## A general Bayesian framework for high fidelity interferometric calibration with applications to 21 cm cosmology

### Peter H. Sims MSI Fellow, McGill University

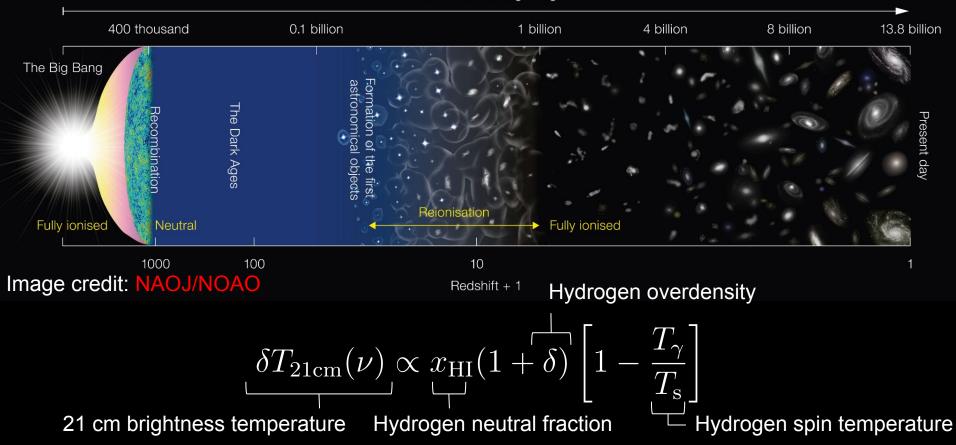
Cosmology from Home, 2022



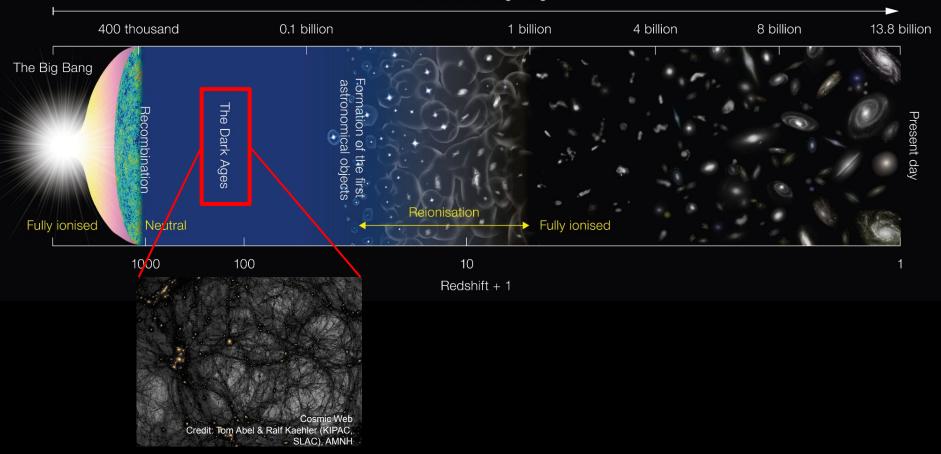




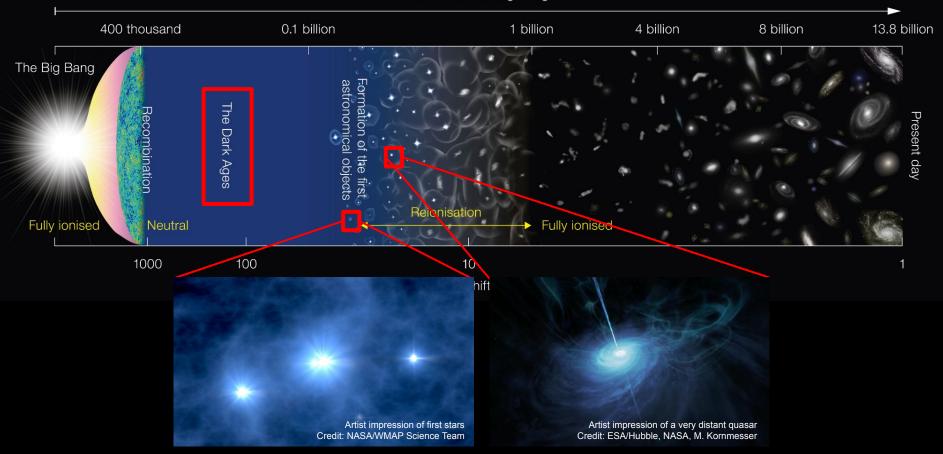
Years after the Big Bang



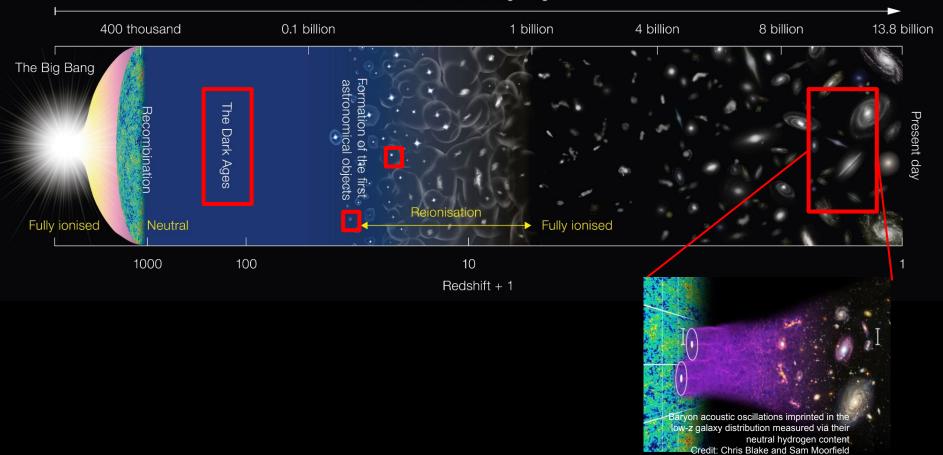
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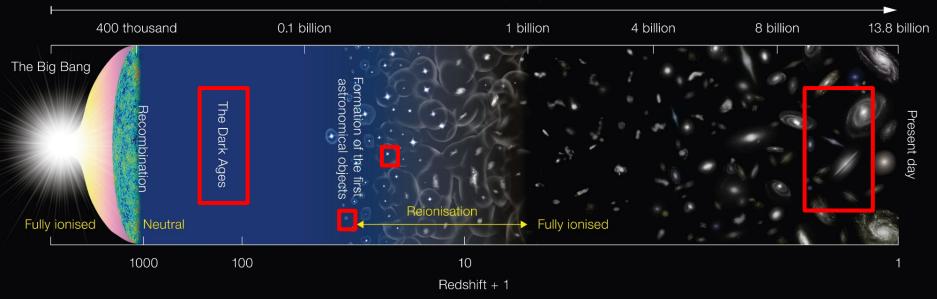
