

Simultaneous Constraints on Foreground and Global 21-cm Models via a Novel Pattern Recognition Technique

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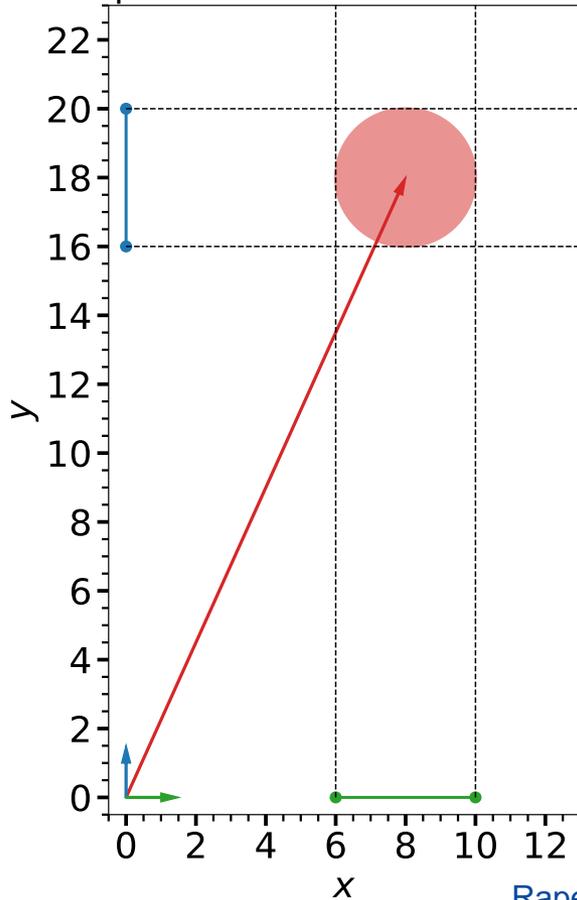


JOINT CONSTRAINTS ON PHYSICAL PARAMETERS

- In [Paper I](#) (Tauscher, Rapetti, Burns & Switzer 2018), we presented a new method based on [Singular Value Decomposition \(SVD\)](#) and [Information Criteria](#) to obtain [spectral constraints](#) on the global 21-cm signal.
- Converting spectral constraints into [constraints on any physical parameter space of choice](#) is presented in [Paper II](#) (Rapetti, Tauscher, Mirocha & Burns, to be submitted).
- This allows us to analytically find a [joint linear fit](#) of the signal and systematics (currently, foreground) to be readily used as starting point (mean and covariance) for our [simultaneous, nonlinear Markov Chain Monte Carlo \(MCMC\) fit](#).

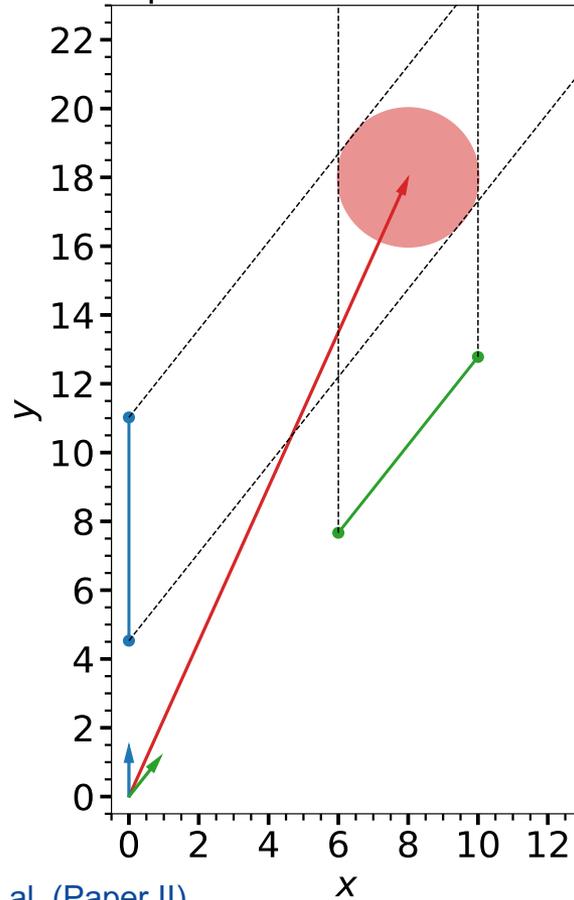
OVERLAP BETWEEN DATA COMPONENTS

Separation without correlation



Rapetti et al. (Paper II)

Separation with correlation



Schematic representation: overlap between signal and systematic modes increases the uncertainties of both components with respect to the **statistical noise (red circle)**.

Data (red vector), **signal (blue)** and **systematics (green)** basis vectors vectors on the origin.

The **blue** and **green** intervals are the 1σ uncertainties on the signal and foreground.

The signal uncertainty is computed by projecting the noise ellipse parallel to the foreground basis vector onto the line defined by the signal basis vector and vice versa.

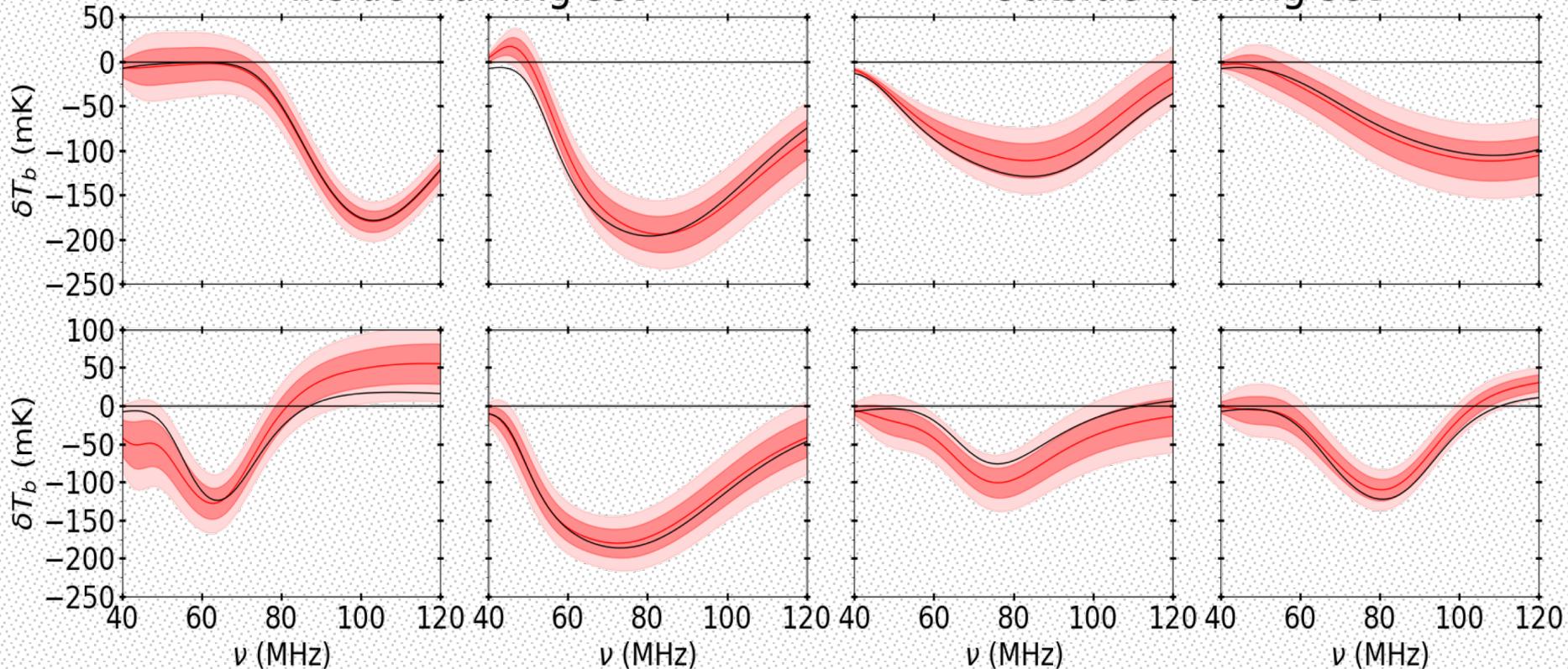
Left: **Minimum** uncertainties for each component (**noise level**) by using **orthogonal** modes. Right: **Larger** uncertainties due to **overlap**.

