

Title: Data Science in Radio Cosmology

Date: Mar 24, 2016 11:00 AM

URL: <http://pirsa.org/16030130>

Abstract: <p>In recent decades probing for the subtle indications of new physics in<br>experimental data has become increasingly difficult. The datasets have gotten<br>much bigger, the experiments more complex, and the signals ever smaller. Success<br>stories, like LIGO and Kepler, require a sophisticated combination of statistics<br>and computation, coupled with an appreciation of both the experimental realities<br>and the theoretical framework governing the data.<br><br>

In this talk I will look broadly at data science in physics, and how and why it<br>has taken an increasingly central role. I'll highlight specifically my current<br>area of research, radio cosmology: discussing why it is one of the most<br>challenging areas for data science, and describing my work developing optimal<br>and efficient statistical methods for turning terabytes of timestreams into<br>cosmology.</p>

# Data Science in Radio Cosmology

Richard Shaw



a place of mind  
THE UNIVERSITY OF BRITISH COLUMBIA



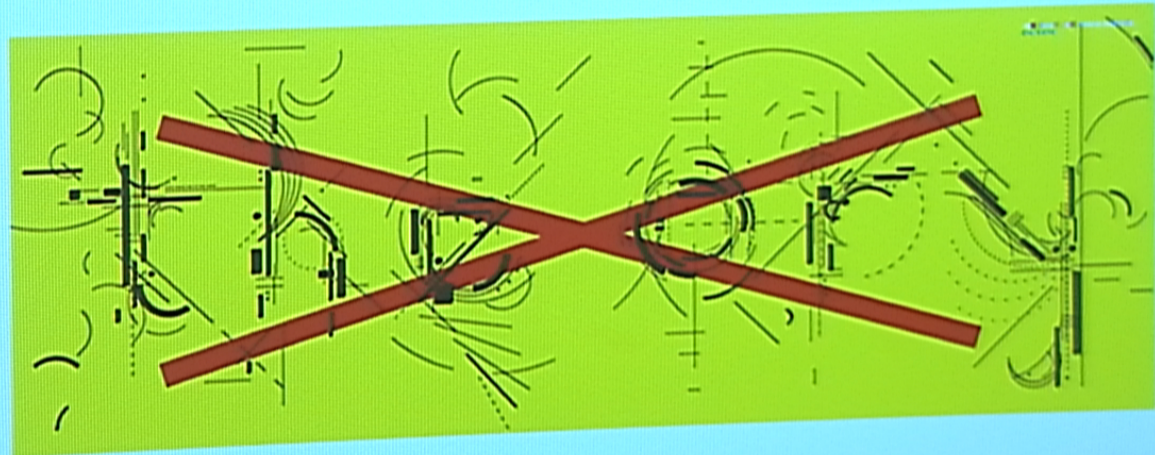
# Data Science

- Combination of:
  - ▶ Statistics
  - ▶ Signal processing
  - ▶ Machine learning
  - ▶ High performance computing
  - ▶ **Physics** (theory and experiment)



CHRIS ANDERSON MAGAZINE 06.23.08 12:00 PM

# THE END OF THEORY: THE DATA DELUGE MAKES THE SCIENTIFIC METHOD OBSOLETE



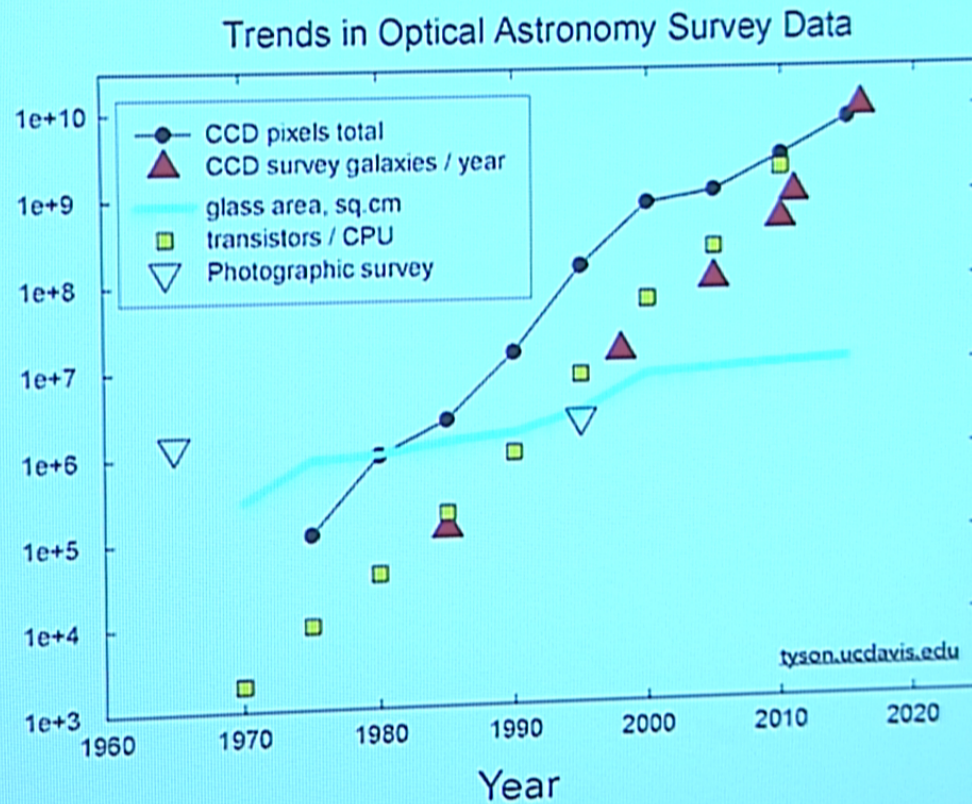


# Data Science *in* Science

- For Science
  - ▶ Model fitting
  - ▶ Model selection
- Tools:
  - ▶ *The fashionable stuff*
  - ▶ Machine learning (classification, regression...)



