

COSMOS

Cosmic Orbital and Suborbital Microwave ObservationS



# CMB data analysis: from foregrounds to Neural Networks



Nicoletta Krachmalnicoff

# Overview

- Analysis of S-PASS data
- Validation of FG models for LiteBIRD
- Neural Networks applied to CMB

# S-PASS data analysis

Krachmalnicoff et al., *A&A*, 2018

<https://arxiv.org/abs/1802.01145>

**PARKES radio telescope: 64 m**

Frequency: **2.3 GHz** (224 MHz BW)

**Sky coverage ~ 50%** (South hemisphere)

Angular resolution ~ **9 arcmin**

## S-PASS team:

G. Bernardi

S. Brown

E. Carretti (PI)

R. Crocker

B. Gaensler

J. Farnes

M. Haverkorn

J. Malereki

M. Kesteven

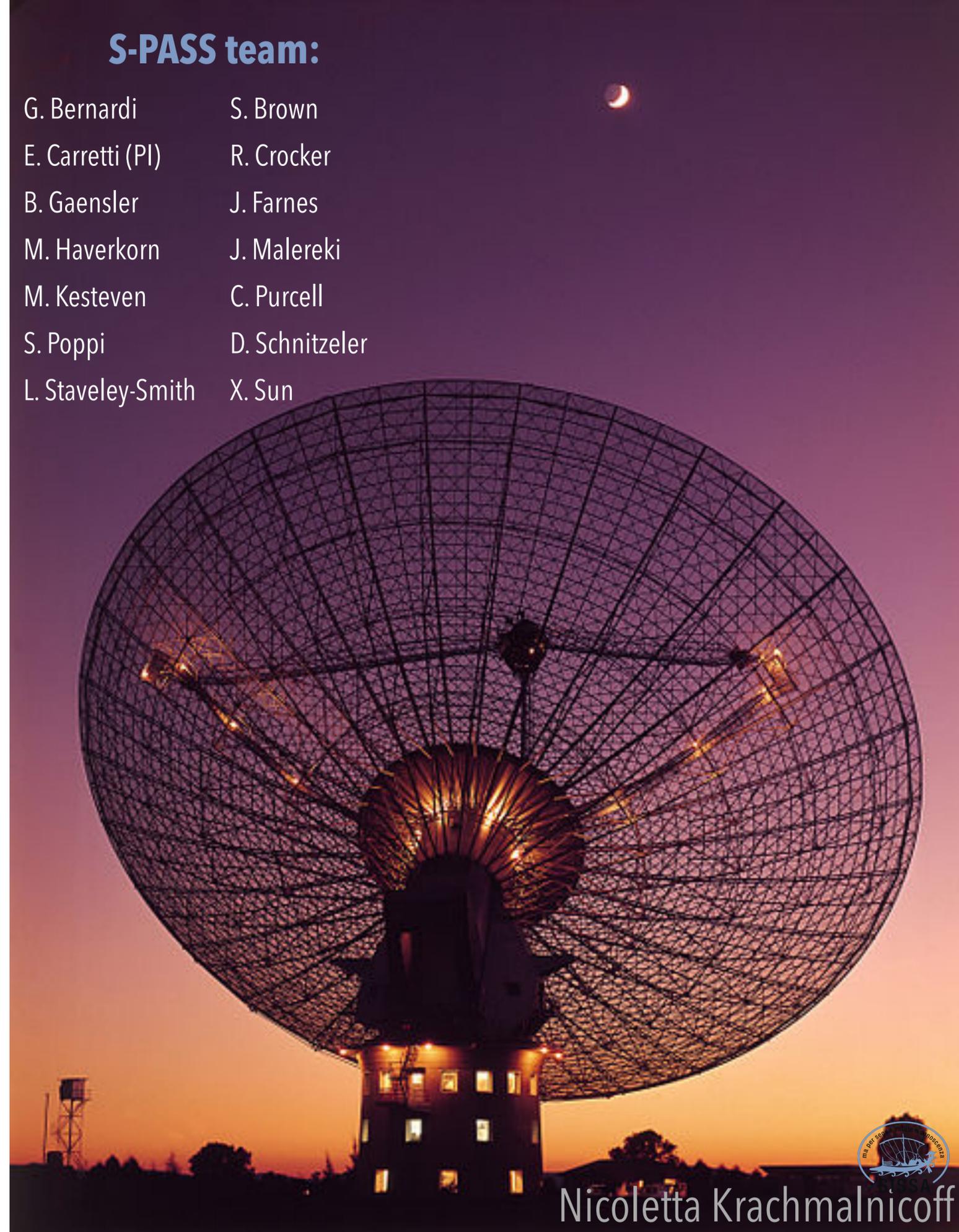
C. Purcell

S. Poppi

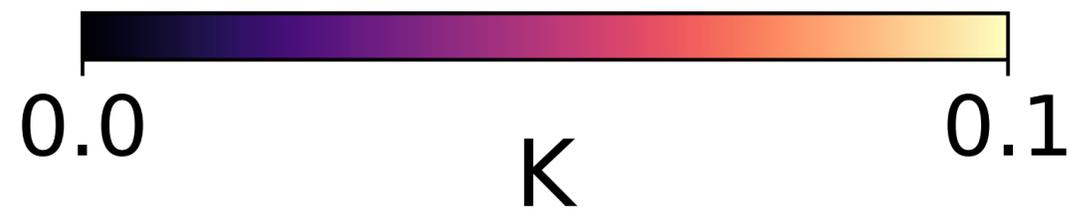
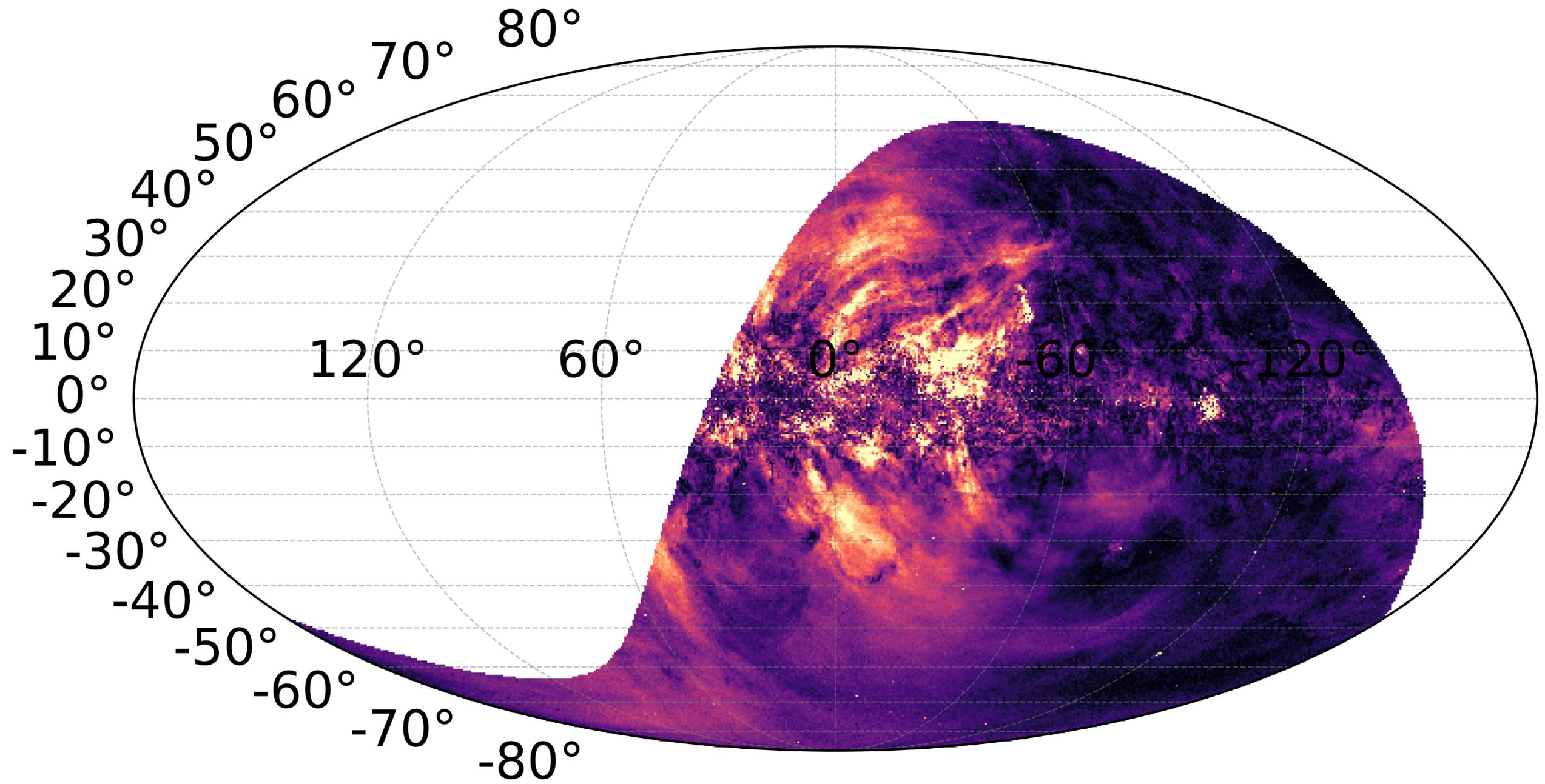
D. Schnitzeler