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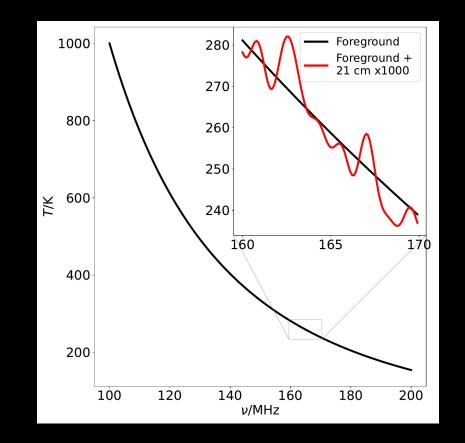
Each of these measurements is reliant on exquisite instrumental calibration, requiring modelling of the instrument bandpass at a spectral fidelity of 1 part in  $10^3$ - $10^6$ 

## 21 cm cosmology data analysis

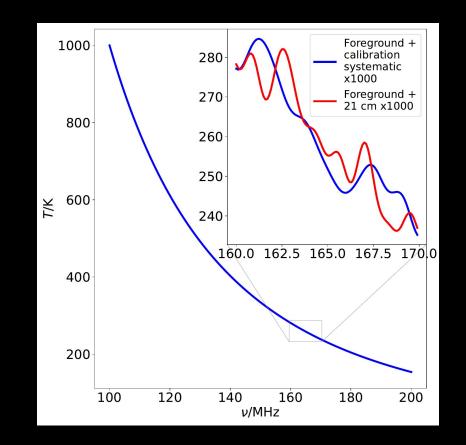


• The goal: extract the redshifted 21 cm signal from data dominated by Galactic synchrotron and bremsstrahlung radiation and extragalactic radio sources that, combined, are 3 - 6 orders of magnitude brighter

