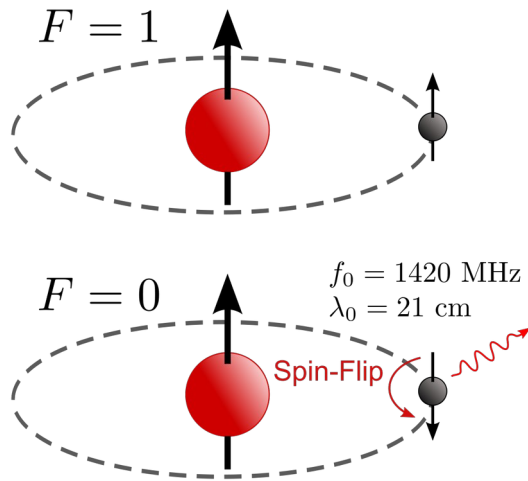




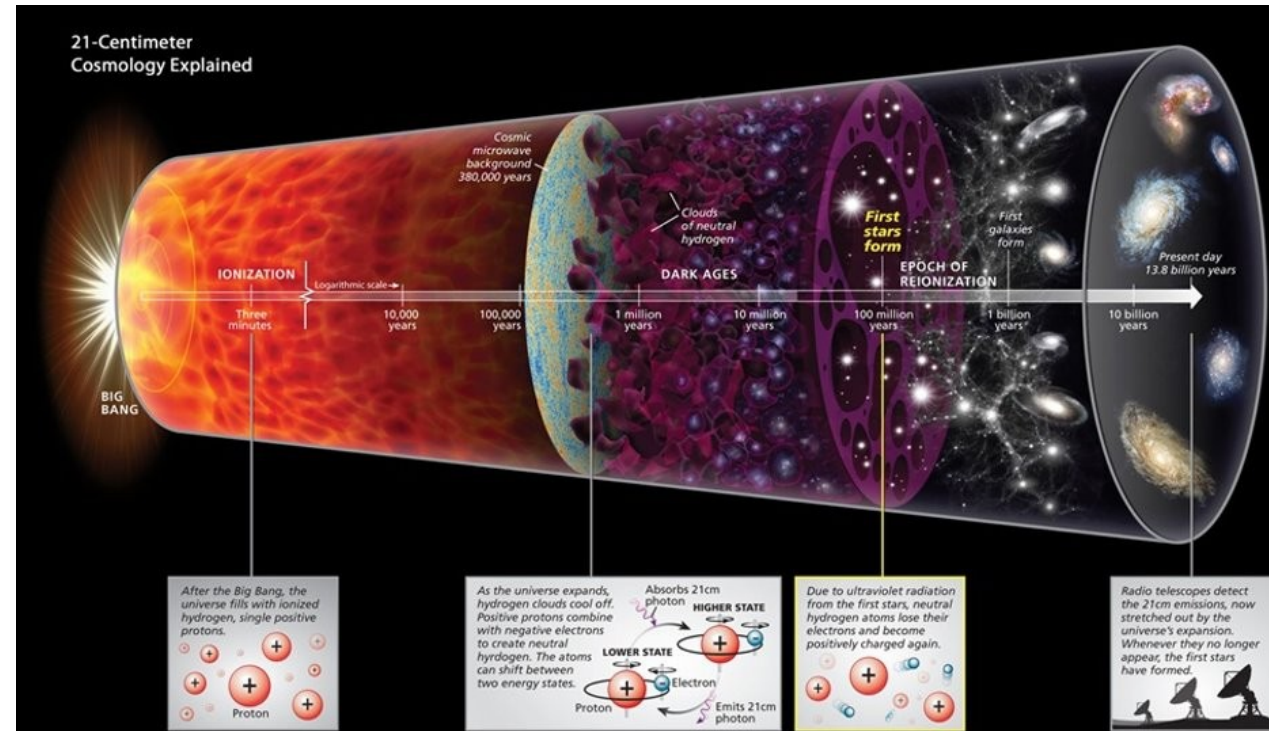
Using Bayesian
data analysis to
model
systematics and
inform experiment
design in Global
21cm
experiments

Dominic Anstey
On behalf of the REACH
collaboration

Global 21cm Experiments



https://en.wikipedia.org/wiki/Hydrogen_line



Roan Kelly, Discover Magazine

Bayesian Data Analysis

$$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$

$$P(\theta_{\mathcal{M}}|\mathcal{D}, \mathcal{M}) = \frac{P(\mathcal{D}|\theta_{\mathcal{M}}, \mathcal{M}) P(\theta_{\mathcal{M}}|\mathcal{M})}{P(\mathcal{D}|\mathcal{M})}$$

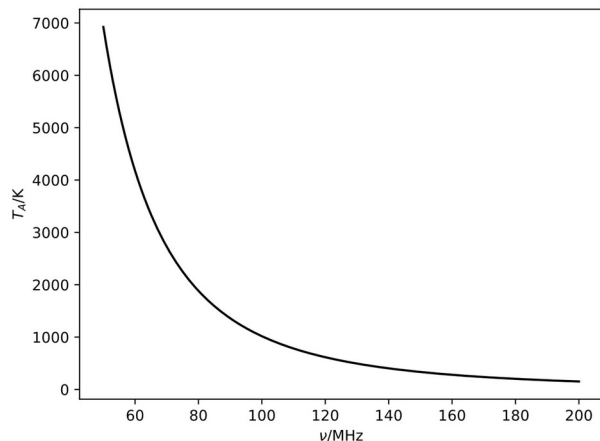
$$\mathcal{P} = \frac{\mathcal{L}\pi}{\mathcal{Z}}$$

$$P(\mathcal{M}|\mathcal{D}) = \frac{P(\mathcal{D}|\mathcal{M}) P(\mathcal{M})}{P(\mathcal{D})} = \mathcal{Z} \frac{P(\mathcal{M})}{P(\mathcal{D})}$$