L. Staveley-Smith

X. Sun

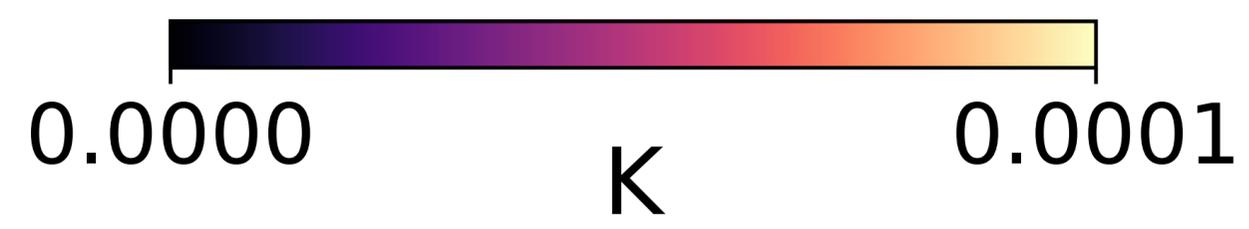
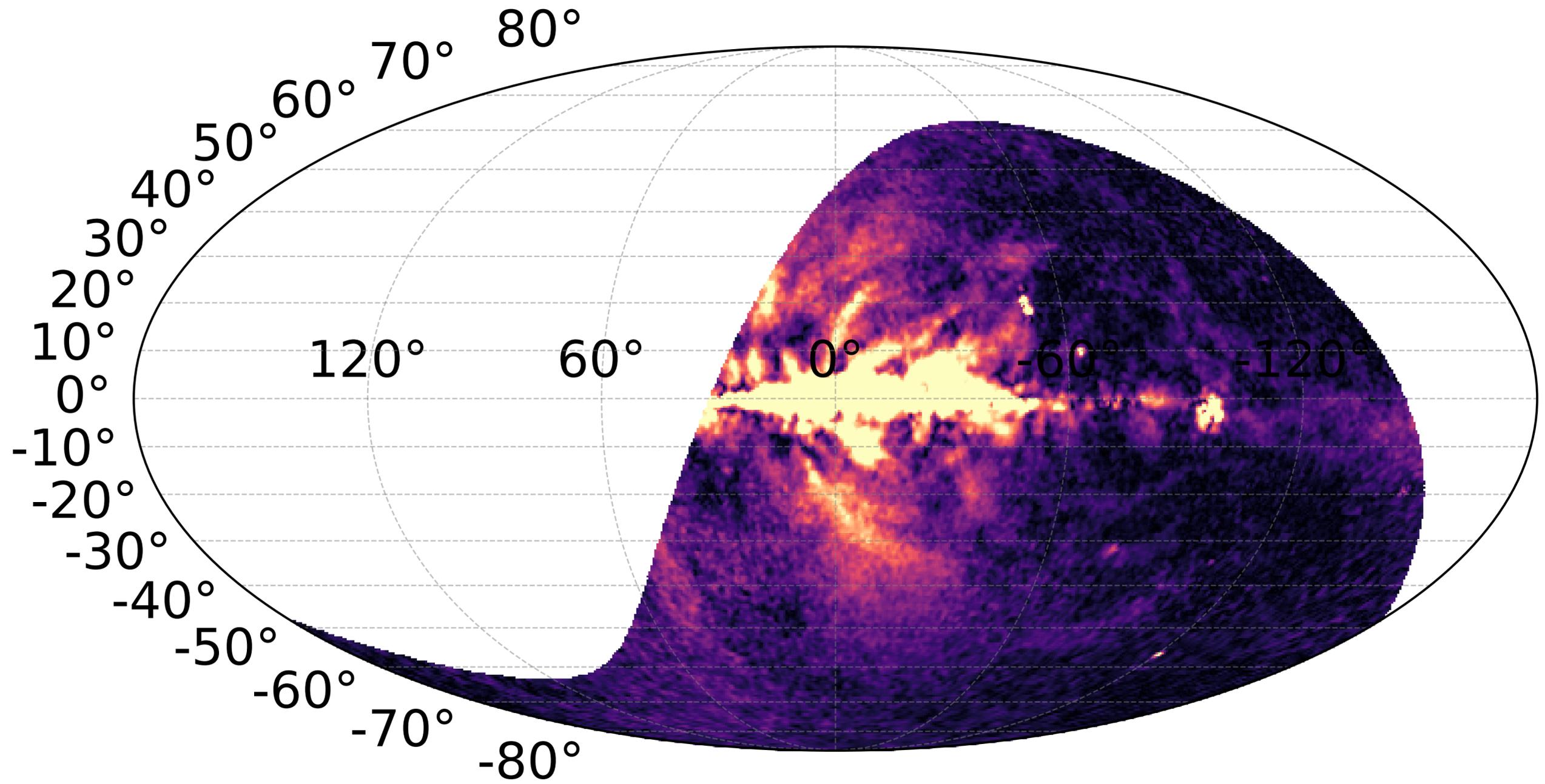


# S-PASS polarized intensity map @2.3 GHz



# WMAP-K polarized intensity map

@23 GHz



# Synchrotron SED

S-PASS / WMAP-K / LFI-30 / WMAP-Ka

2.3 GHz

23 GHz

28.4 GHz

33 GHz

$|b| > 20^\circ$

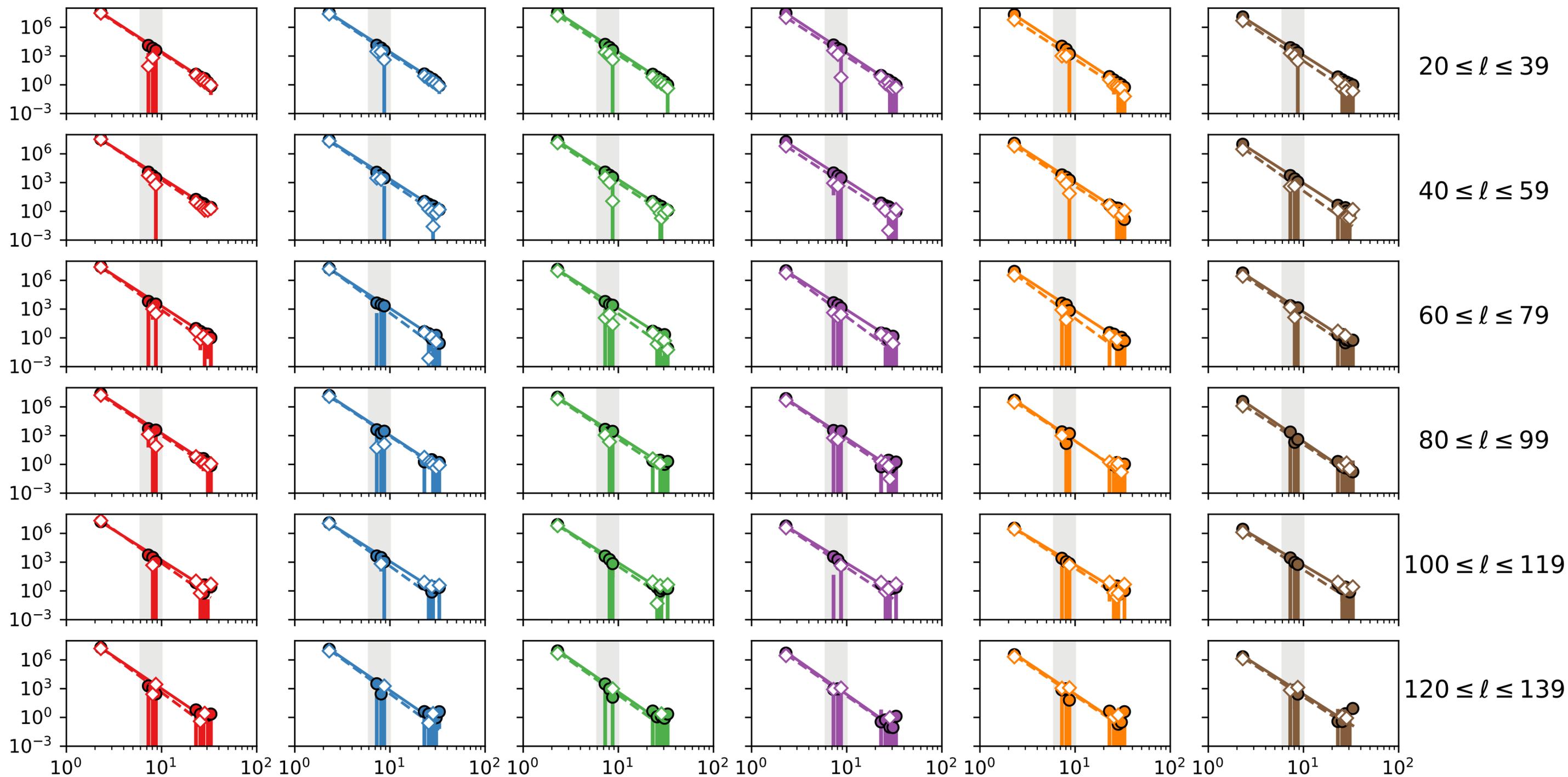
$|b| > 25^\circ$

$|b| > 30^\circ$

$|b| > 35^\circ$

$|b| > 40^\circ$

$|b| > 50^\circ$

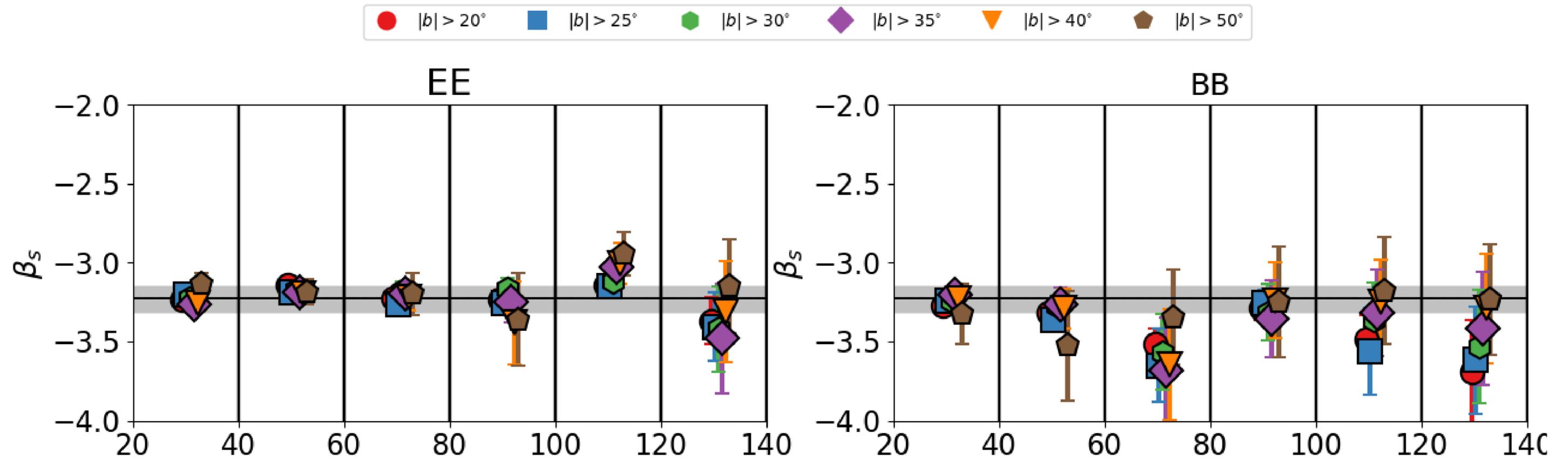


$D_\ell(\nu_1\nu_2) [\mu K^2]$

Effective Frequency:  $\sqrt{\nu_1\nu_2}$  [GHz]