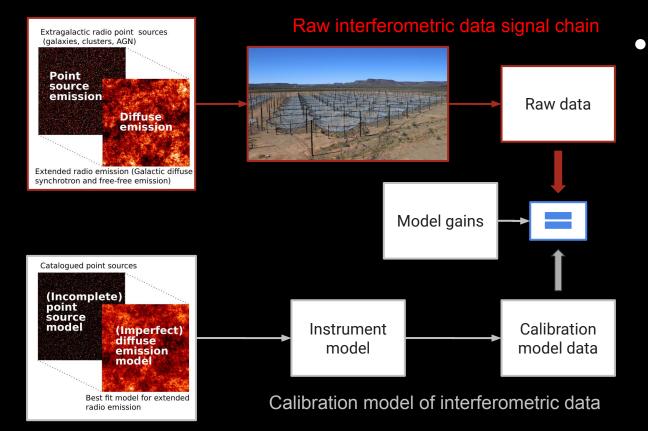
## 21 cm cosmology data analysis



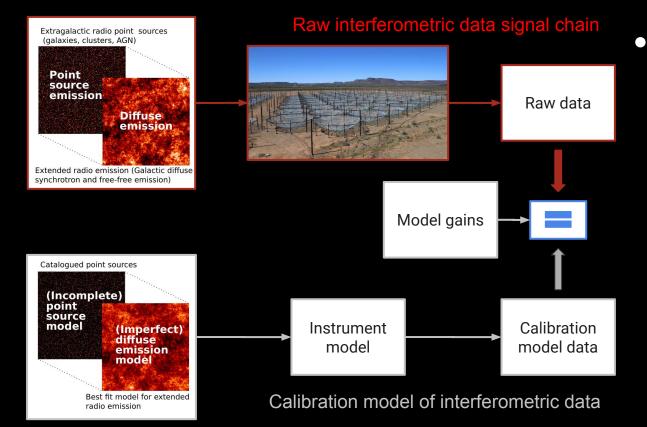
## The calibration challenge



## How interferometers are calibrated

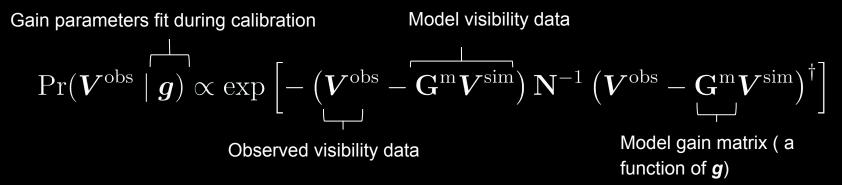


Interferometers are calibrated by fitting the raw data with a calibration model of the data comprised of 1) a known sky model, 2) a known instrument model and 3) a set of unknown per-antenna complex gains to be determined in the fit



**Errors and** incompleteness in the calibration sky model, due to uncertainties in low-frequency sky maps and flux-density limited catalogues of extragalactic sources, translate to biased per-antenna complex gains parameters, resulting in spurious spectral structure in the data

#### Define a complex Gaussian likelihood for sky calibration:



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Gain parameters fit during calibrationModel visibility data
$$\Pr(V^{obs} \mid g) \propto \exp\left[-\left(V^{obs} - G^m V^{sim}\right) N^{-1} \left(V^{obs} - G^m V^{sim}\right)^{\dagger}\right]$$
Observed visibility dataModel gain matrix ( a function of g)

$$m{V}^{
m obs} = {f G}^{
m true}m{V}^{
m true} + m{n}$$
  
If  $m{V}^{
m sim} = m{V}^{
m true}$ , then  $\lim_{m{n}
ightarrow m{0}} {f G}^{
m m} = {f G}^{
m true}$ 

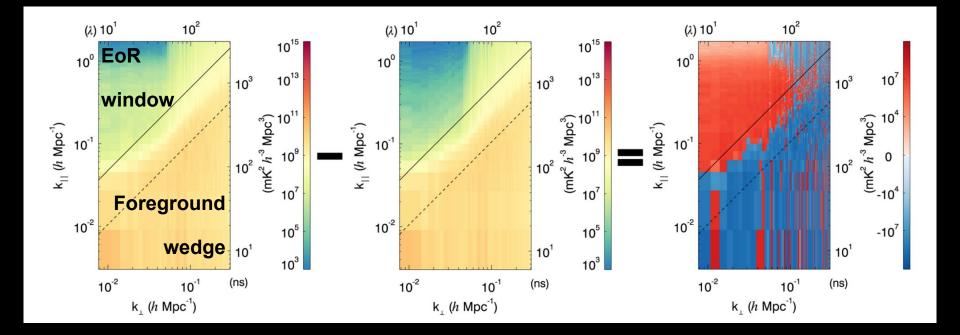
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Observed visibility dataModel gain matrix ( a function of g)

 $egin{aligned} V^{ ext{obs}} = ext{G}^{ ext{true}} V^{ ext{true}} + oldsymbol{n}_{ ext{rue}} \end{aligned}$ 