LINEAR SIGNAL EXTRACTION WITH PYLINEX

Tauscher et al. (2018)

Inside training set

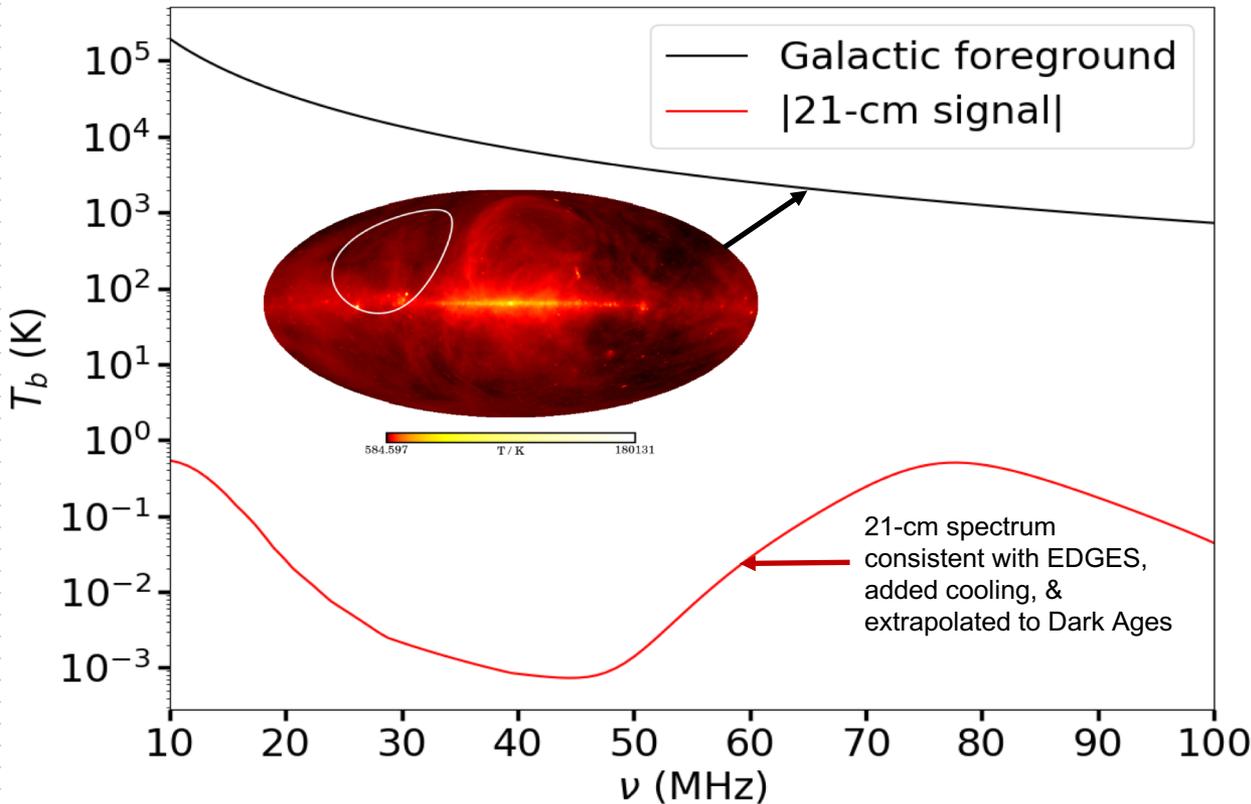
Outside training set



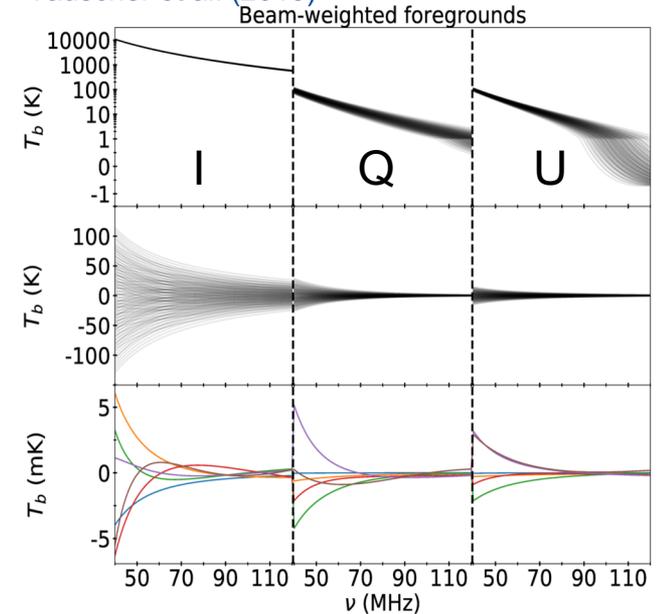
Signal Linear Estimates from SVD eigenmodes. **Black** curves: Input signals. **Red** curves: Signal estimates. **Red** bands: posterior 68/95% confidence intervals. Left: 4 input signals from the *ares* set. Right: 4 from the *tanh* set (e.g. Harker et al. 2016).

Find the code `pylinex` in this link: <https://bitbucket.org/ktausch/pylinex>

CHALLENGES OF GLOBAL 21-CM OBSERVATIONS



Tauscher et al. (2018)