# Synchrotron SED

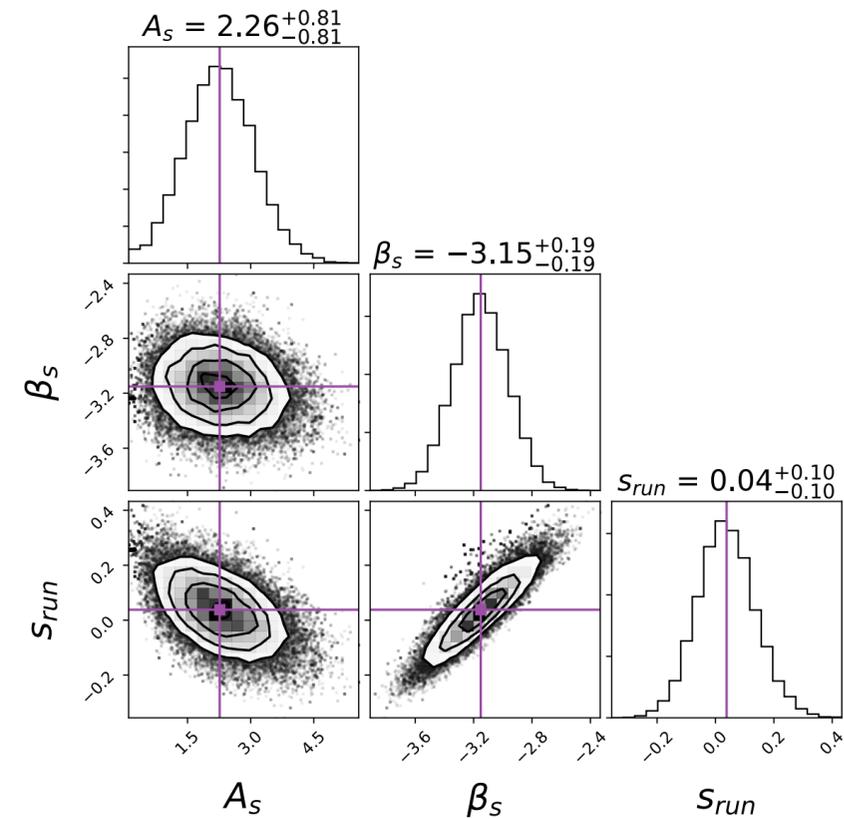


$$\beta_s = -3.22 \pm 0.08$$

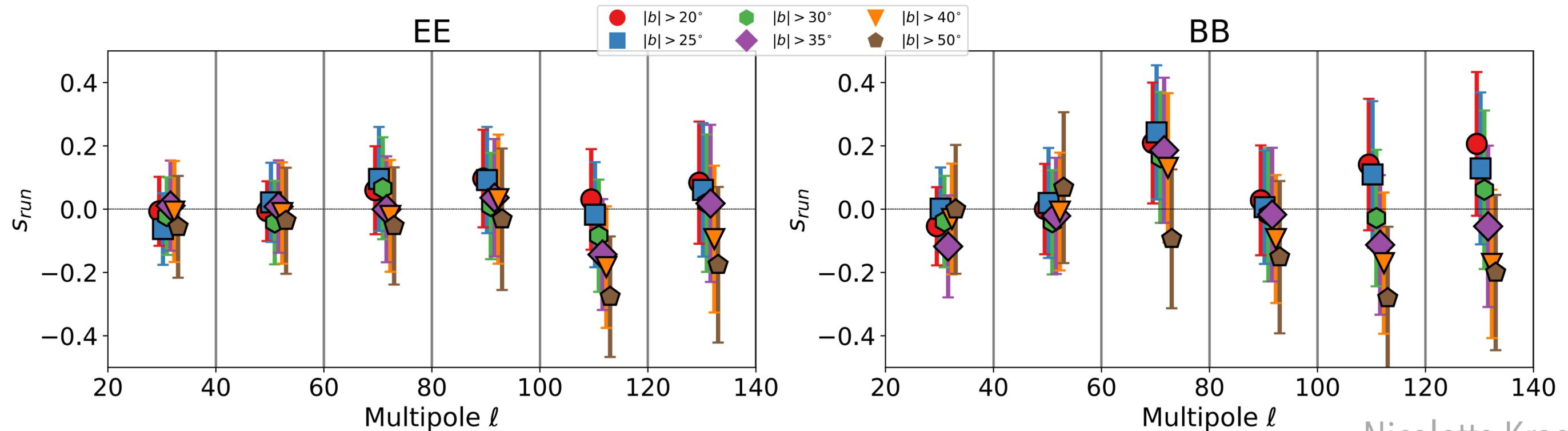
- ◆ Constant along the multipole range and for E and B-modes
- ◆ In agreement with constraints coming from WMAP and Planck

# Constraints on curvature

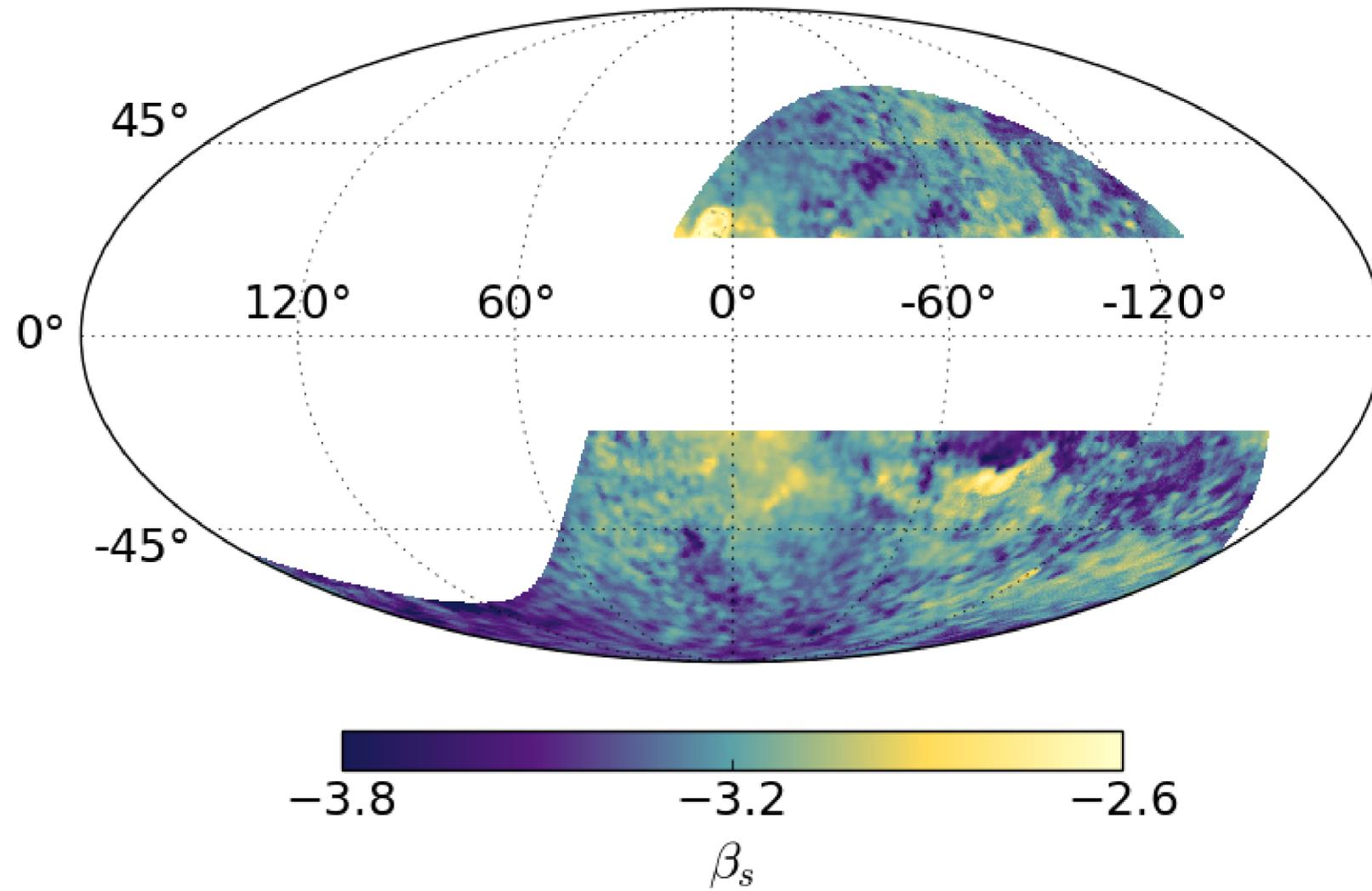
$$D_\ell(\nu_1 \times \nu_2) = A_s \left( \frac{\nu_1}{\nu_0} \right)^{\beta_s + s_{run} \log(\nu_1/\nu_0)} \left( \frac{\nu_2}{\nu_0} \right)^{\beta_s + s_{run} \log(\nu_2/\nu_0)}$$



