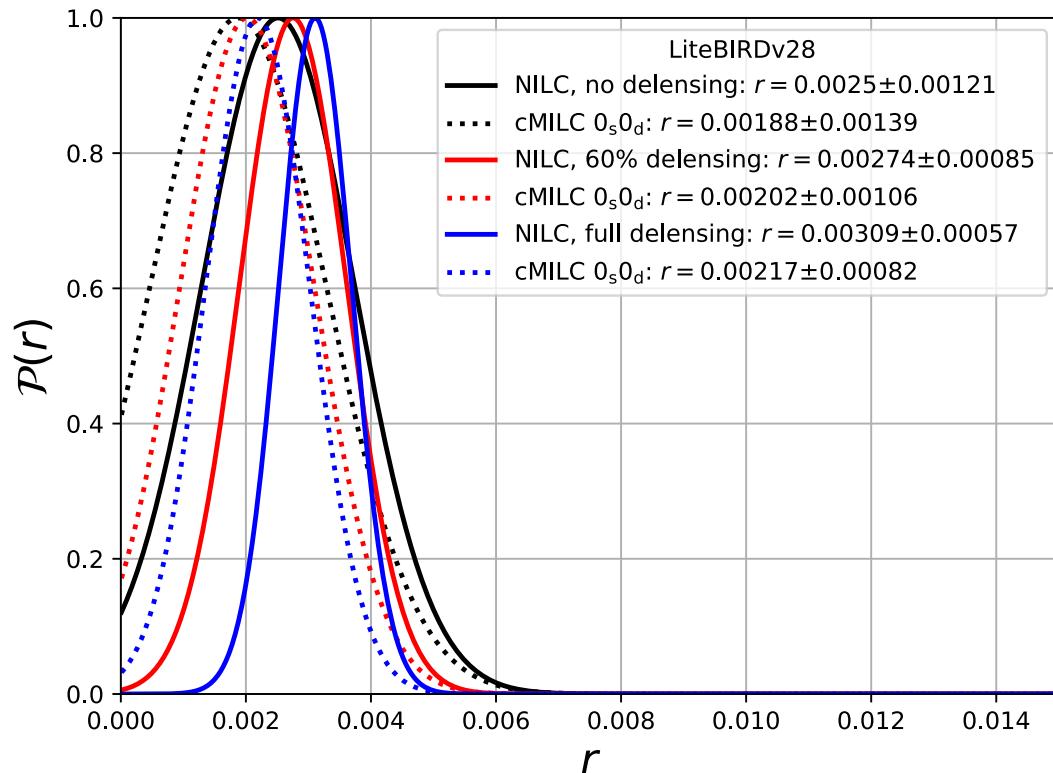


*Biased detection  $r \simeq 3 \times 10^{-3}$   
due to residual foreground  
contamination*

# Likelihood constraints on $r = 0$

## (LiteBIRD simulation)

cMILC ( $f_{\text{sync}}, f_{\text{dust}}$ )