**Problem:** if the simulated calibration model is incomplete (sources missing, diffuse emission missing or imperfectly modelled),

$$oldsymbol{V}^{ ext{sim}} 
eq oldsymbol{V}^{ ext{true}}$$
 then,  $\lim_{oldsymbol{n} o oldsymbol{0}} \operatorname{G}^{ ext{m}} 
eq \operatorname{G}^{ ext{true}}$ 



The residual power spectrum derived using an incomplete sky model to calibrate the data has excess power in the 'EoR window' Figure adapted from: Barry et al. 2016

## Solutions to the sky model incompleteness problem

1. Better sky models

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Better sky models
 A very smooth instrument bandpass

## Solutions to the sky model incompleteness problem

- Better sky models
   A very smooth instrument bandpass
  - 3. Novel calibration algorithms

## BayesCal

- Fully Bayesian calibration framework to jointly estimate, with the calibration solutions, a statistical model of the incomplete component of the calibration sky model constrained by a prior on the expected angular power spectrum of this sky model component, to eliminate sky-model incompleteness bias
- First two papers currently in review (Sims, Pober & Sievers 2022a,b)

A Bayesian approach to high fidelity interferometric calibration I: mathematical formalism

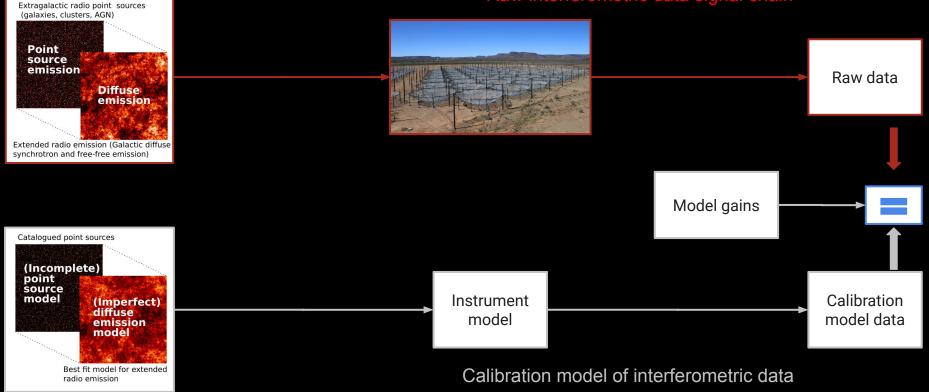
Peter H. Sims, <sup>1,2</sup> Jonathan C. Pober, <sup>3</sup> and Jonathan L. Sievers<sup>1,2</sup> <sup>1</sup>McGill Space Institute, McGill University, 3550 University Street, Montreal, QC H3A 2A7, Canada <sup>2</sup>Department of Physics, McGill University, 3600 University Street, Montreal, QC H3A 2T8, Canada <sup>3</sup>Department of Physics, Brown University, Providence, RI 02912, USA

A Bayesian approach to high fidelity interferometric calibration II: demonstration with simulated data

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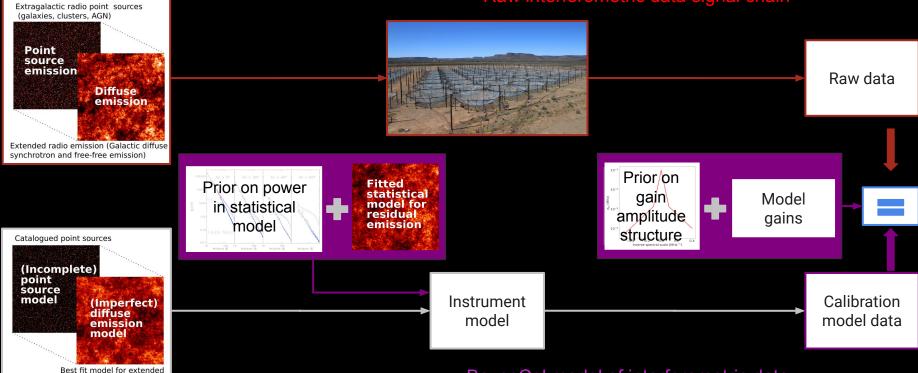
## BayesCal framework