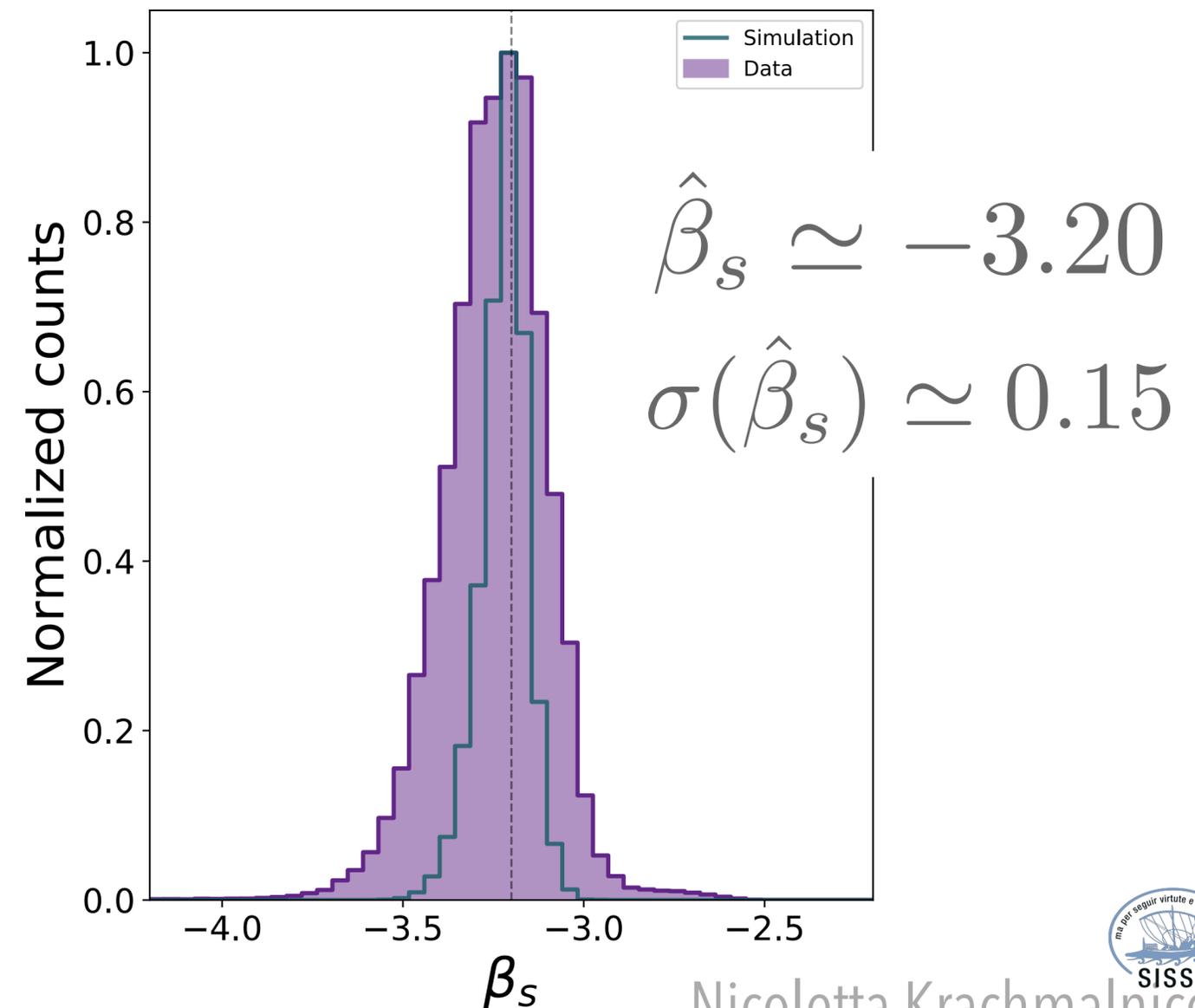
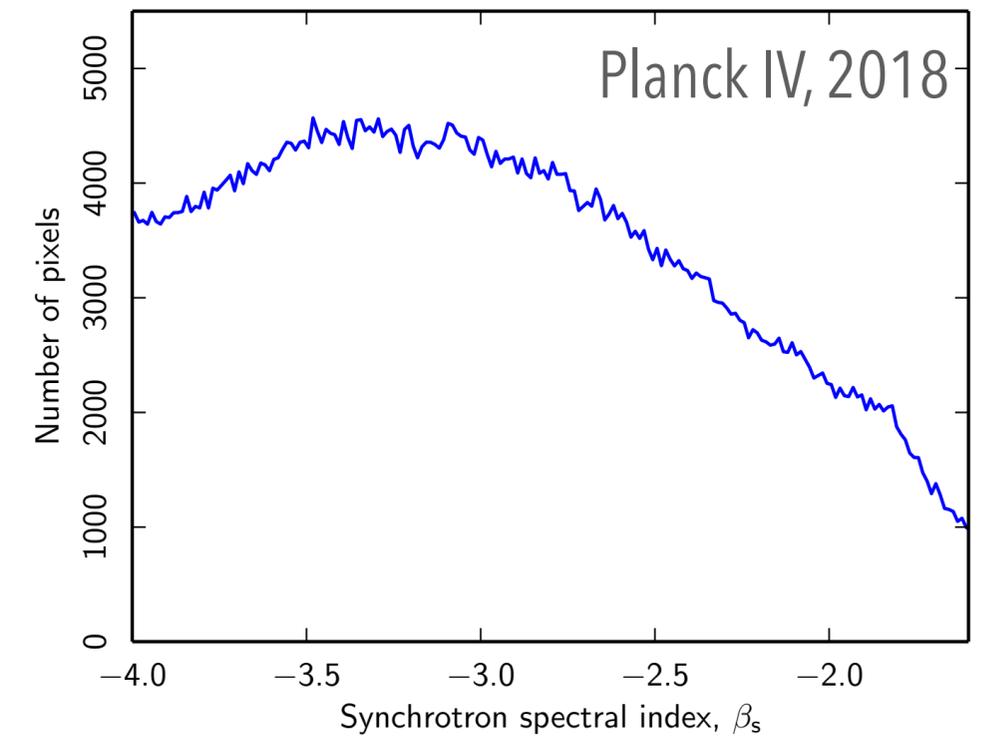
- ◆ Strong degeneracy between  $\beta_s$  and  $s_{run}$
- ◆ **Gaussian prior on spectral index** from WMAP and Planck:  $\beta_s = -3.13 \pm 0.13$
- ◆  $s_{run}$  **compatible with zero**, with  $1\sigma$  errors between 0.07 and 0.14
- ◆ More data at intermediate frequencies are needed (C-BASS in south, QUIJOTE and C-BASS in north)