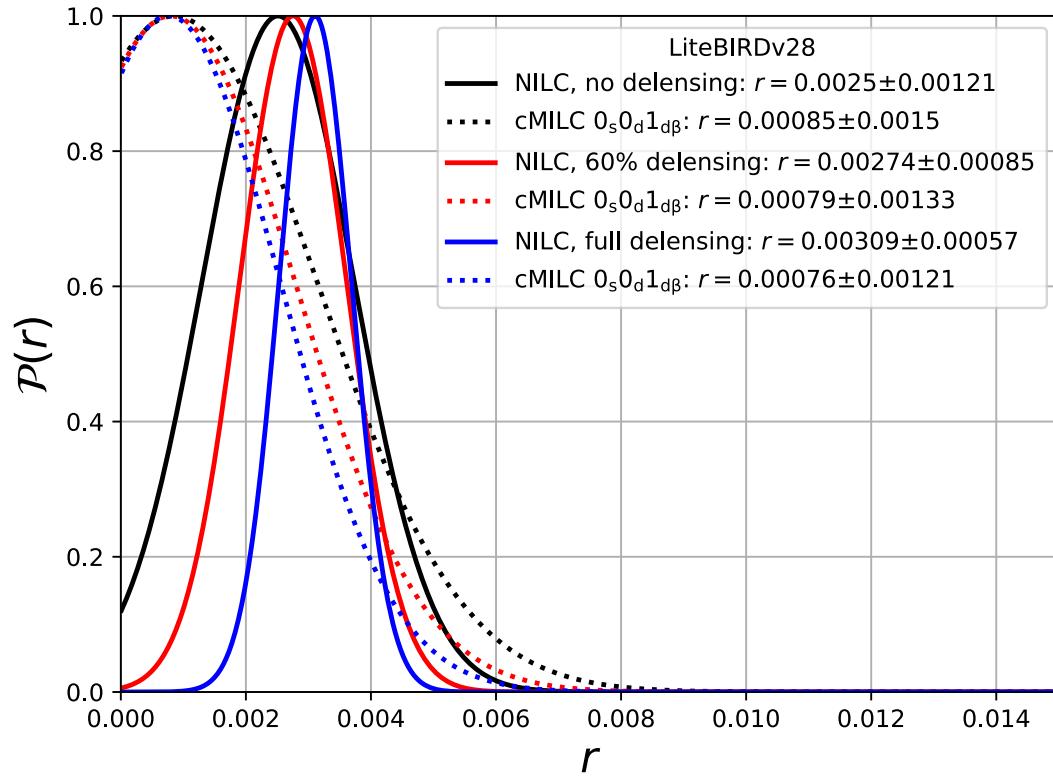


*By deprojecting moments, cMILC progressively erases residual foreground biases*

# Likelihood constraints on $r = 0$

## (LiteBIRD simulation)

cMILC ( $f_{\text{sync}}, f_{\text{dust}}, \partial_\beta f_{\text{dust}}$ )

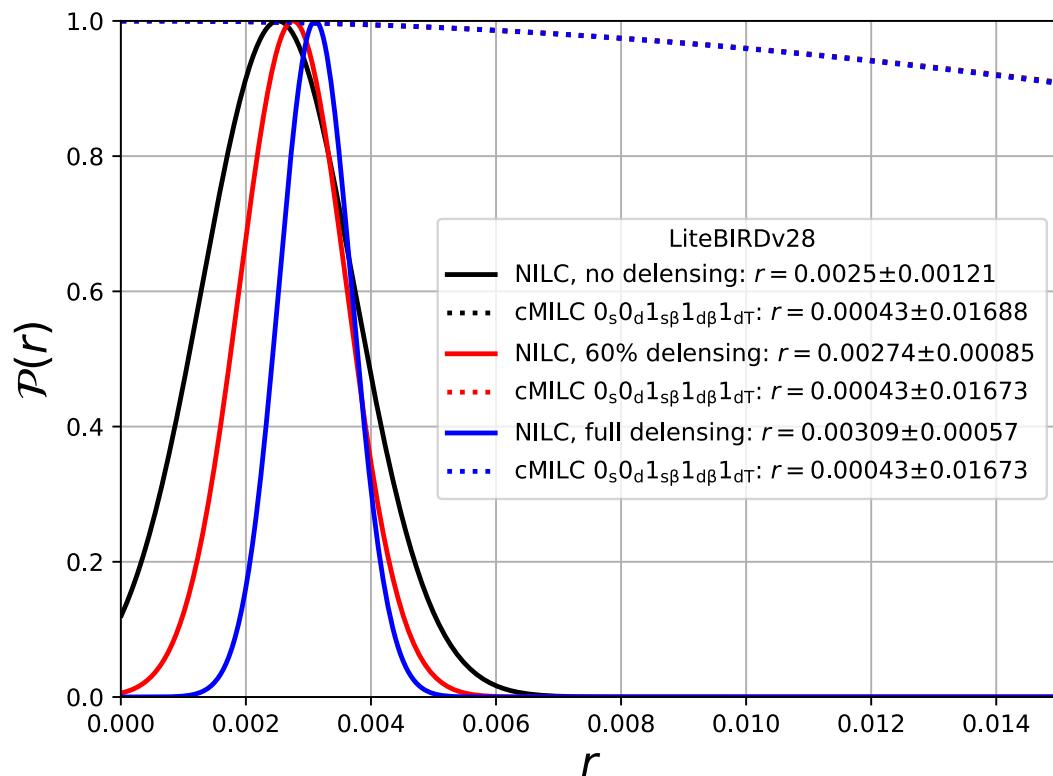


*By deprojecting moments, cMILC progressively erases residual foreground biases*

# Likelihood constraints on $r = 0$

## (LiteBIRD simulation)

cMILC ( $f_{\text{sync}}, f_{\text{dust}}, \partial_\beta f_{\text{sync}}, \partial_\beta f_{\text{dust}}, \partial_T f_{\text{dust}}$ )