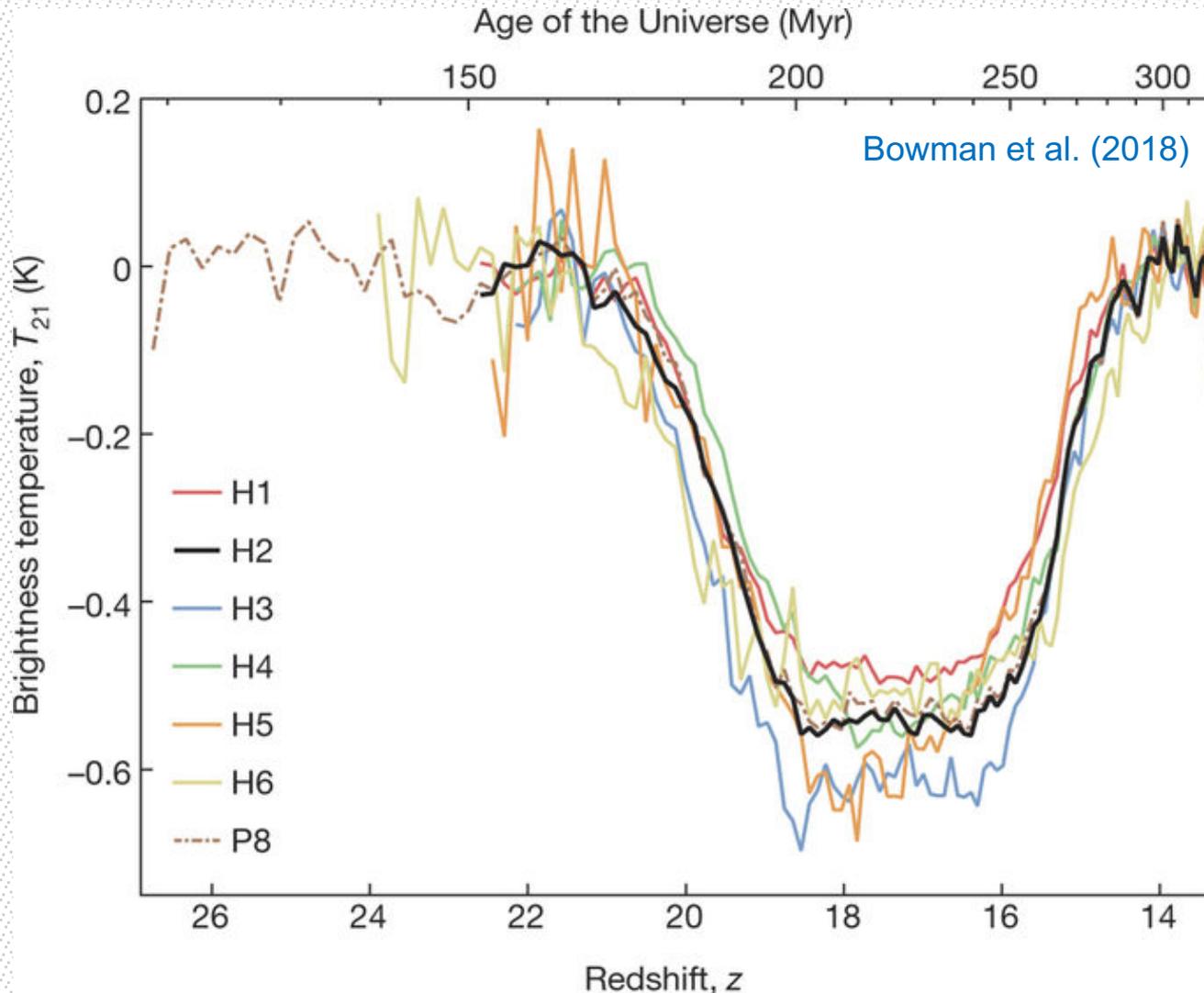
Foreground Characteristics

- Spectrally smooth
- Spatial structure
- Polarized

Signal Characteristics

- Spectral structure
- Spatially isotropic
- Unpolarized

MOTIVATION FOR MODELS: OBSERVATIONAL MEASUREMENTS

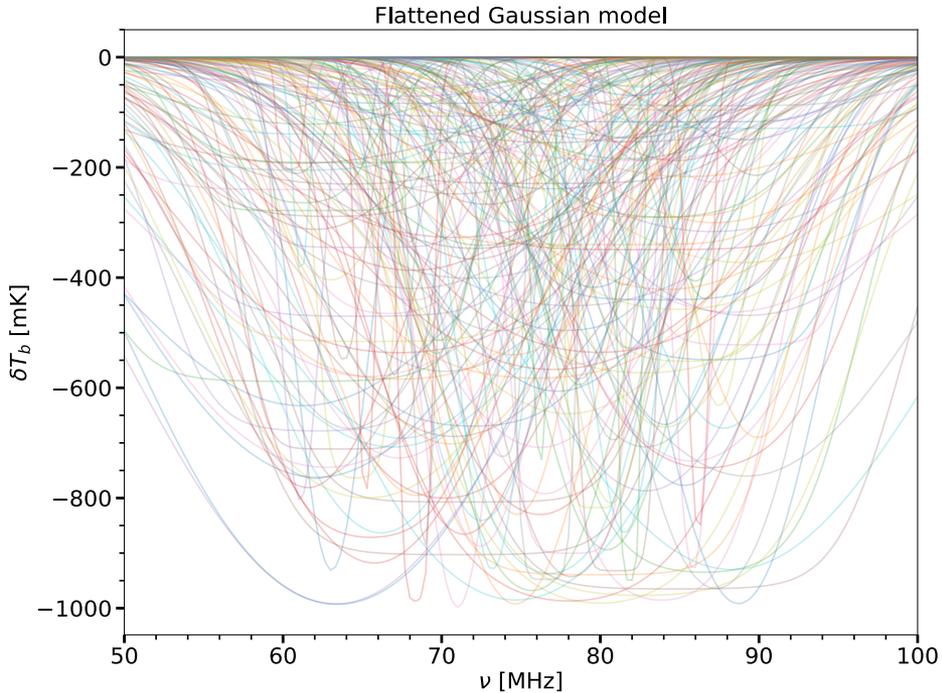
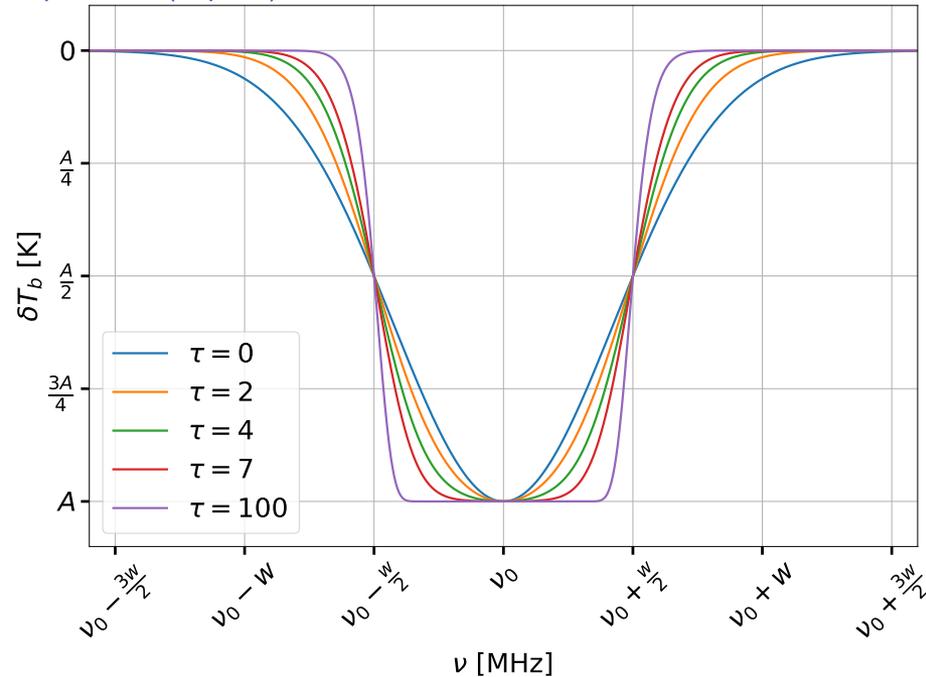


EDGES

measured a **78 MHz absorption profile** at a frequency consistent with those expected for a Cosmic Dawn signal in the global 21-cm spectrum **using a flattened Gaussian model.**

GLOBAL 21-CM MODELS: FLATTENED GAUSSIAN

Rapetti et al. (Paper II)



The flattened Gaussian model depends on parameters: A (amplitude), ν_0 (central freq.), w (FWHM), and τ (flattening). The first three shift and scale the signal while τ (at constant w) determines how long around ν_0 the signal stays near its maximum depth.

Sample of 200 curves from the training set for the flattened Gaussian model.

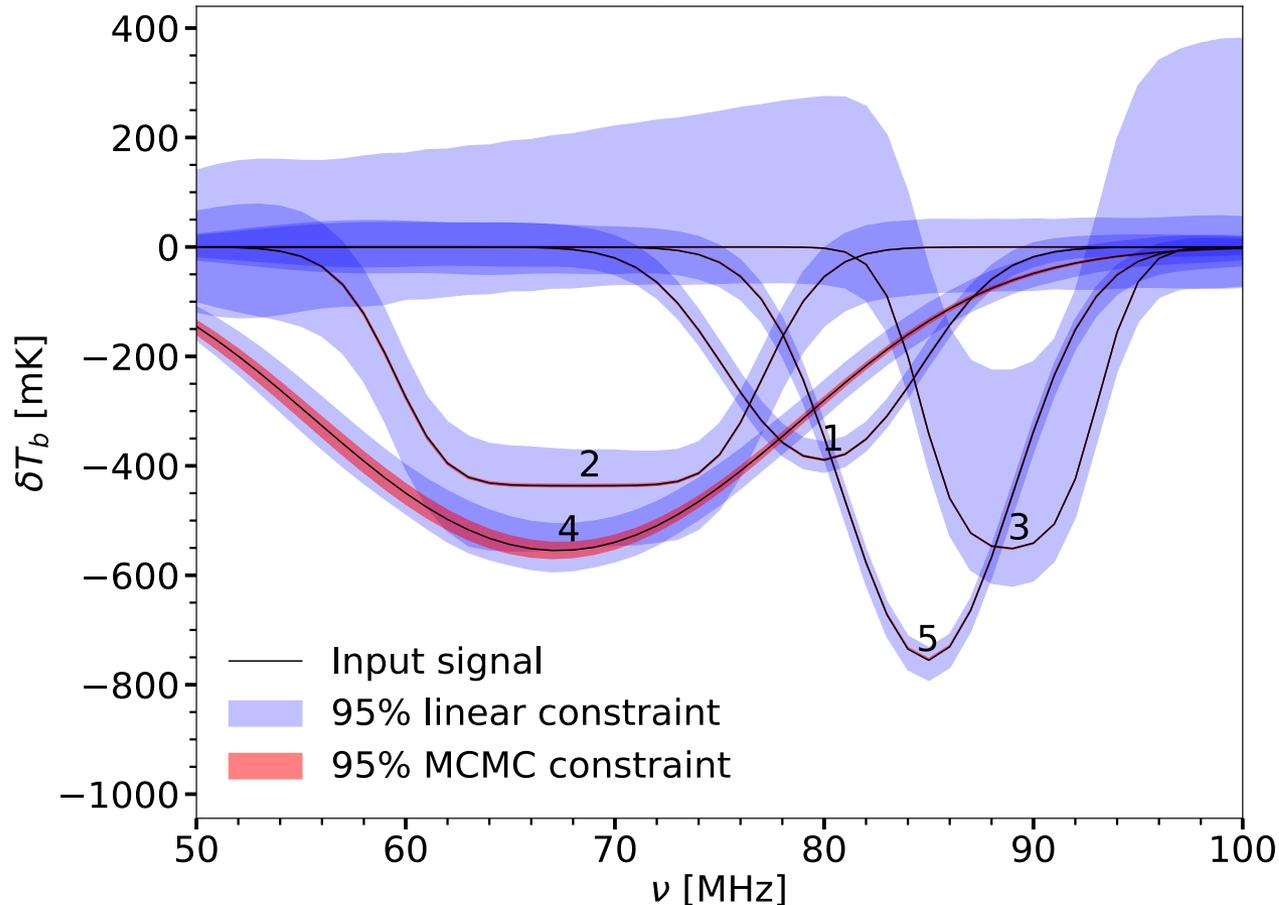
A uniform (-1, -0.1) K
 ν_0 uniform (60, 90) MHz
 w uniform (1, 30) MHz
 τ exponential (1)