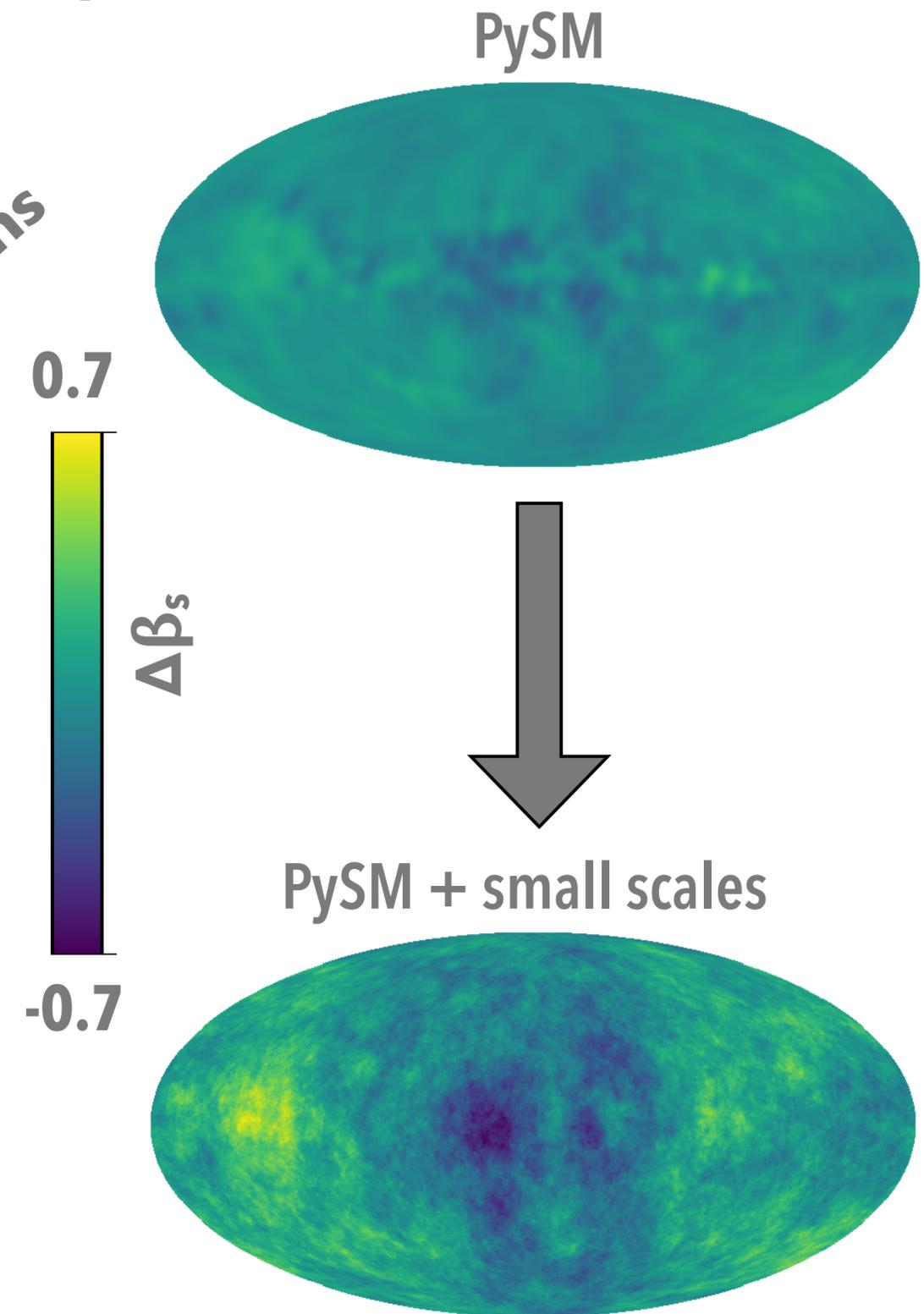
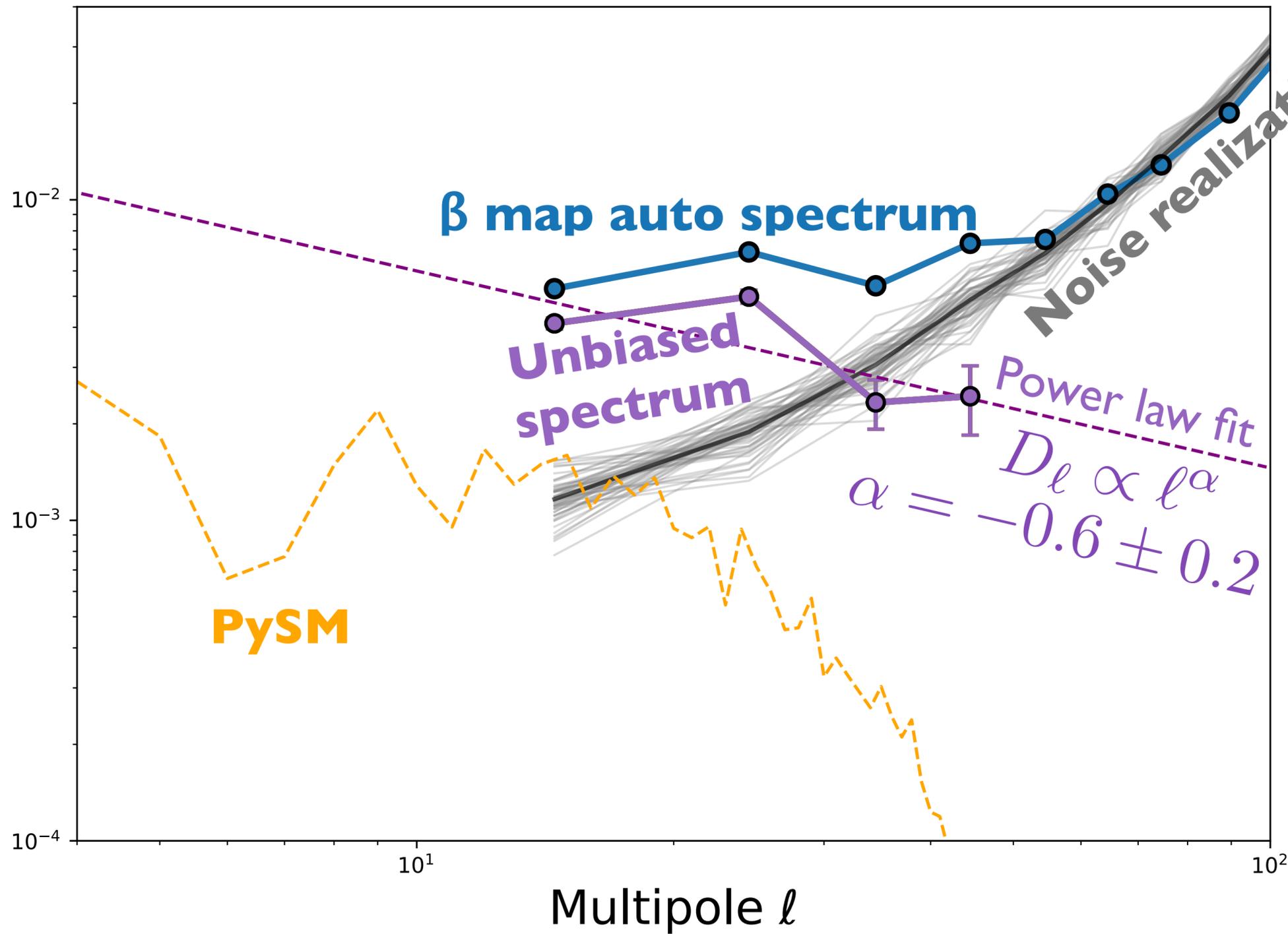
# Synchrotron spectral index map



- ◆ Power law fit in range **2.3 - 33 GHz**
- ◆ Fit in each pixel in **total polarized intensity** taking into account the noise bias
- ◆ **Angular resolution of 2°**
- ◆ Sky coverage  $\sim 30\%$
- ◆ No prior



# Power spectrum of spectral index map

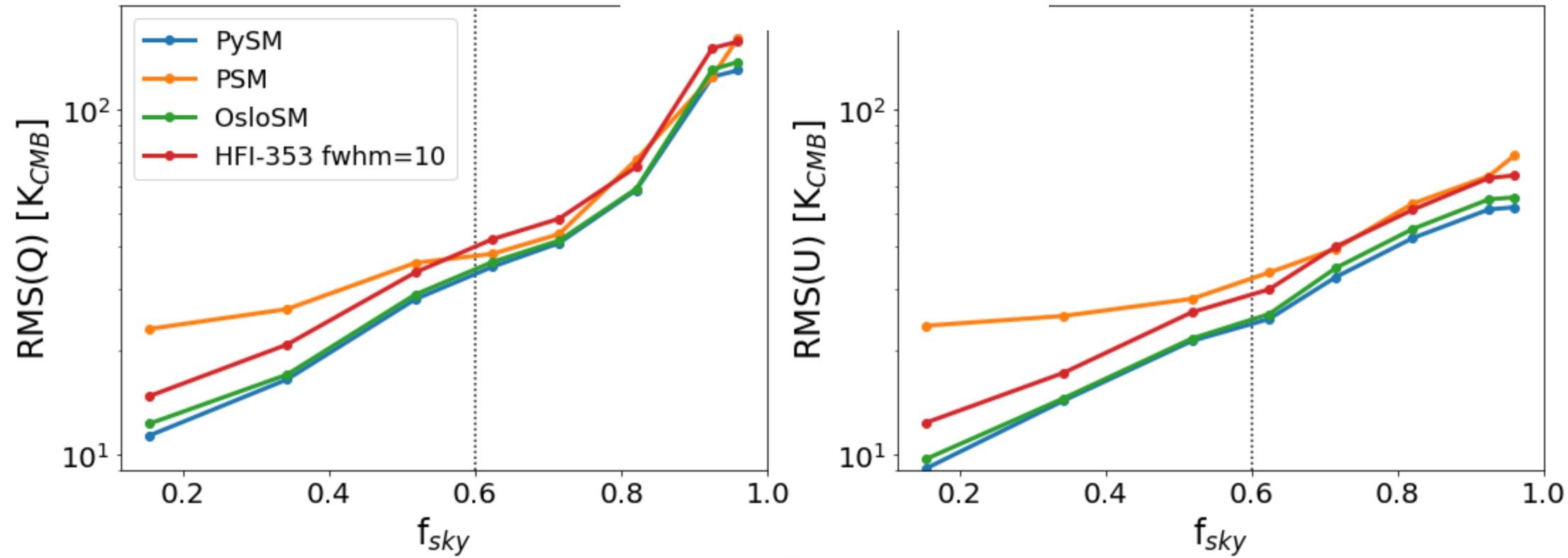


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- **Validation of FG models for LiteBIRD**
- Neural Networks applied to CMB

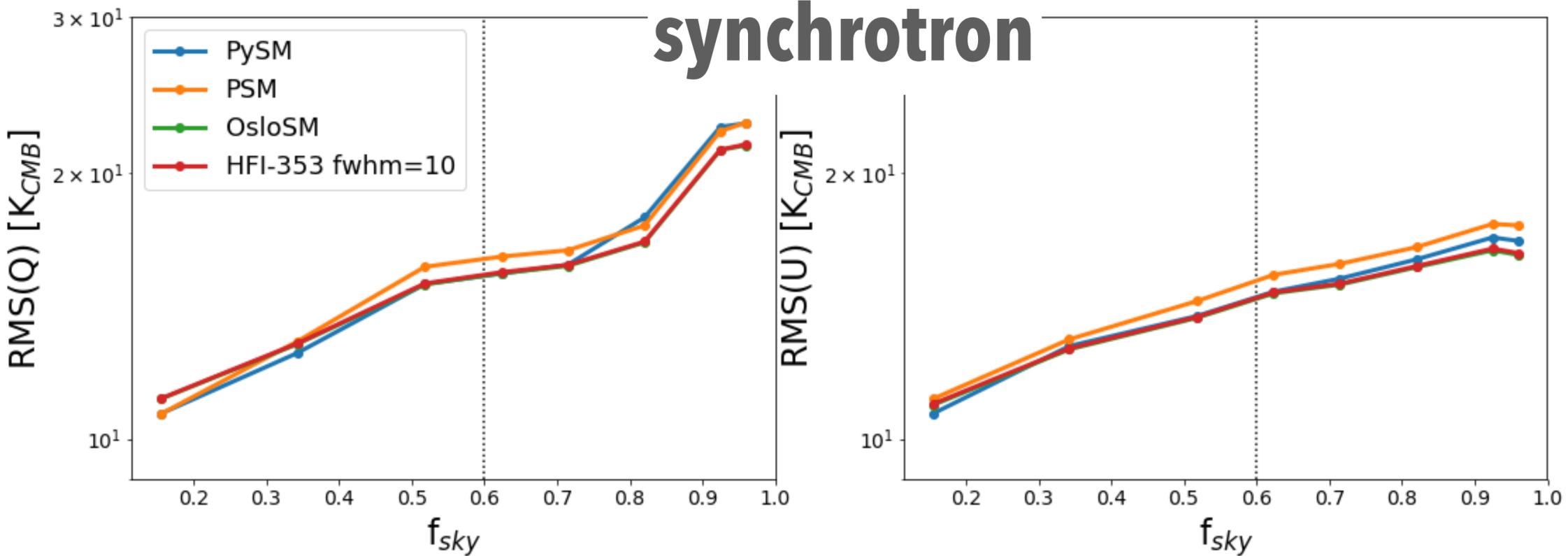
# Validation of FG sky models (for LB but not only)