Sims, Pober & Sievers 2022a

## BayesCal framework



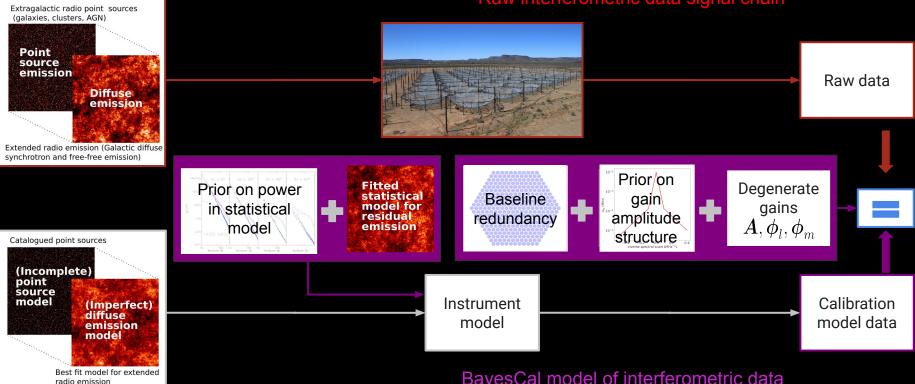
radio emission

#### Raw interferometric data signal chain

BayesCal model of interferometric data

Sims, Pober & Sievers 2022a

## BayesCal framework

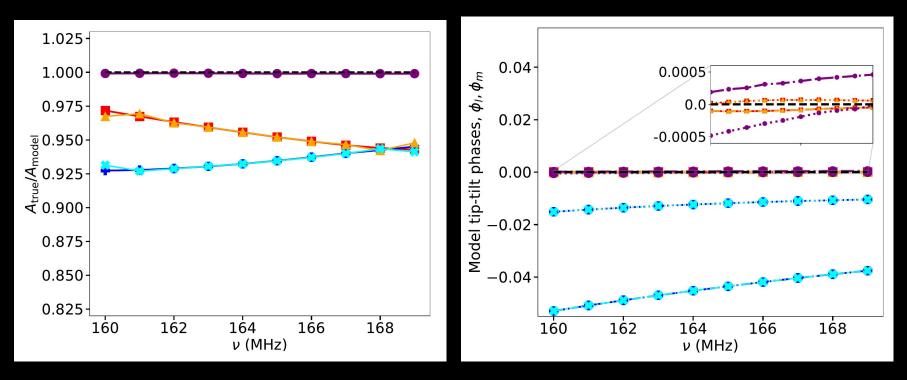


#### Raw interferometric data signal chain

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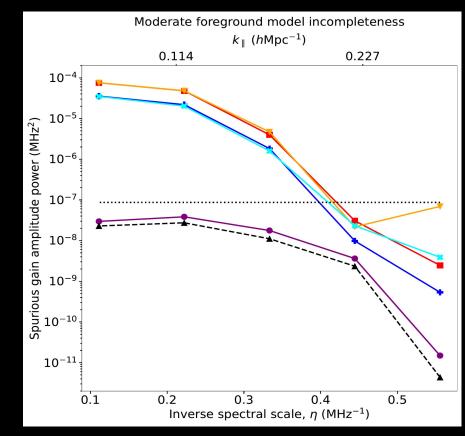
Sims, Pober & Sievers 2022a

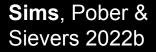
BayesCal calibration solutions (average bias mitigation)



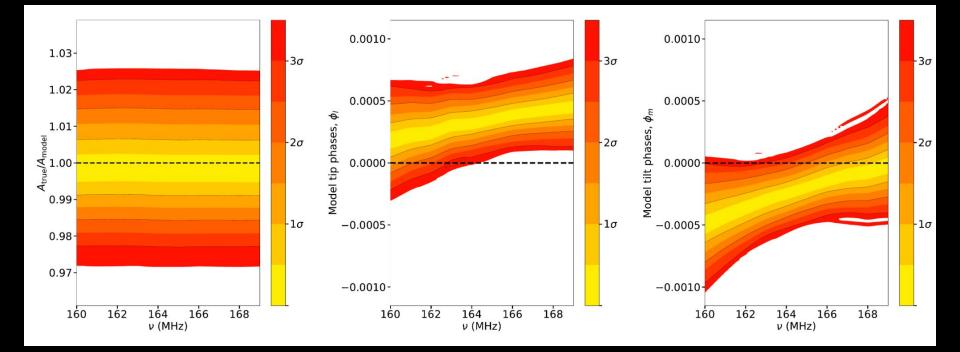
Sims, Pober & Sievers 2022b

# BayesCal calibration solutions (spurious power mitigation)





## BayesCal calibration solutions (full posteriors)



Sims, Pober & Sievers 2022b

## Thank you for listening!

- Intrinsic spectral structure (no instrumental effects)
- Spectral structure after standard calibration

 Spectral structure after calibration using BayesCal