MCMC CONDITIONAL FIT OF 21-CM SIGNAL MODEL OVER SVD FOREGROUND TERMS

- Our conditional MCMC **marginalizes** over the **SVD foreground modes** at each step, while **efficiently** exploring only the physical parameter space.
- This calculation is **exact (not an approximation)** and allows for the natural separation of **linear nuisance parameters without a need for a parametric model** and nonlinear signal parameters to be MCMC sampled.
- This properly accounts for **overlaps** between the **signal and the systematics** (beam-weighted foreground, receiver, etc.).
- **Other experiments** in physics/astrophysics/cosmology could benefit from our current set of novel solutions.

FLATTENED GAUSSIAN MODEL: LINEAR AND MCMC SPECTRAL CONSTRAINTS

Rapetti et al. (Paper II) Flattened Gaussian model constraints



Pipeline spectral constraints for five random flattened Gaussian cases successfully recovered:

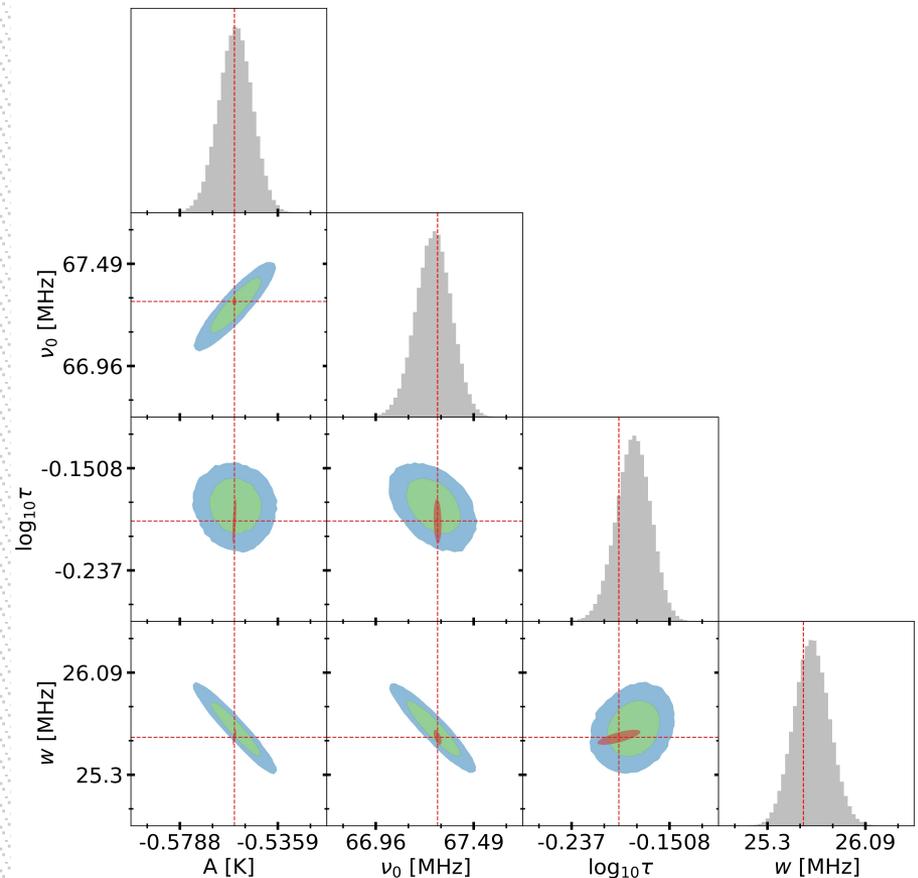
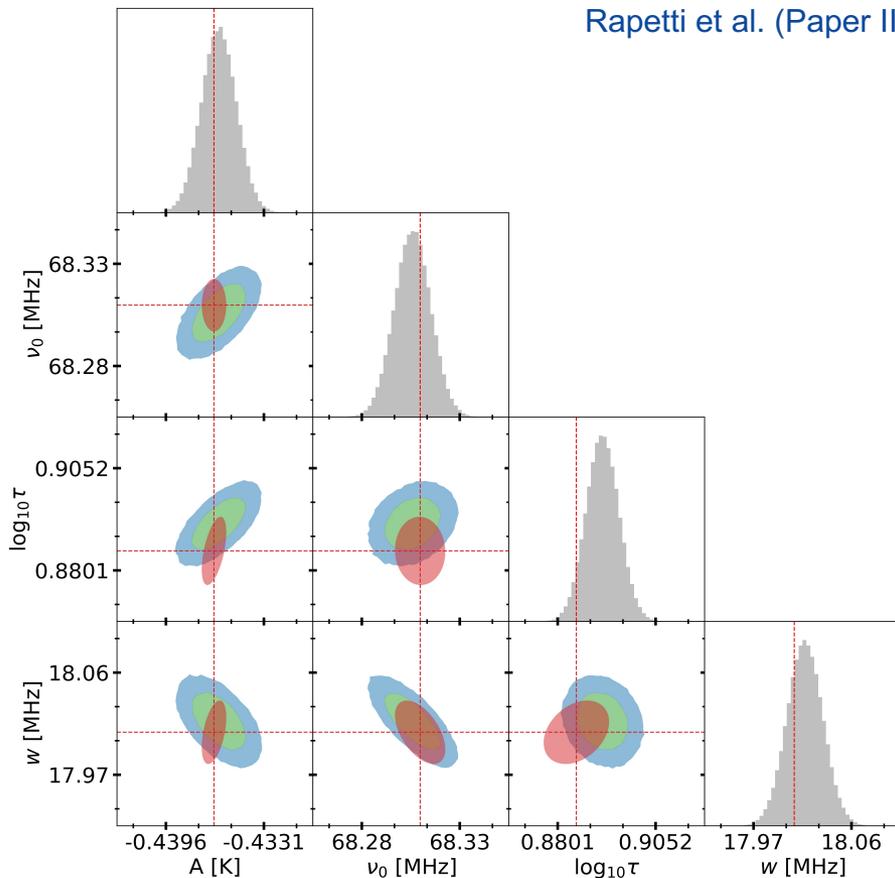
Blue bands: 95% confidence intervals from the **linear fit**, with SVD signal and foreground modes.

Red bands: 95% confidence intervals from the **MCMC fit**, with the **full nonlinear signal model** and SVD foreground modes.

Note that for the linear fit, the 95% confidence intervals correspond to 8.75σ .