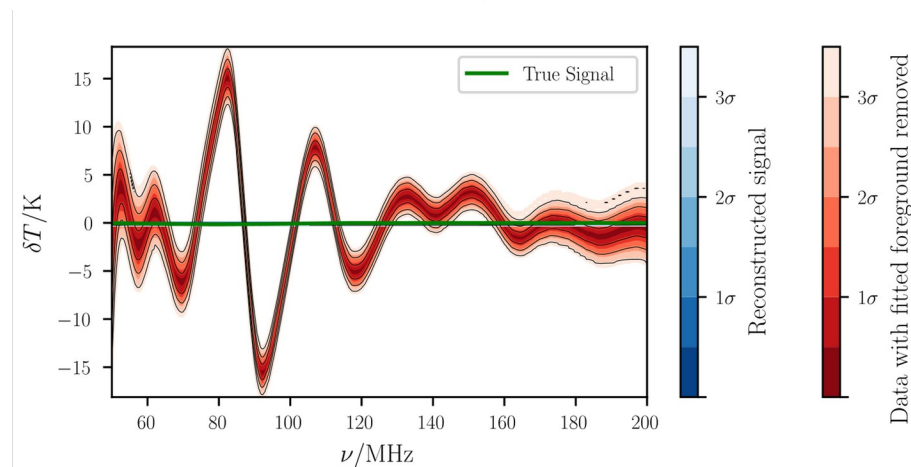
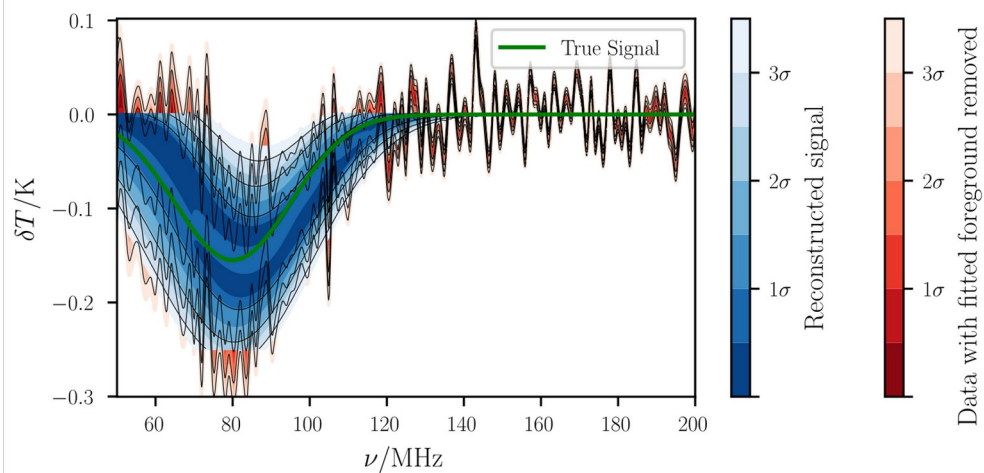
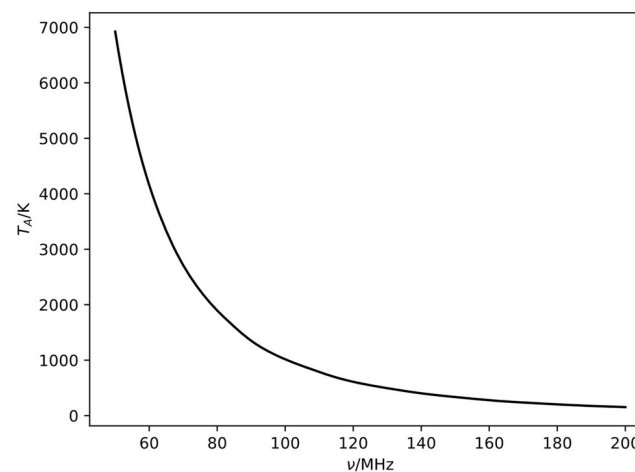


Radio Foregrounds and Chromaticity

Achromatic antenna



Chromatic antenna



REACH Data Analysis Pipeline

- Generate a model of the sky radio emission based on physical parameters

$$T_{\text{sky}} (\Omega, \nu, \theta_{\text{F}})$$

- Convolve this with a model of the antenna beam to produce a parameterised foreground model that includes chromatic distortions

$$T_{\text{F}} (\nu, \theta_{\text{F}}) = \frac{1}{4\pi} \int_0^{4\pi} D (\Omega, \nu) T_{\text{sky}} (\Omega, \nu, \theta_{\text{F}}) d\Omega$$