## thermal dust



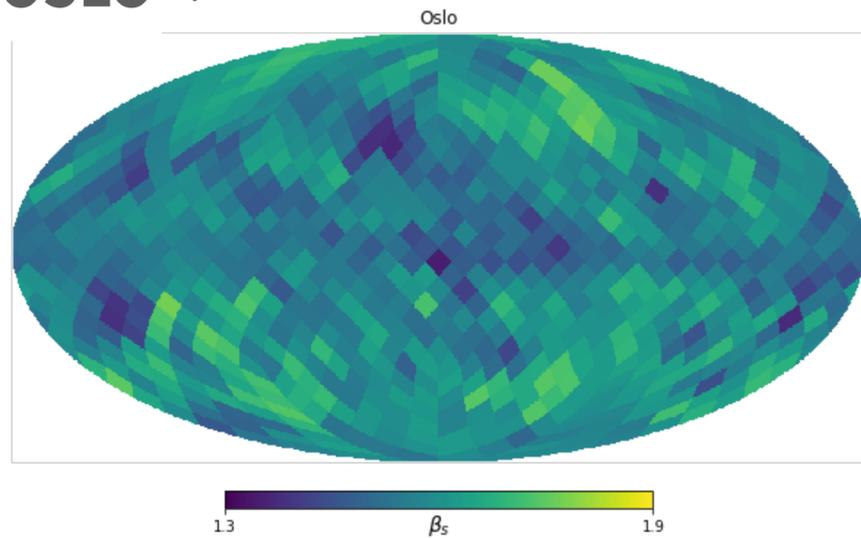
Q and U RMS of **thermal dust** and **synchrotron templates** as a function of the sky masks

## synchrotron

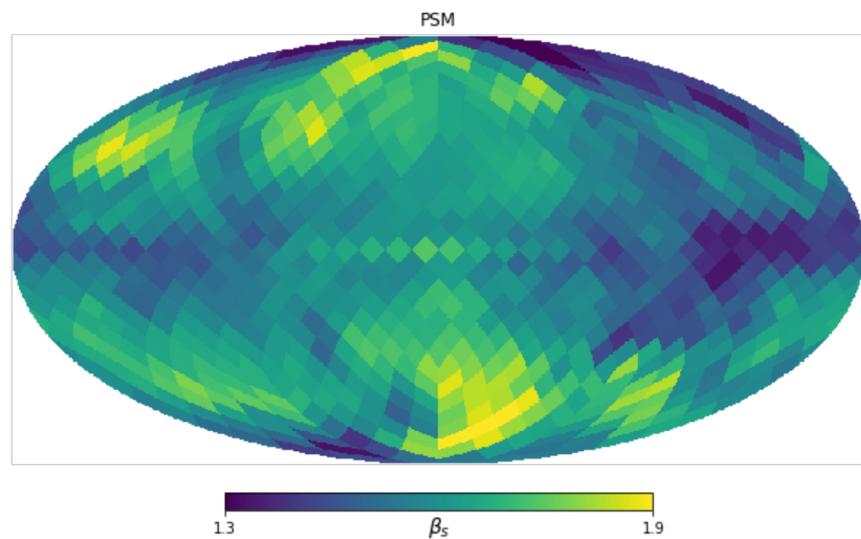


# $\beta_d$ maps

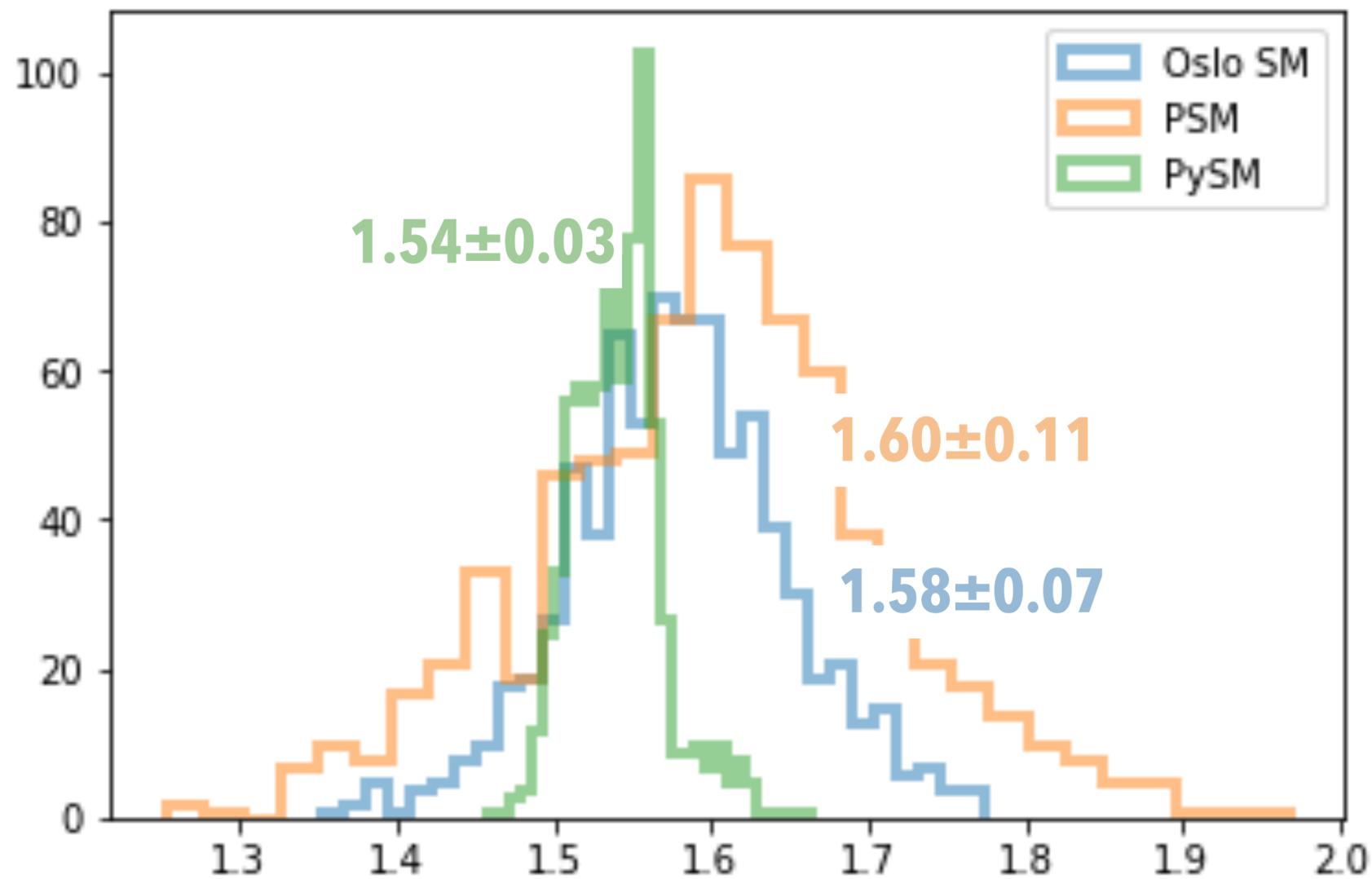
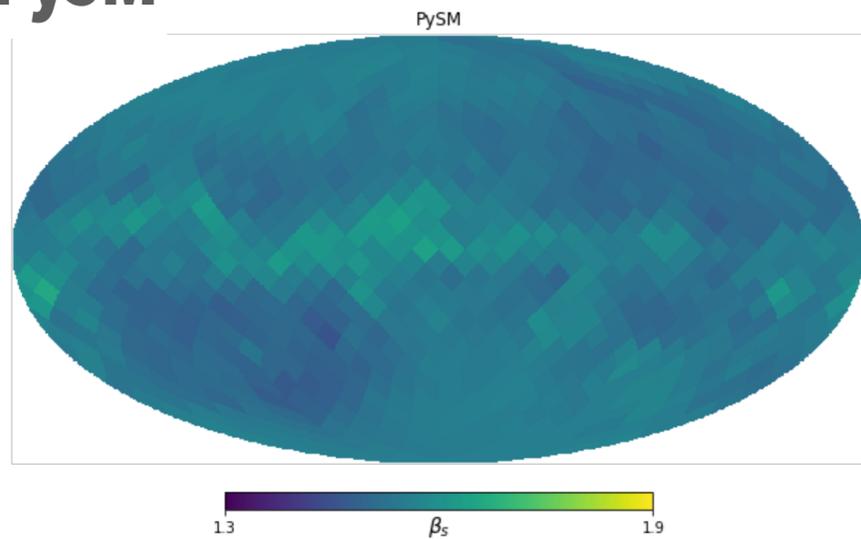
OSLO ?



PSM GNILC from temperature?