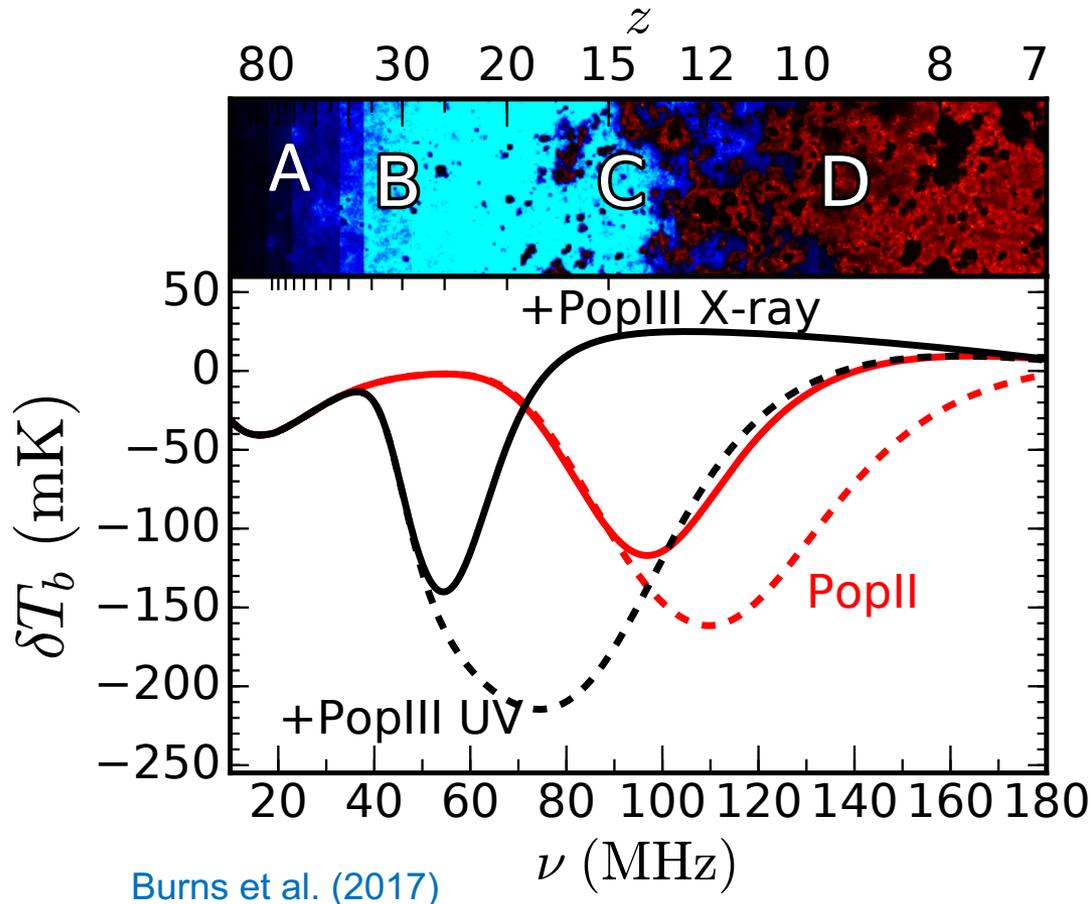
FLATTENED GAUSSIAN MODEL: FULL MCMC PARAMETER CONSTRAINTS

Rapetti et al. (Paper II)



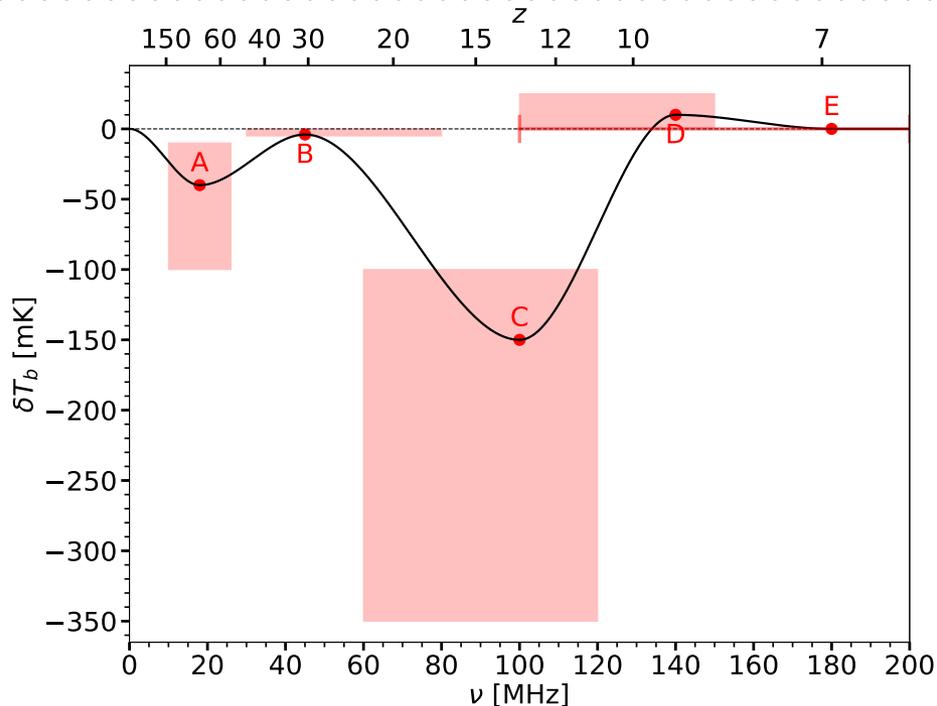
1D (gray) and 2D (68/95%) MCMC posterior parameter constraints. The red, dashed lines mark the input parameters. The left (right) plot corresponds to case 2 (4) before. The red contours represent 95% errors for statistical noise alone. In case 4, systematics clearly play a larger role than in case 2.

MOTIVATION FOR MODELS: HYDROGEN COSMOLOGY THEORY



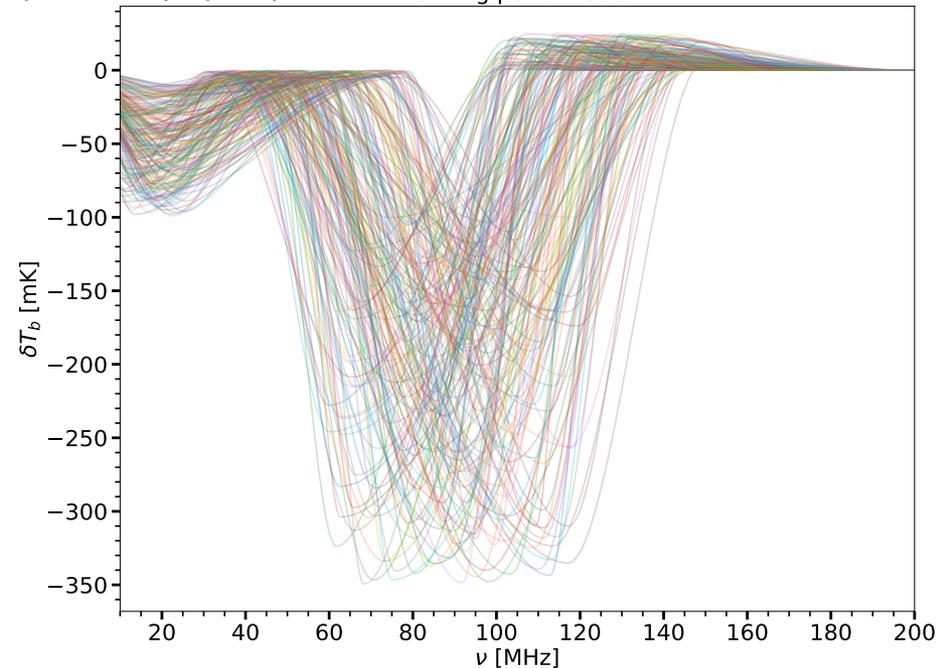
- dT_b is a combination of temperatures: T_S spin, T_k kinetic, T_α Lyman-a, T_γ background (CMB).
- **A: Expansion** recouples $T_S \rightarrow T_\gamma$
- **B: First stars** Ly-a emission couples back $T_S \rightarrow T_k$
- **C: Heating sources** including initial **black hole** accretion drive $T_k \rightarrow T_\gamma$
- **D: Reionization** removes signal ($x_{HI} \rightarrow 0$).