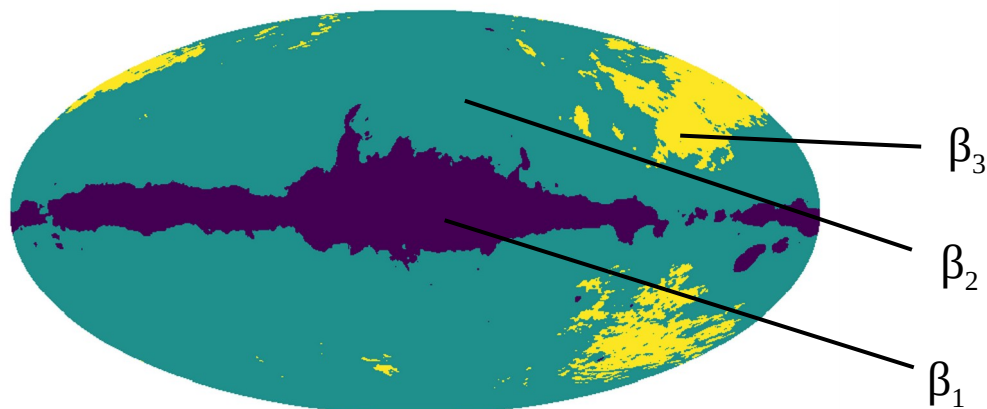
- Use this model in a Nested Sampling fit

$$\log \mathcal{L} = \sum_i \left[-\frac{1}{2} \log (2\pi\theta_\sigma) - \frac{1}{2} \left(\frac{T_{\text{data}} (\nu_i) - T_{\text{F}} (\nu_i, \theta_{\text{F}}) - T_{\text{S}} (\nu_i, \theta_{\text{S}})}{\theta_\sigma} \right)^2 \right]$$