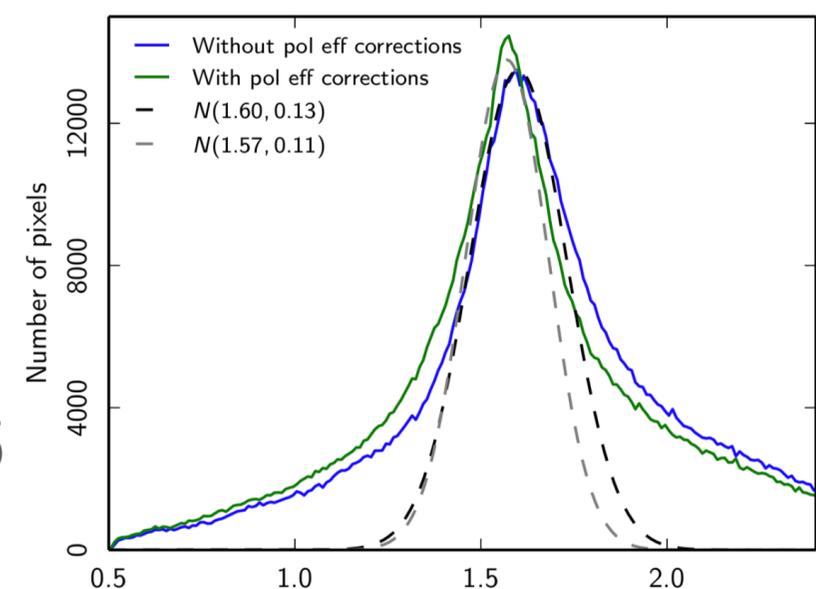
PySM ?



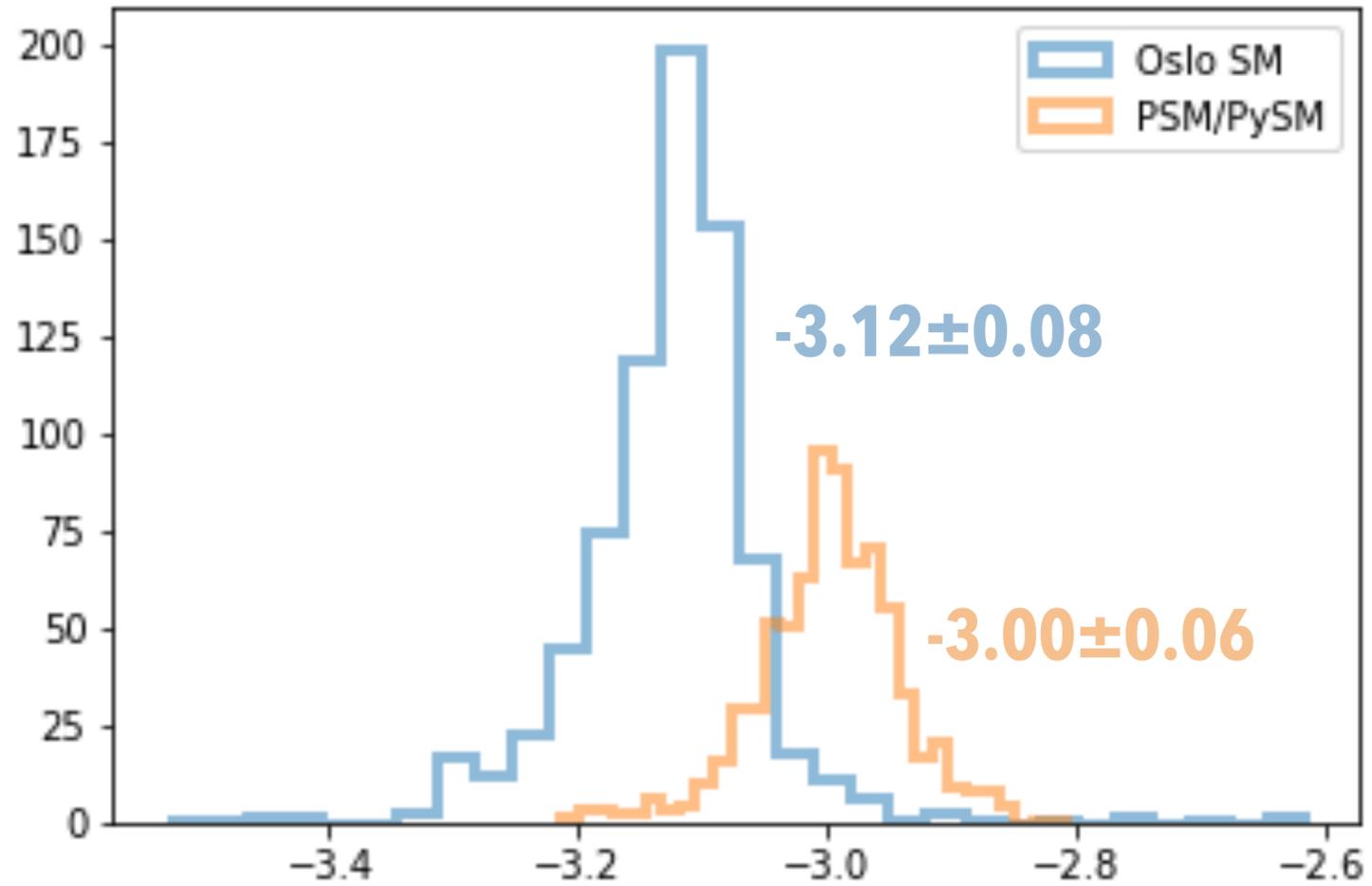
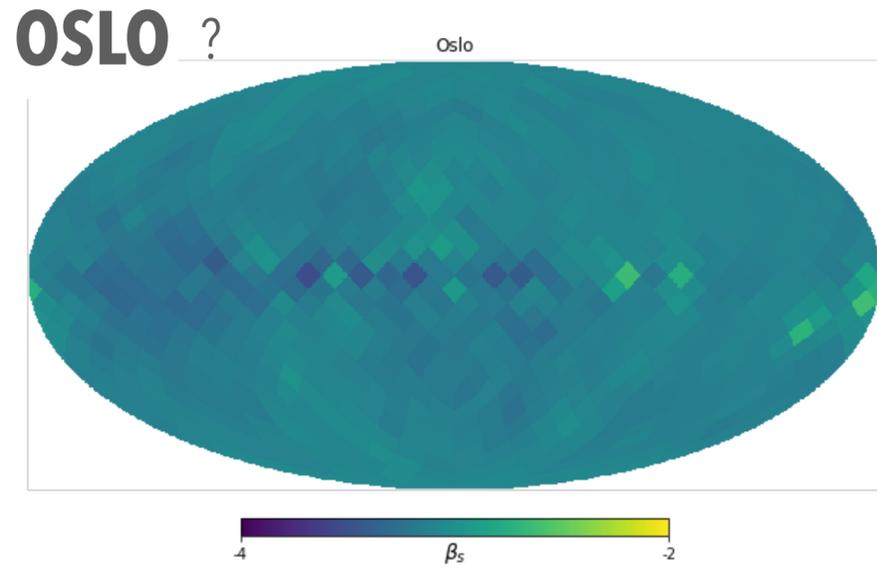
## Constraints from data:

Planck 2018, XI:  $\beta_d = 1.53 \pm 0.02$   
 $\beta_d^P - \beta_d^I = 0.05 \pm 0.03$   
 (from spectra)

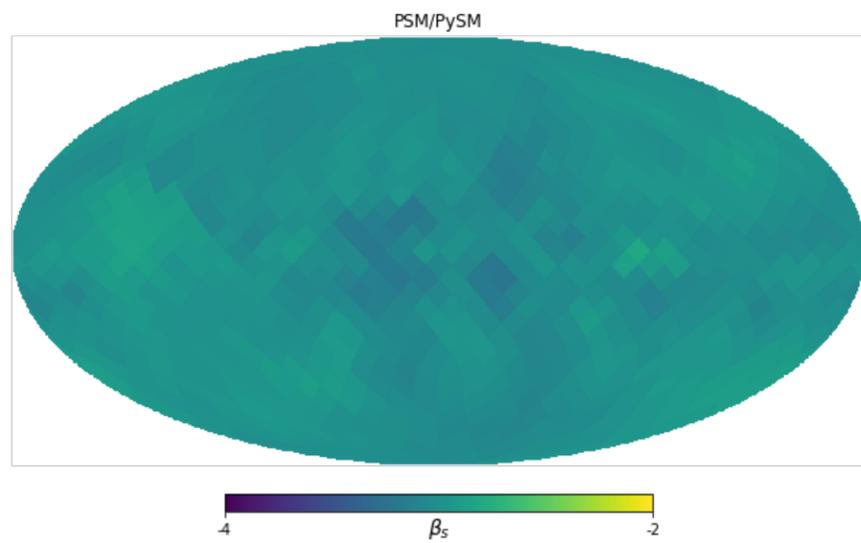
Planck 2018, IV:  $\beta_d = 1.55 \pm 0.05$   
 (from spectra)