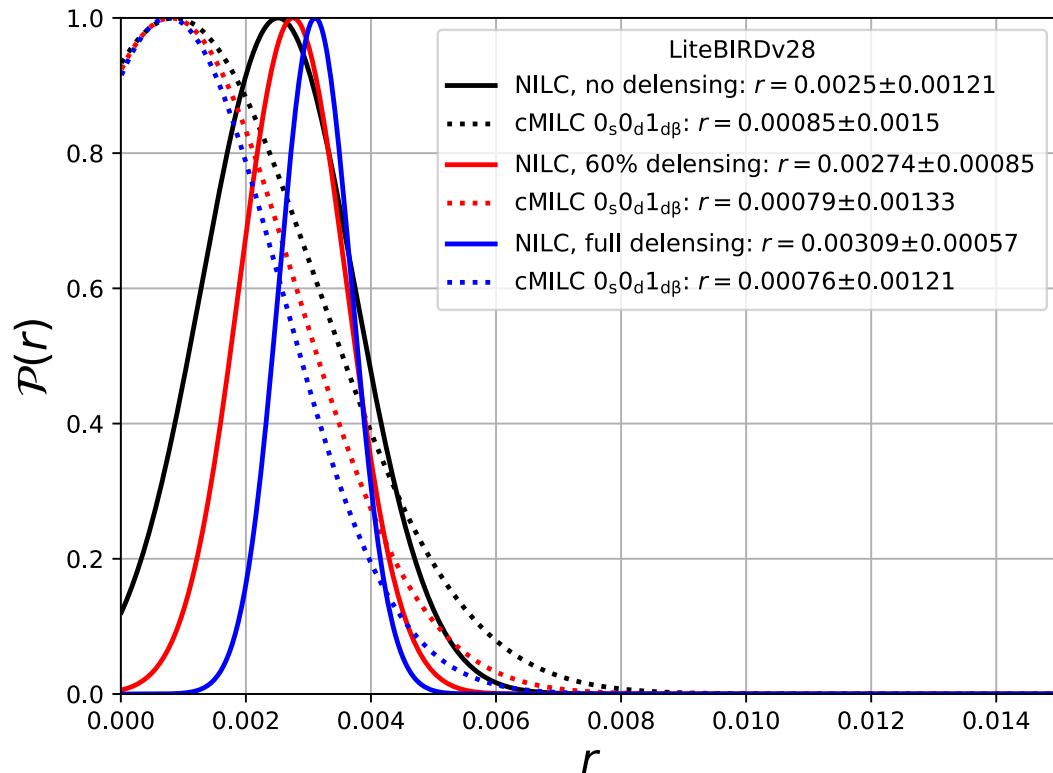


*Deprojecting dust temperature moments  
results in large noise penalty  
(LiteBIRD frequencies < 400 GHz)*

# Likelihood constraints on $r = 0$

## (LiteBIRD simulation)

cMILC ( $f_{\text{sync}}, f_{\text{dust}}, \partial_\beta f_{\text{dust}}$ )



**Sweet spot !**  
*Foreground bias removed without paying much noise penalty*

# Conclusions

- ❑ New faint signal regimes for component separation: B-mode, relativistic SZ, etc
- ❑ Poor knowledge of the spectral properties of foregrounds at the targeted sensitivities
- ❑ Extended frequency coverage is essential to get evidence of incorrect foreground modelling and biased signal detections
- ❑ Huge amplitude discrepancies between foregrounds and signals bring out new challenges:
  - Tiny modelling errors on foregrounds = large errors/biases on cosmological signals
  - Spectral distortions of the foregrounds complicate the picture
- ❑ “Thinking outside the box” for component separation is needed to overcome these challenges:
  - Moment expansion of the foregrounds
  - Foreground deprojection instead of global variance minimization
  - Thinking beyond spectral modelling for component separation to tackle spectral degeneracies