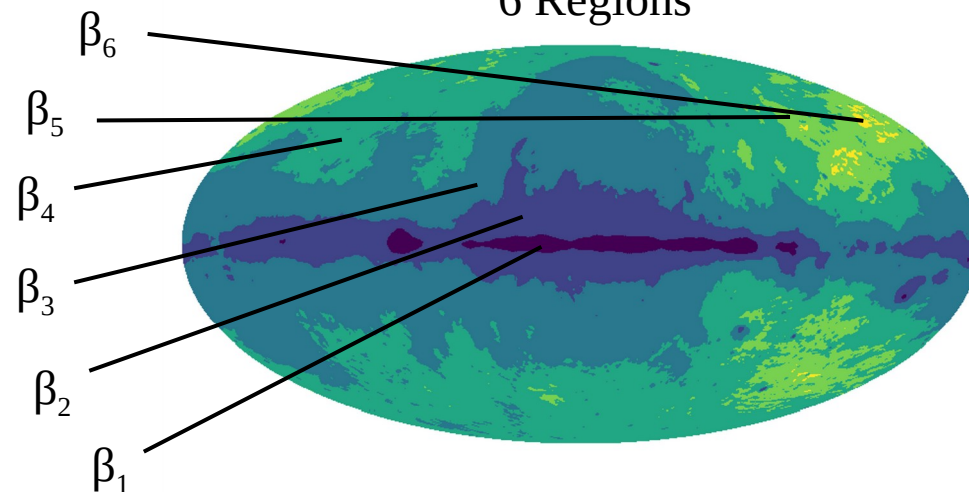
Parameterised Sky Modelling

Parameterise by spectral index

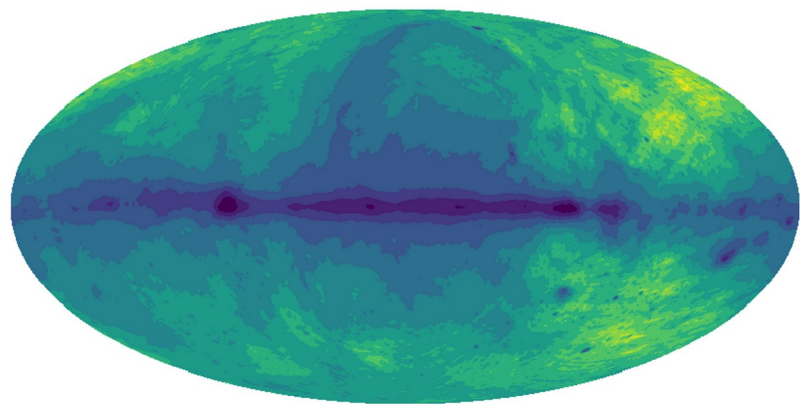
3 Regions



6 Regions

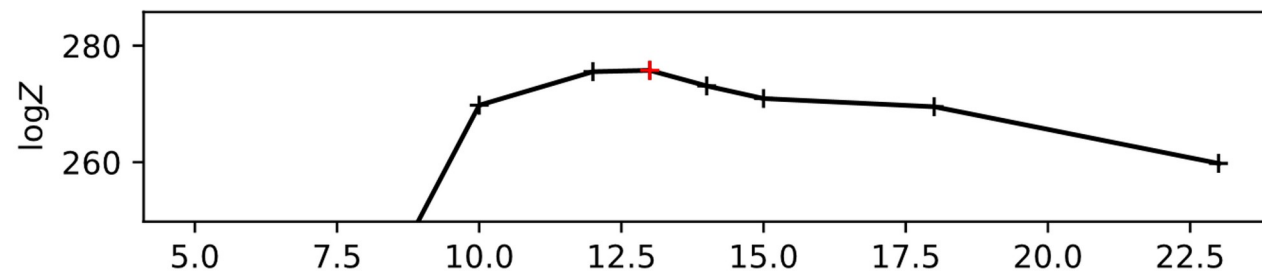
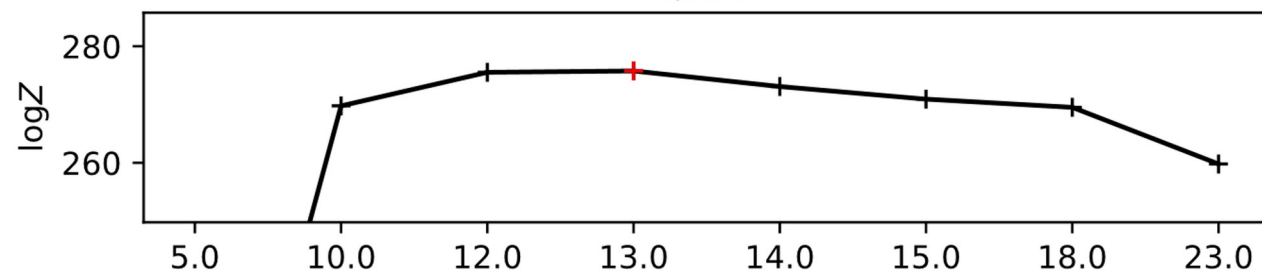
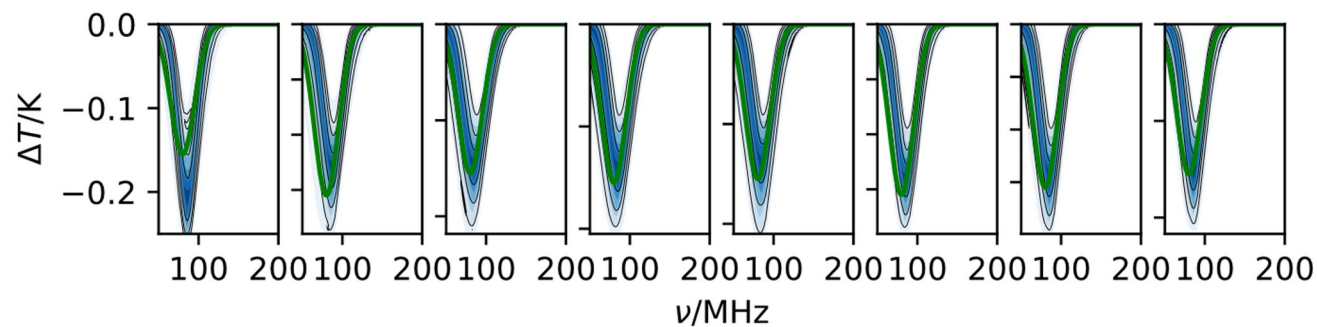