# $\beta_s$ maps



**PSM/PySM** from WMAP pol + Haslam I



## Constraints from data:

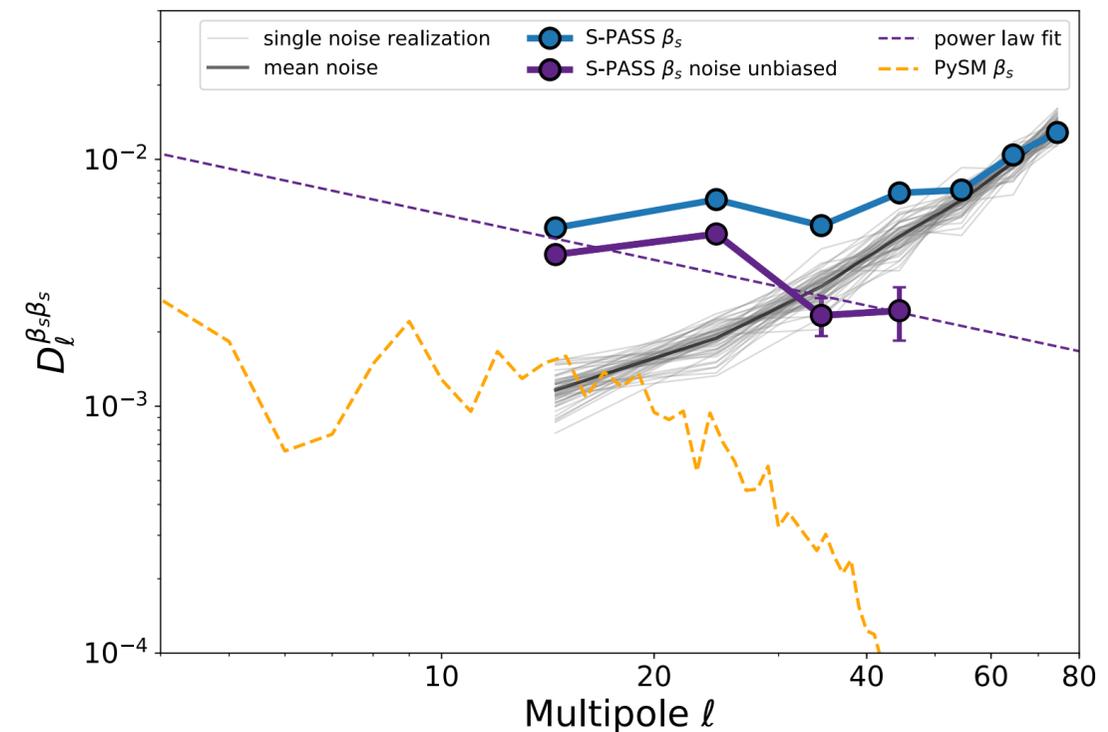
Planck 2018, XI:  $\beta_s = -3.13 \pm 0.13$   
(from spectra)

Planck 2018, IV:  $\beta_s = -3.10 \pm 0.10$   
(from spectra)

S-PASS:  $\beta_s = -3.22 \pm 0.08$   
(from spectra)

S-PASS:  $\beta_s = -3.20 \pm 0.15$   
(from maps)

C-BASS ? :  $\beta_s \sim -3.0$



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- **Neural Networks applied to CMB**

# Neural Networks

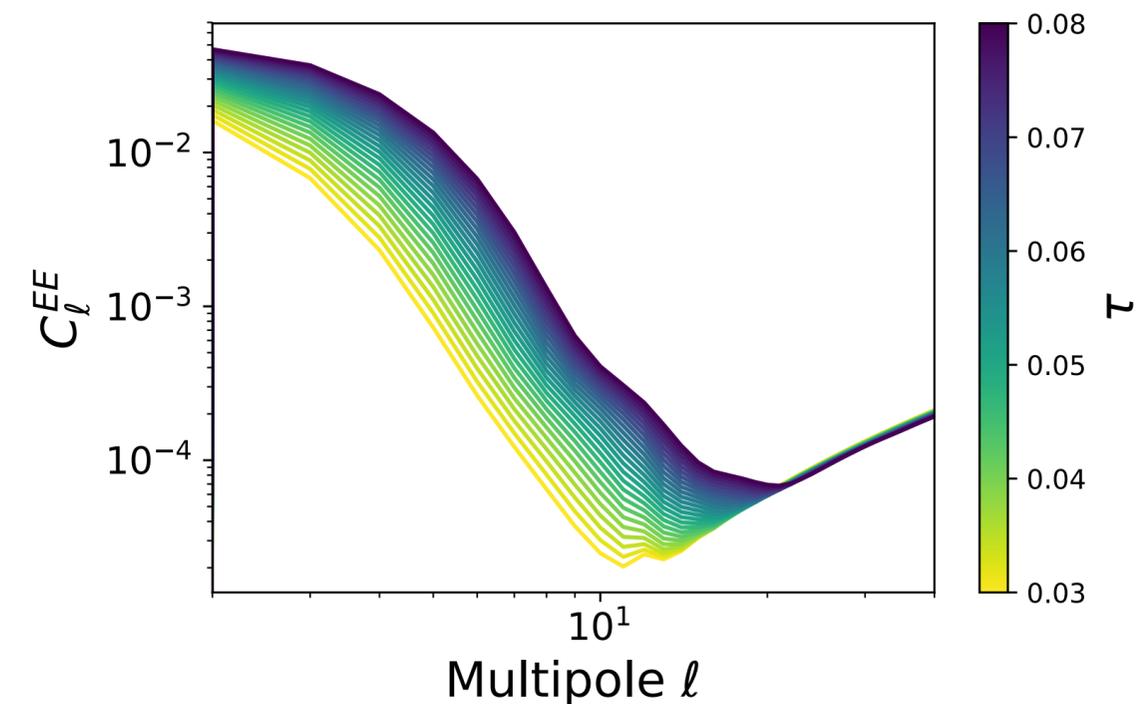
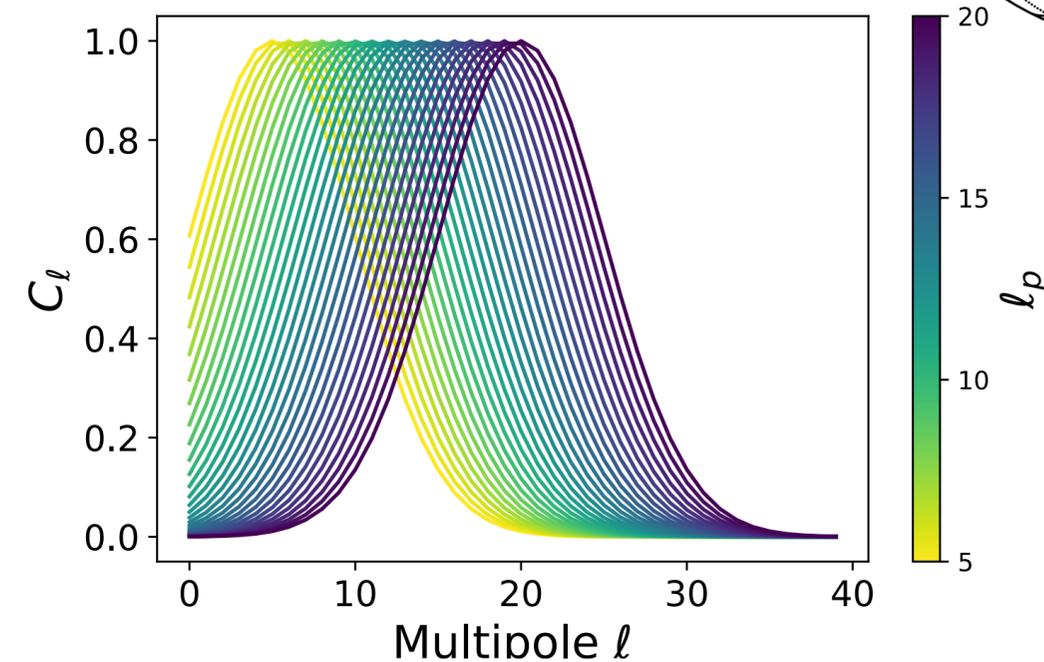
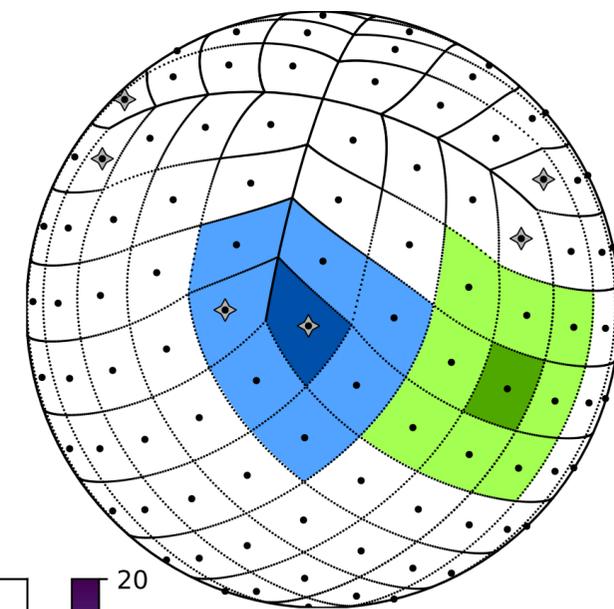
- **What are Neural Networks?**
  - ▶ mathematical operators able to approximate complex functions that map input into output
- **How do they work?**
  - ▶ by recursively applying non-linear functions to linear combination of input elements
  - ▶ weights of the linear combination must be optimized during training
  - ▶ set of inputs for which the output is known is needed to do the training
- **Why to apply NNs to CMB data analysis?**
  - ▶ NNs are experiencing fast growth in many fields of science
  - ▶ Many applications (from data flagging to component separation and parameter estimation) could be performed with NNs
  - ▶ Different approach with respect to standard methods, good for validation!

# NNs for parameter estimation

Krachmalnicoff & Tomasi, 2019, submitted to A&A

<https://arxiv.org/abs/1902.04083>

- Developed **pixel-based algorithm** to apply **Convolutional Neural Network to HEALPix maps**
- Able to **fit for cosmological parameters directly from maps** on full/partial sky, from temperature or polarization map, and in presence of noise
- first **test on estimation of tau on simple simulations** showing promising results



# Summary and prospective

- **S-PASS data analysis at 2.3 GHz**
  - ▶ most advance characterization of synchrotron emission in polarization for CMB
  - ▶ synchrotron is complex
  - ▶ joint analysis with C-BASS and QUIJOTE
- **FG model validation**
  - ▶ ongoing work for LiteBIRD
  - ▶ important also for other CMB experiments, setting the framework for a common sky modeling
- **NNs applied to CMB**
  - ▶ preliminary work proved the feasibility of the approach
  - ▶ full test on tau to be done soon, with detailed comparison with standard methods in realistic case
  - ▶ Possibile application also for component separation under study