GLOBAL 21-CM MODELS: TURNING POINT



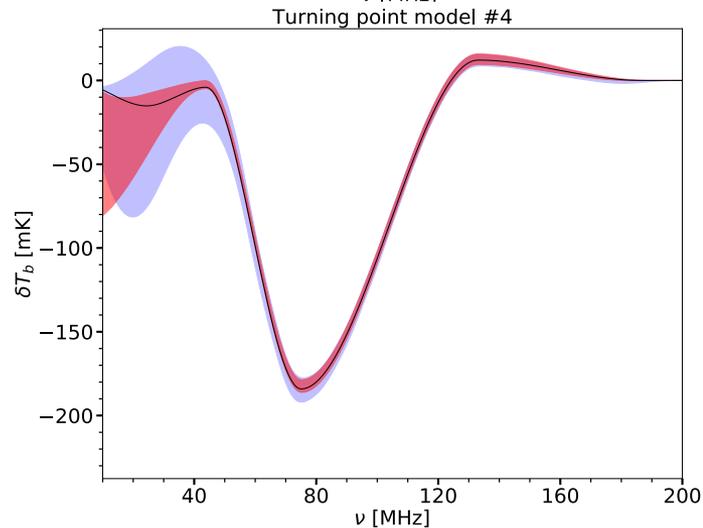
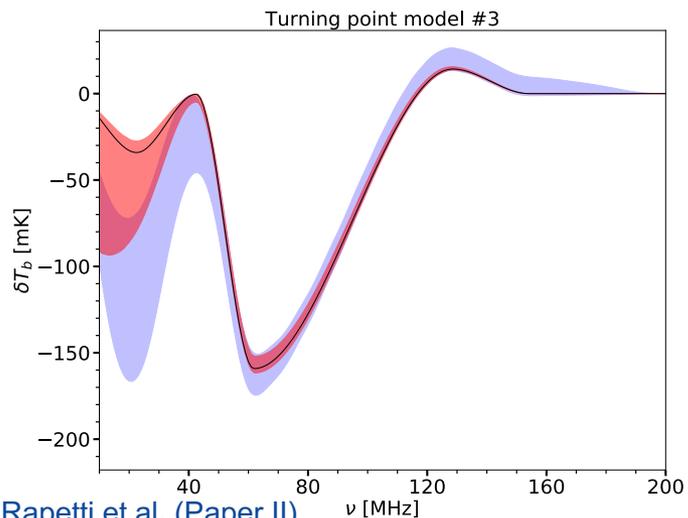
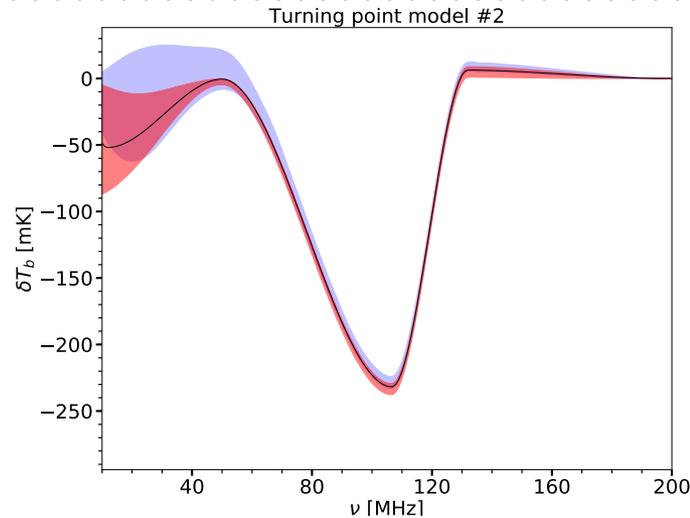
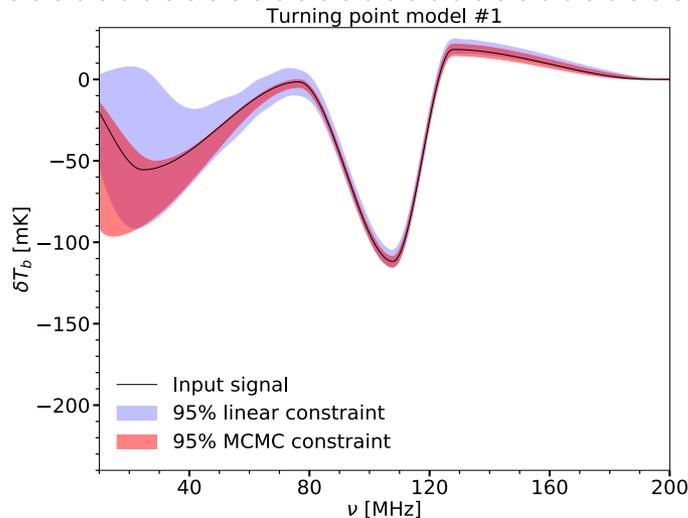
Rapetti et al. (Paper II)

Turning point model



- **Black line:** typical model with movable **red dots** defining a spline interpolation, for which δT_b and its derivative are also fixed to 0 at $\nu = 0$.
- Broadly speaking, **A** represents **Dark Ages**, **B** **Cosmic Dawn**, **C-D** the epoch of heating & **D-E** the epoch of reionization.
- The **red filled regions** around A-D show the allowed positions of the points to build the training set (200 samples, right panel). The **red horizontal line** with vertical bars on its ends marks the same for E.
- Adjacent frequencies are forced to be at least 10 MHz apart.

TURNING POINT MODEL: LINEAR AND MCMC SPECTRAL CONSTRAINTS



Pipeline spectral constraints for four random cases successfully recovered:

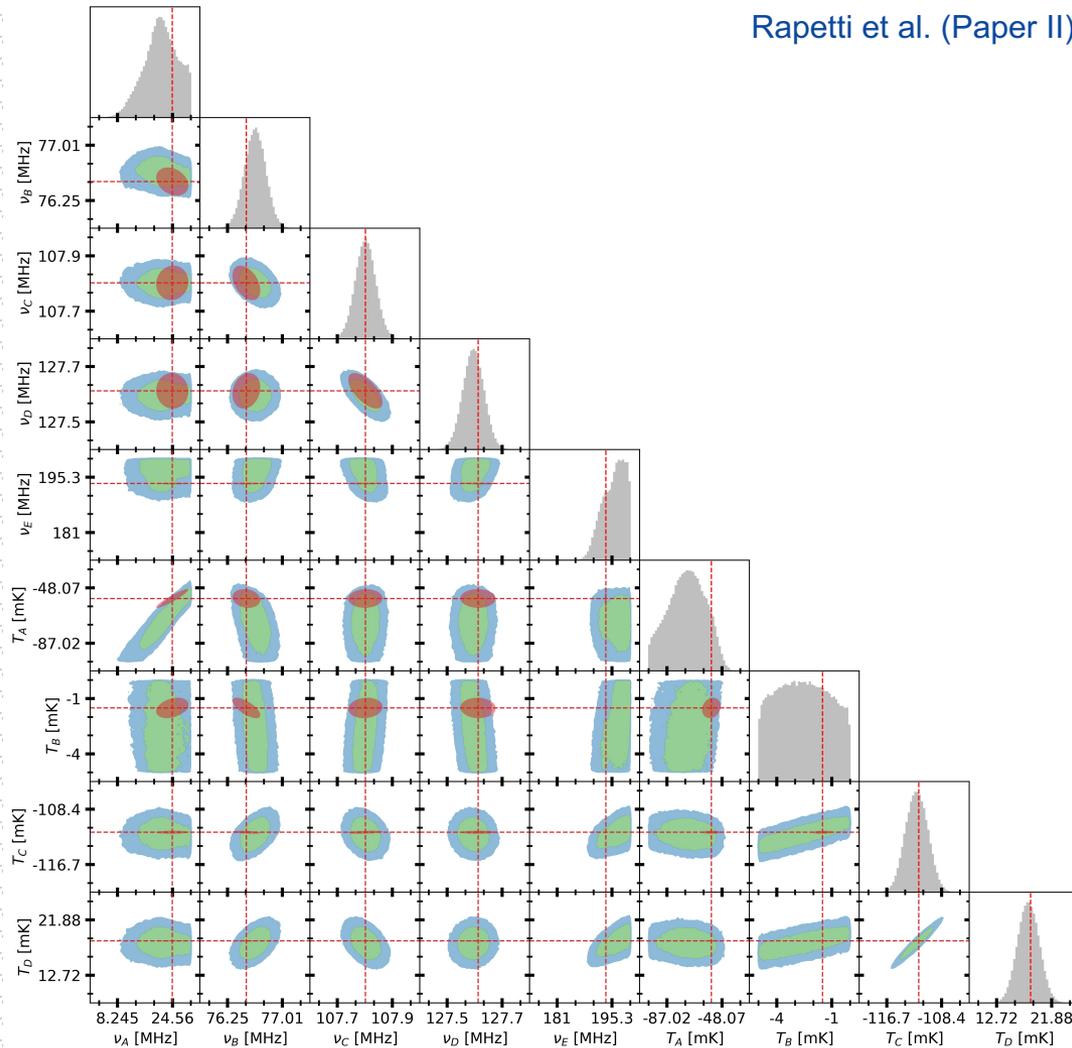
Blue bands: 95% confidence intervals from the linear fit, with SVD signal and foreground modes.