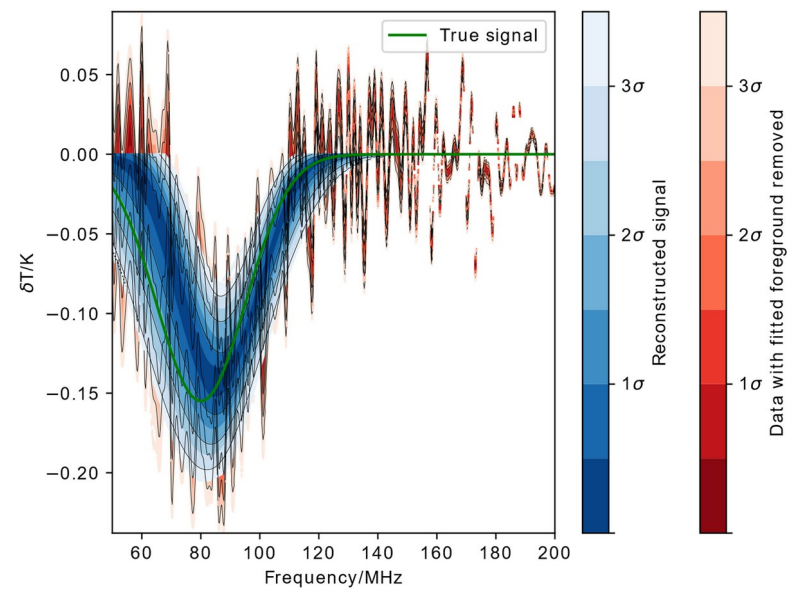
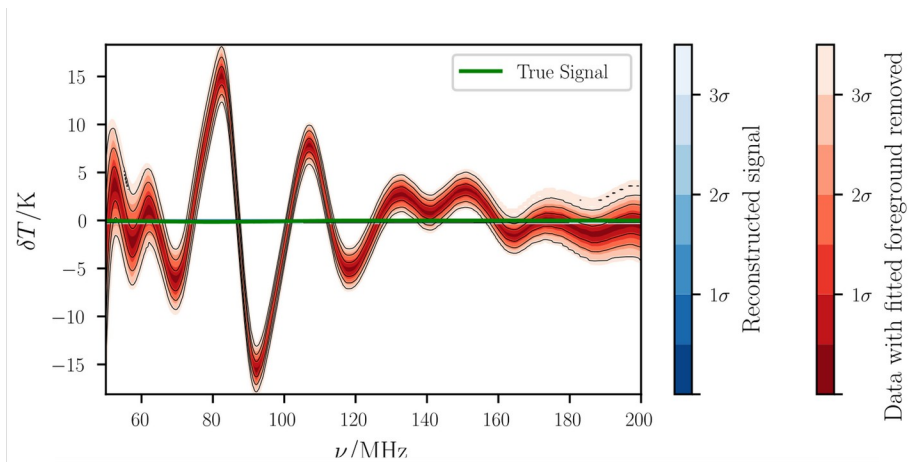


12 Regions



$$T_{\text{sky}}(\Omega, \nu) = \left[\sum_{n=1}^N M_n(\Omega) (T_{\text{base}}(\Omega) - T_{\text{CMB}}) \left(\frac{\nu}{\nu_{\text{base}}} \right)^{-\beta_n} \right] + T_{\text{CMB}}$$

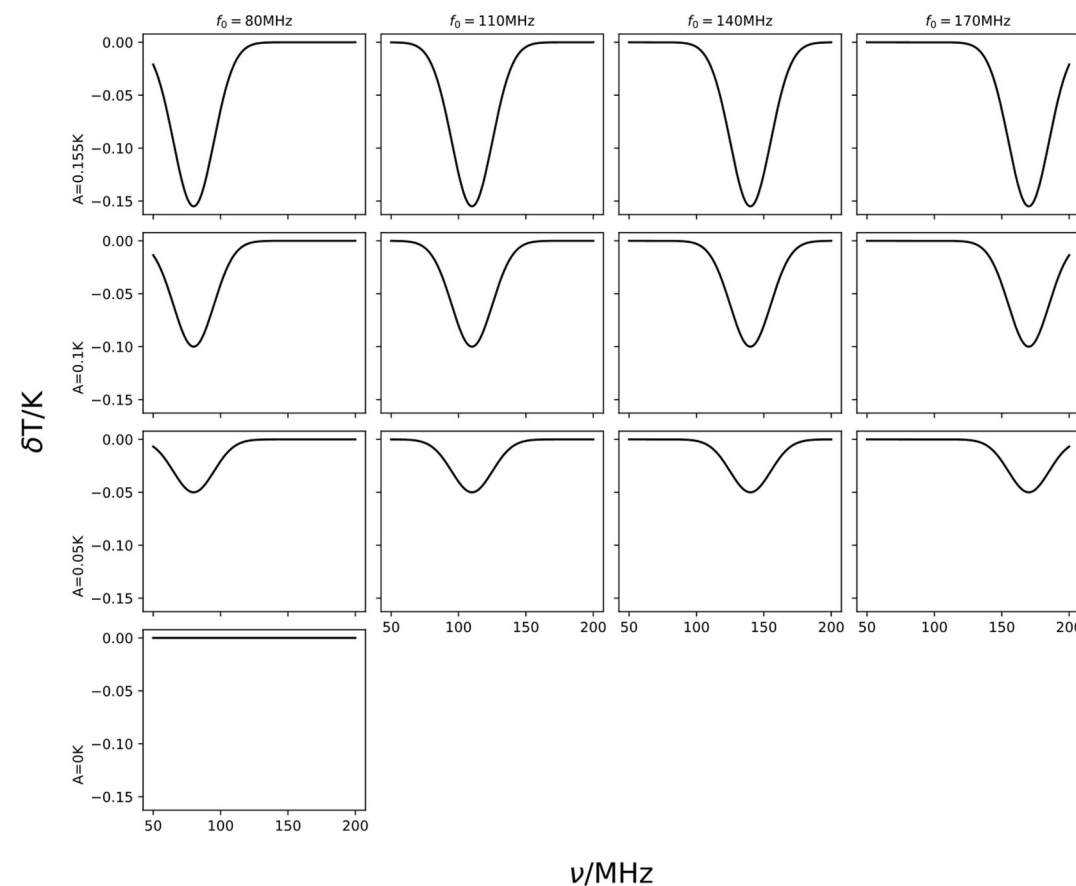
Performance



Informing Experiment Design

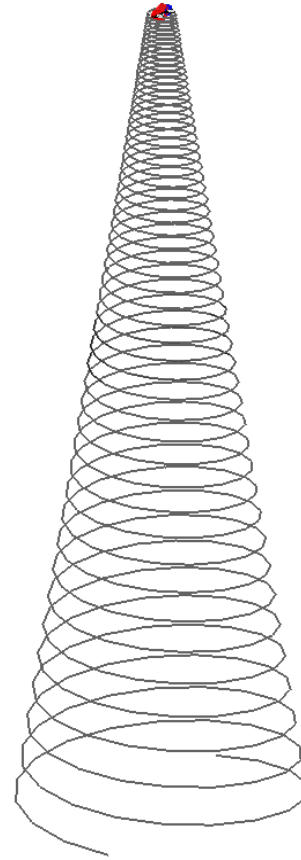
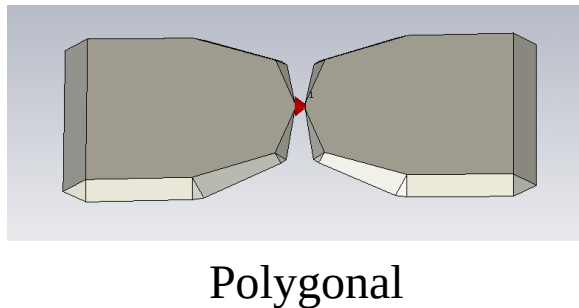
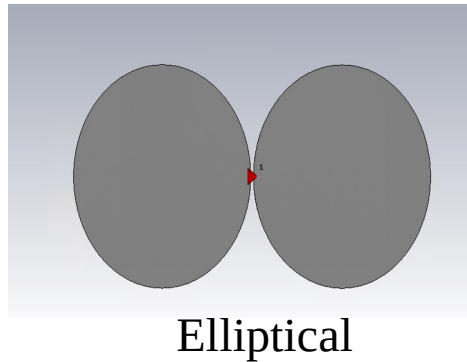
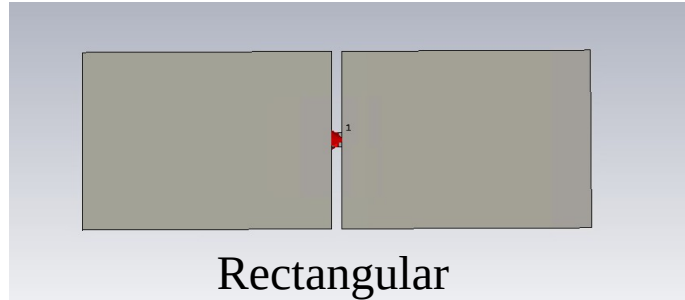
- Simulate observation data for a range of potential antenna designs
- Insert a range of mock 21cm signals into the data
- Attempt to recover each signal using the described pipeline

This allows the accuracy of each recovered signal to be quantified relative to the true signal. We can also quantify the statistical confidence with which the signal is detected.