Red bands: 95% confidence intervals from the MCMC fit, with the full nonlinear signal model and SVD foreground modes.

For the linear fit, the 95% intervals correspond to 2.5σ .

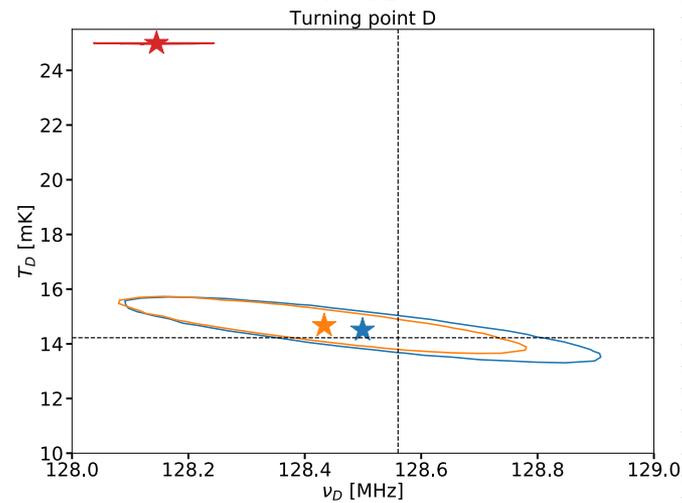
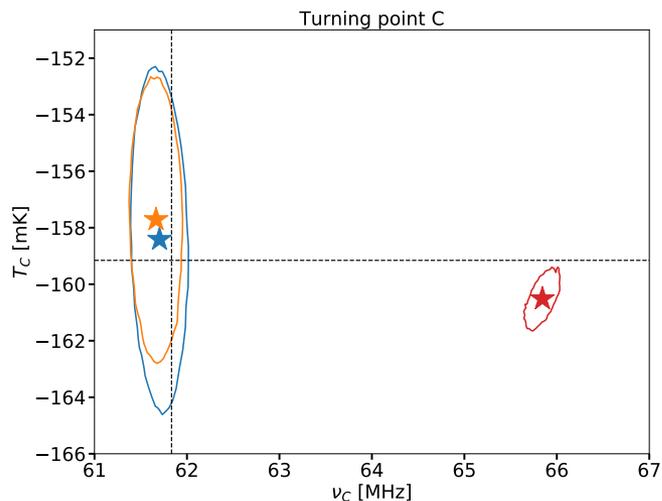
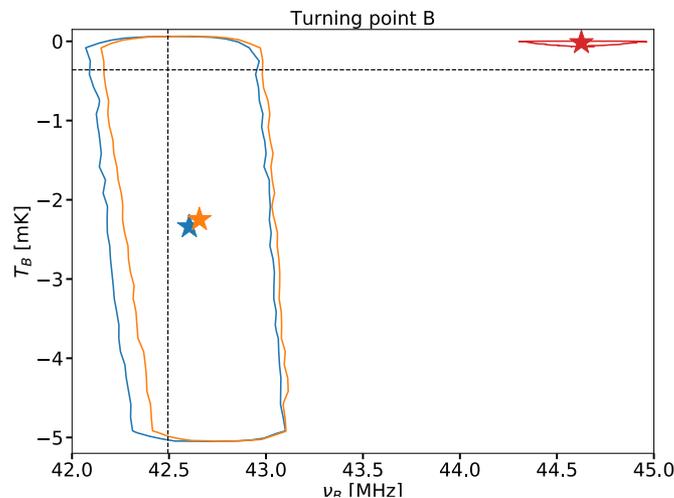
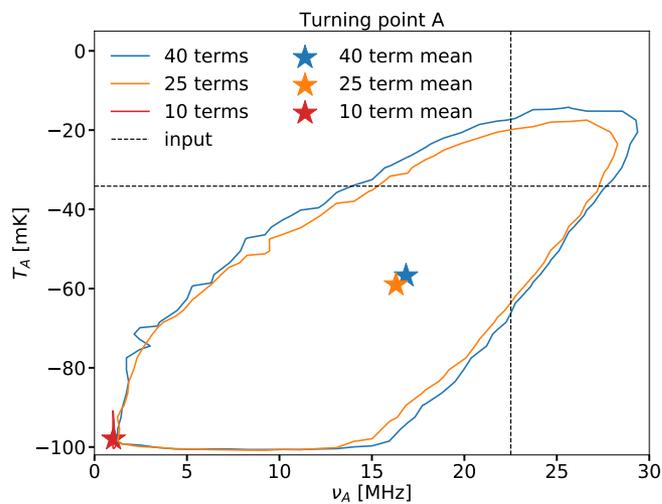
TURNING POINT MODEL: FULL MCMC PARAMETER CONSTRAINTS

Rapetti et al. (Paper II)



- 1D (gray) and 2D (68/95%) MCMC posterior parameter constraints.
- The red, dashed lines mark the input parameters.
- The red ellipses represent 95% confidence contours when only the statistical noise (Fisher-matrix estimated) obscures the signal.
- All intervals assume 800 hours of integration.
- Note e.g. that the temperature of turning point B, allowed to only vary from -5 to 0 mK, is not constrained within the prior, and the temperature C is well constrained.

TURNING POINT MODEL: NUMBER OF MARGINALIZED FOREGROUND PARAMETERS



- 95% constraints on A-D if marginalizing over **10 (red)**, **25 (orange)** & **40 (blue)** terms in the MCMC fit.

- The dashed lines indicate the input parameters.

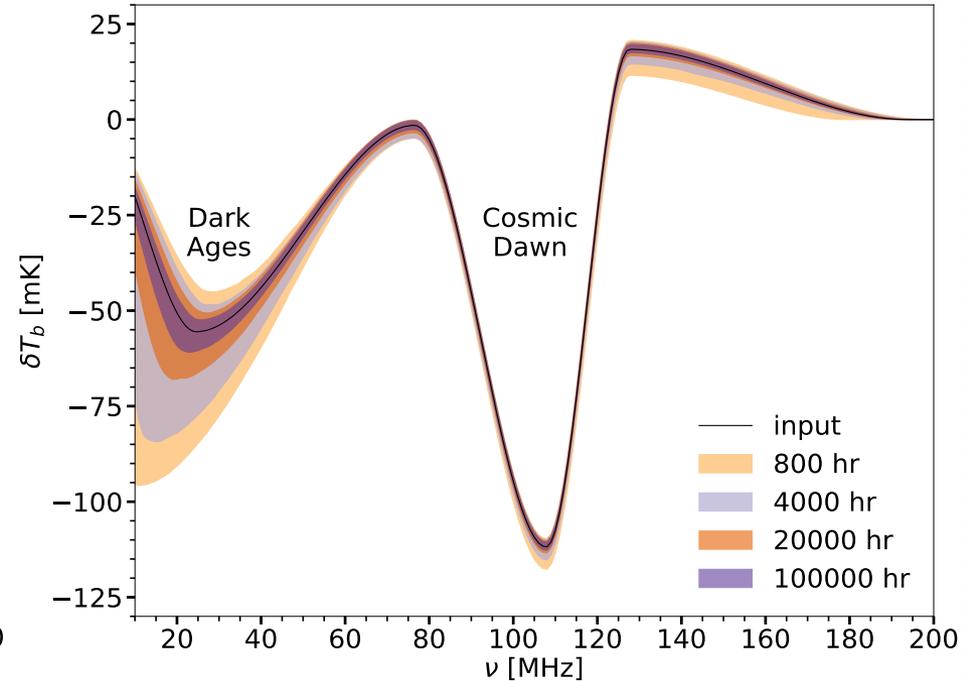
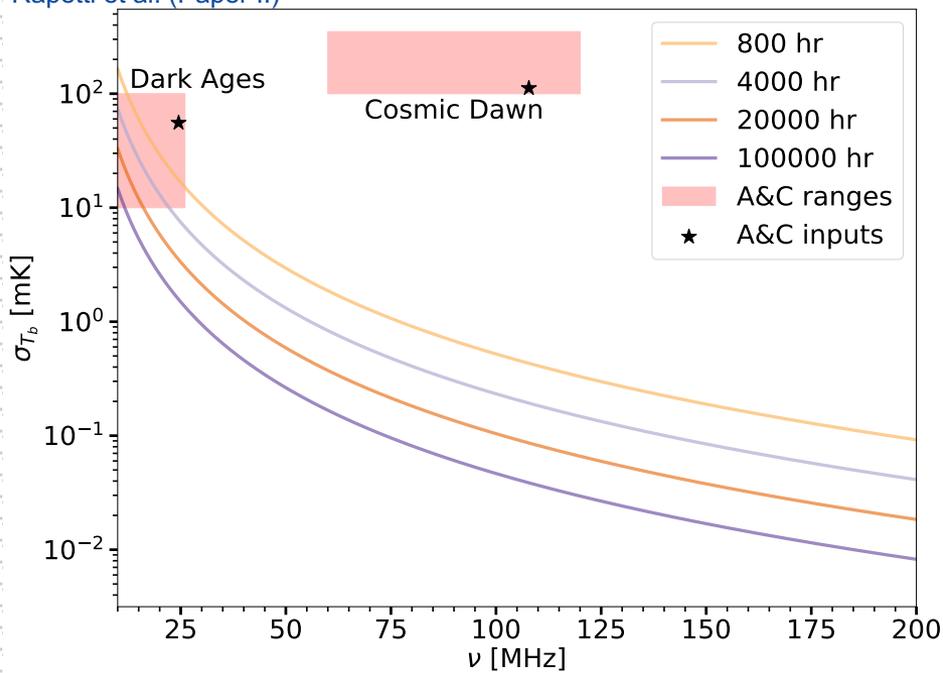
- **10 terms** are not sufficient to explain the foreground (in the linear fit, 24 were chosen), so the signal is biased with spuriously tight constraints.