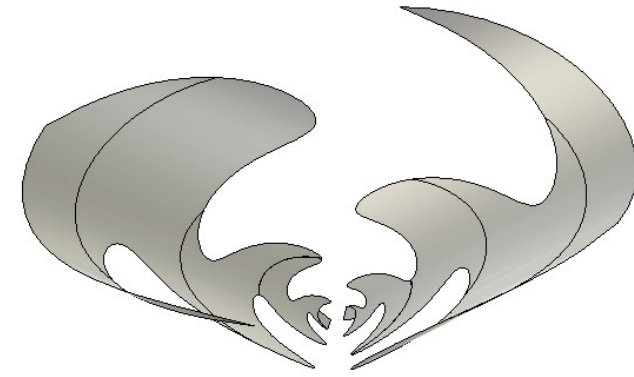


Tested Antennae

Dipoles:



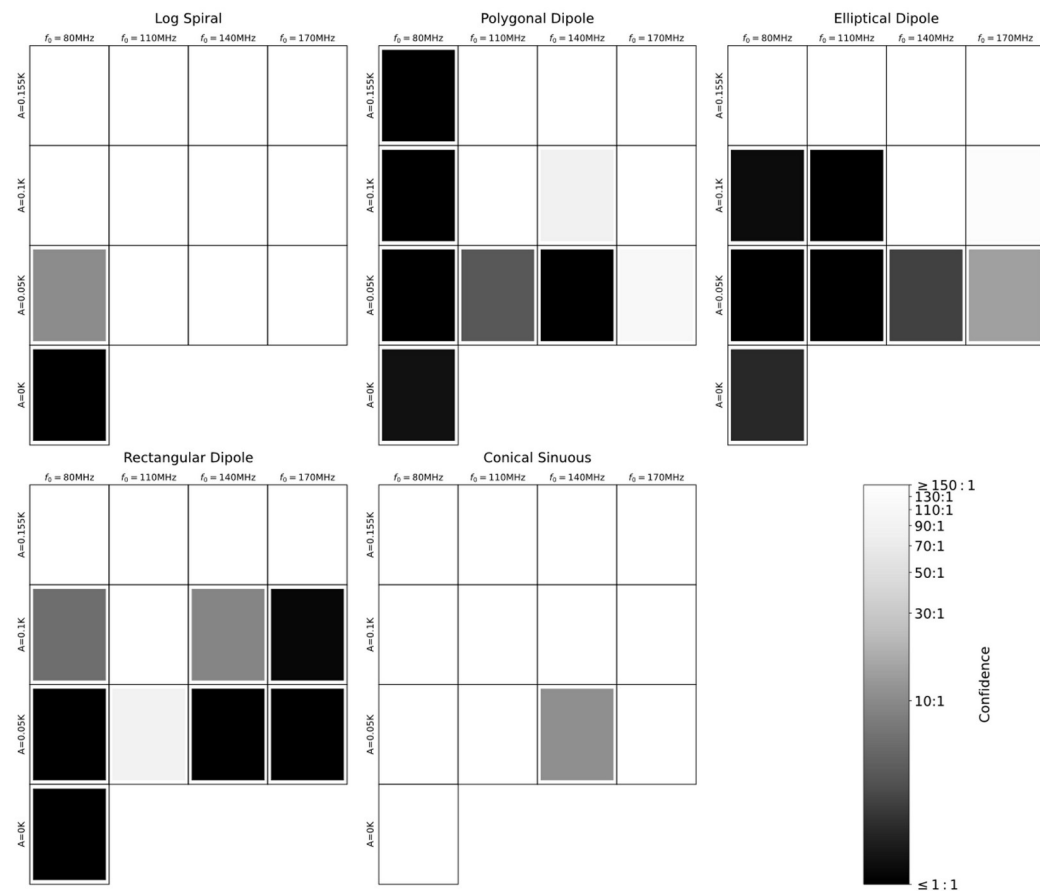
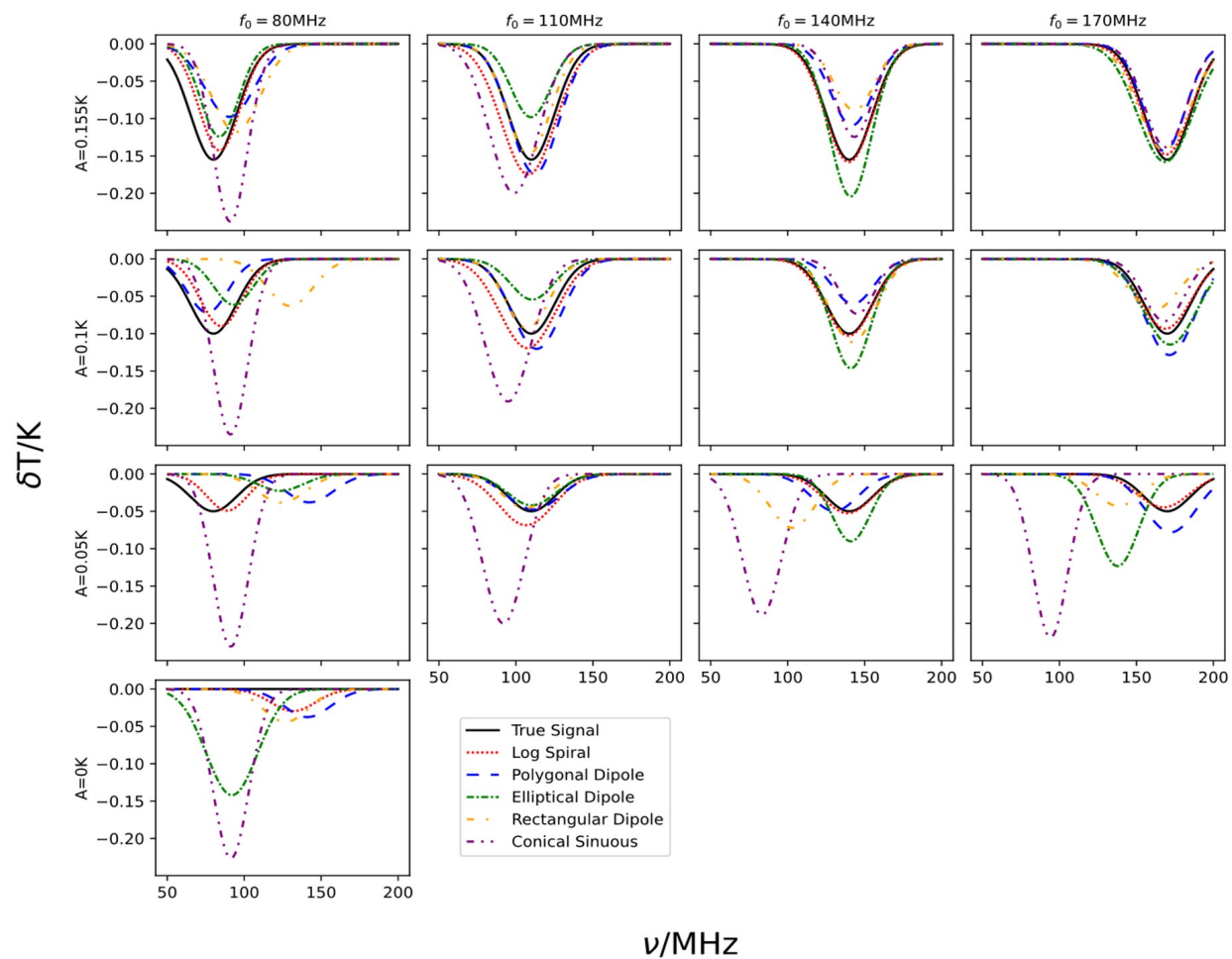
Conical
Logarithmic Spiral