- For **25 terms**, the signal is recovered with realistic errors.

- For **40 terms**, there is no qualitative change thanks to the use of foreground priors.

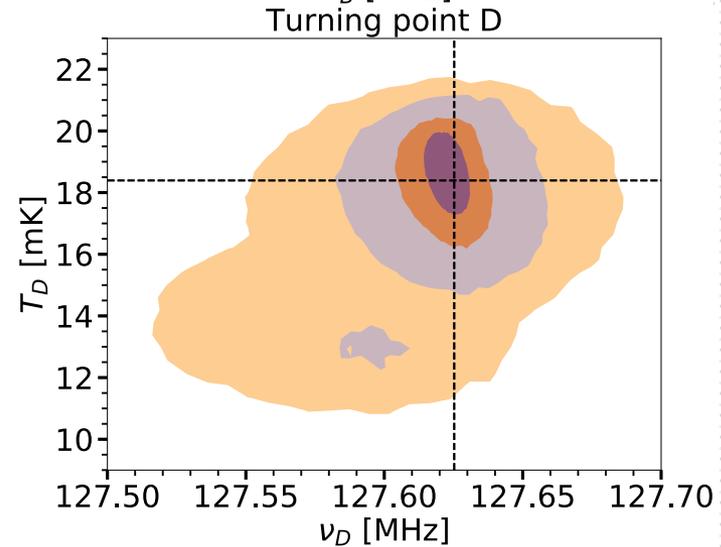
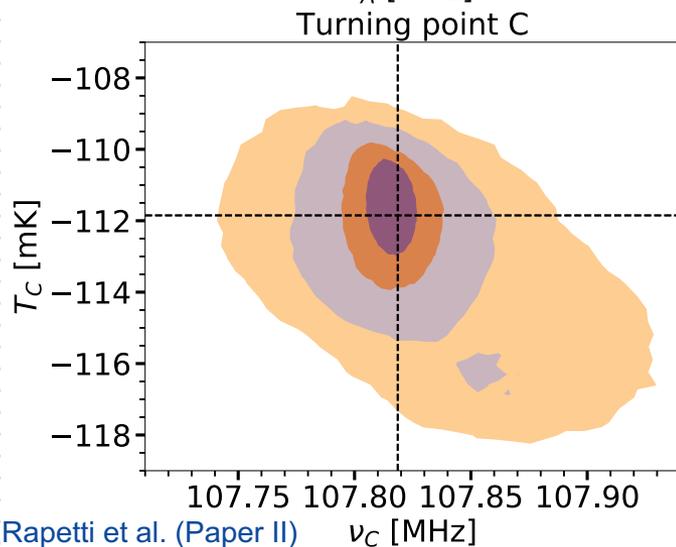
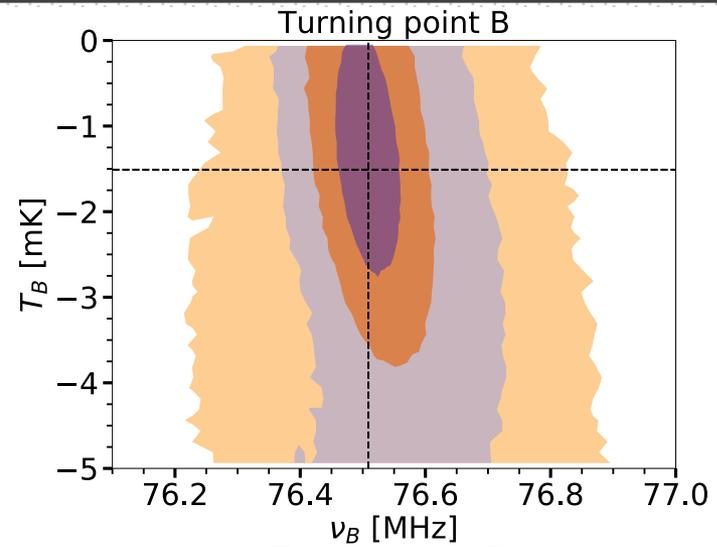
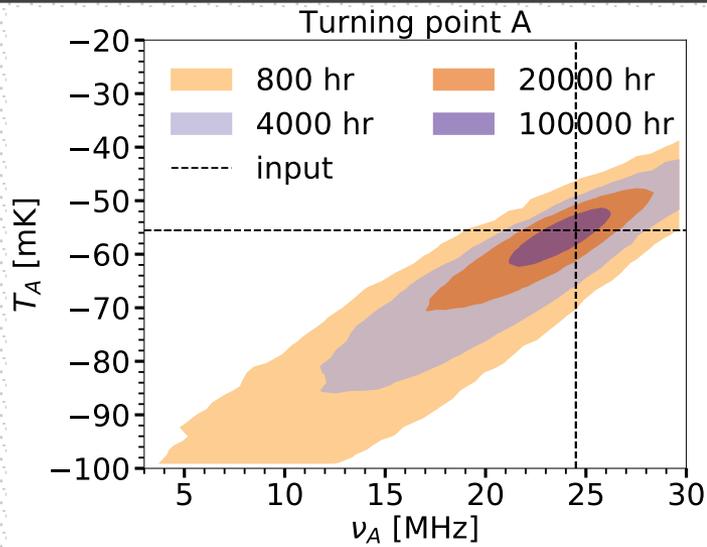
TURNING POINT MODEL: INCREASING THE INTEGRATION TIME

Rapetti et al. (Paper II)



- Left: 1σ noise levels for the factor of 5 increases of integration times from our reference of 800 hours. The **red rectangles** indicate the **allowed values** for turning points A (Dark Ages) and C (Cosmic Dawn) and the **black stars** the **input values**.
- Right : Full uncertainties in frequency space for four different integration times with the same random seed for noise generation.

TURNING POINT MODEL: INCREASING THE INTEGRATION TIME



Rapetti et al. (Paper II)

SUMMARY



- We employ a **linear, fast, analytic methodology** to separate the global 21-cm signal from systematics, with which it can have large overlaps, to estimate the **starting point of a full MCMC search** of any selected **physical signal model**.
- We utilize the linear **SVD foreground terms** to properly and efficiently (in terms of convergence) account for this modeling by **marginalizing over these generally large number of parameters** at each step of our MCMC signal calculation.
- We **test our novel pipeline** on two physically motivated signal models, **flattened Gaussian** (observationally based) and **turning point** (theoretical), and **successfully recover the input parameters** for multiple random cases.
- **EDGES, CTP, DAPPER, FARSIDE, etc.** measurements should benefit from this **statistically rigorous, robust pipeline** which is able to extract the 21-cm signal while modeling the systematics using **detailed training sets** from theory, simulations and observations.