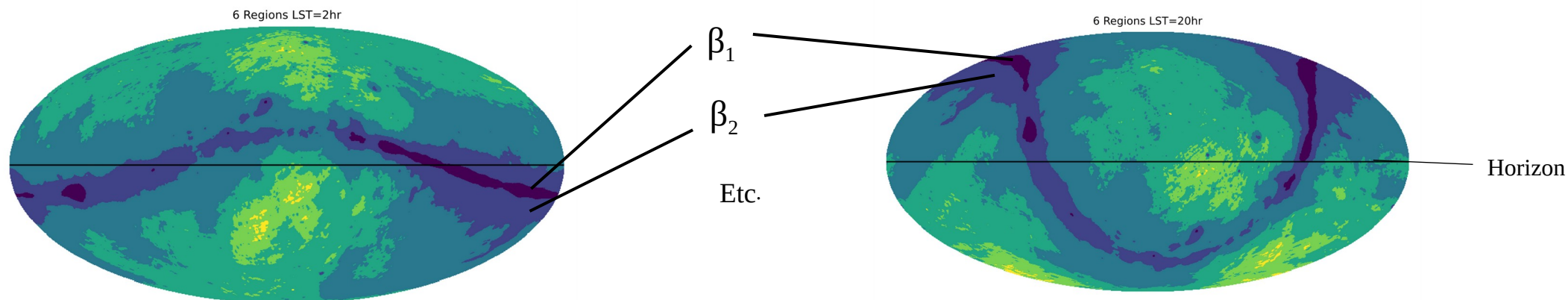
Inverse Conical
Sinuous

Antenna patterns and images
provided by John Cumner and
Quentin Guerring

Results

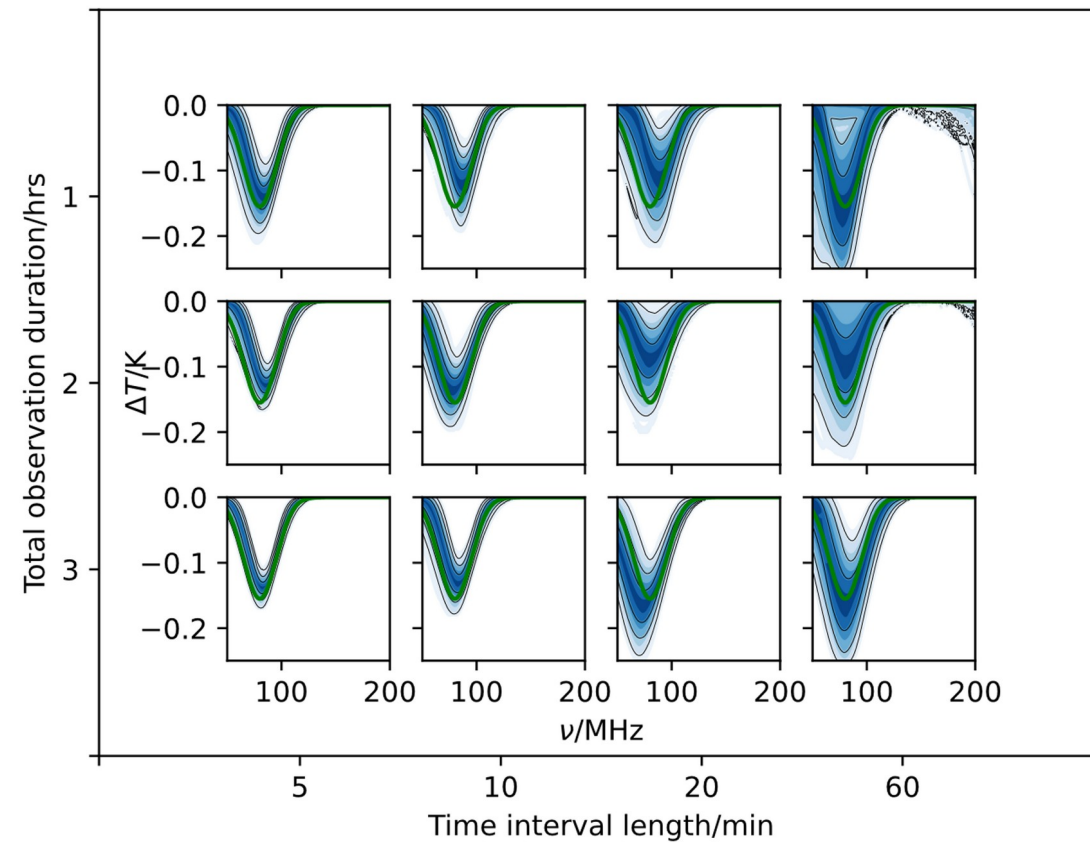
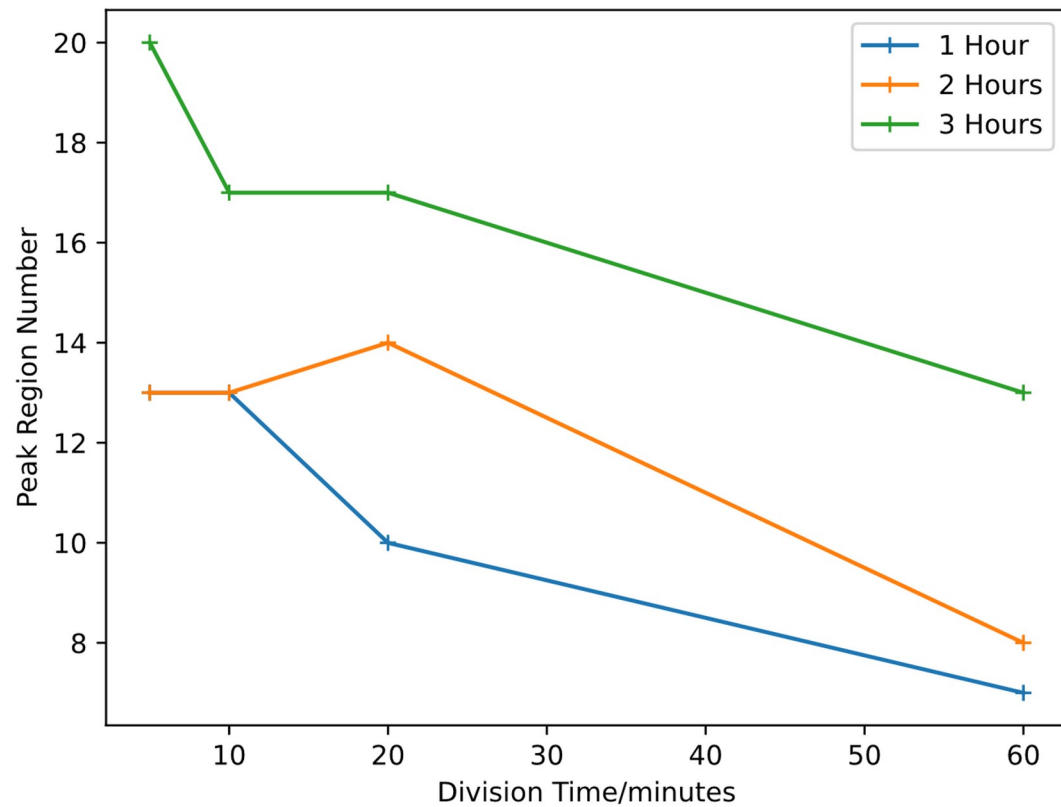


Improving Fit Accuracy



$$\log \mathcal{L} = \sum_i \sum_j \left[-\frac{1}{2} \log (2\pi\theta_\sigma) - \frac{1}{2} \left(\frac{T_{\text{data}}(\nu_i, t_j) - T_{\text{F}}(\nu_i, t_j, \theta_{\text{F}}) - T_{\text{S}}(\nu_i, \theta_{\text{S}})}{\theta_\sigma} \right)^2 \right]$$

Improving Fit Accuracy



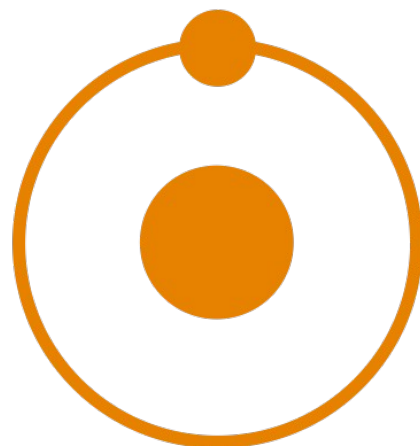
Conclusions

- The REACH pipeline uses physical models of foregrounds to correct for chromatic distortions
- Analysing simulated data with this pipeline can indicate efficacy of a given antenna in a Global 21cm experiment
- Demonstrated this process with 5 example antennae.
- Physically motivated parameters allow multiple time-dependent data sets to be fit simultaneously
- Demonstrated the improvement in foreground modelling and signal fitting that this provides.



Antenna EM simulations
provided by John Cumner
and Quentin Gueuning

Plots produced using fgivenx
tool: Handley, 2018



REACH