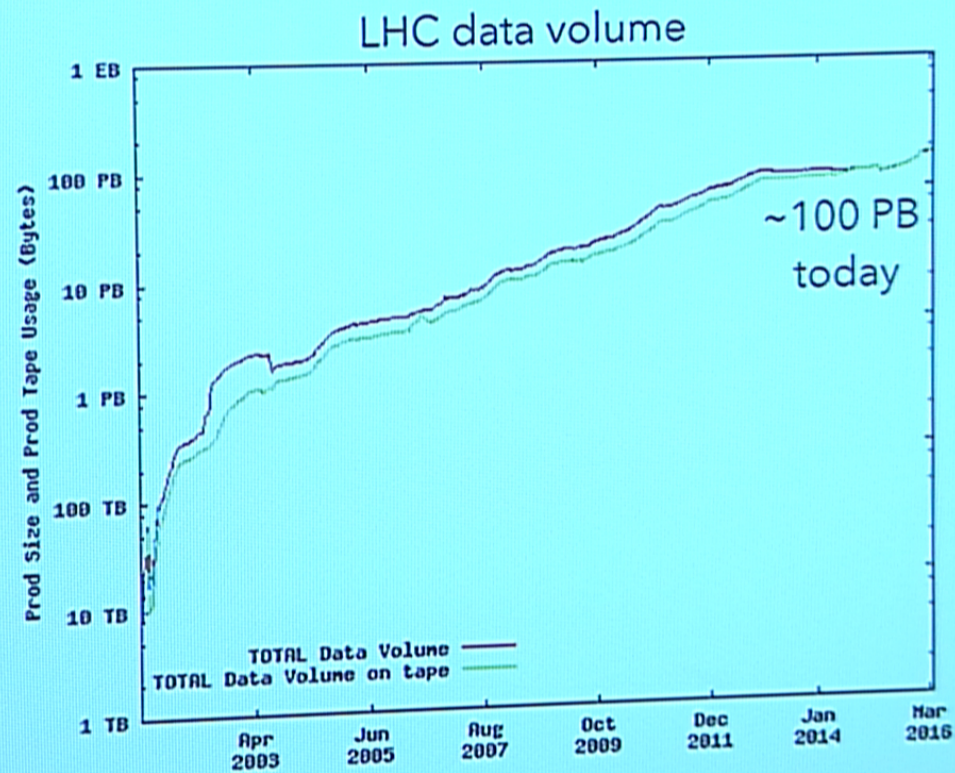
# Why now? Data acquisition...



Exponential growth in data acquisition



# Data growth...



Exponential growth in data size



# Why now?

- Huge increase in accumulation of data
- Volume and complexity means it is becoming a specialised just to do anything with it.



# LHC

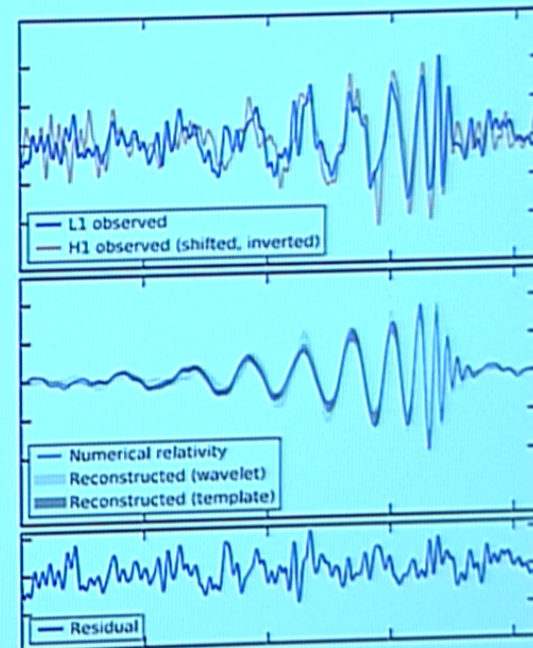
- Need to identify which collisions are interesting
- Each collision is described by many tens of parameters
- Machine classification problem (Boosted Decision Trees)
  - ▶ Trained on simulated data
  - ▶ Selects between events of interest and background events
- Used in real-time in software triggers (LHCb)
- Used for event selection for Higgs detection (CMS)

Gligorov 2014, CMS Collaboration 2012



# LIGO event detection

- Real-time matched filter
  - ▶ Effectively search against 250k template waveforms
  - ▶ Look for peaks in likelihood ratio, keep those that exceed some false positive rate
  - ▶ Keep only events coincident within 15ms in both detectors (~light travel time)



Abbott et al. 2016



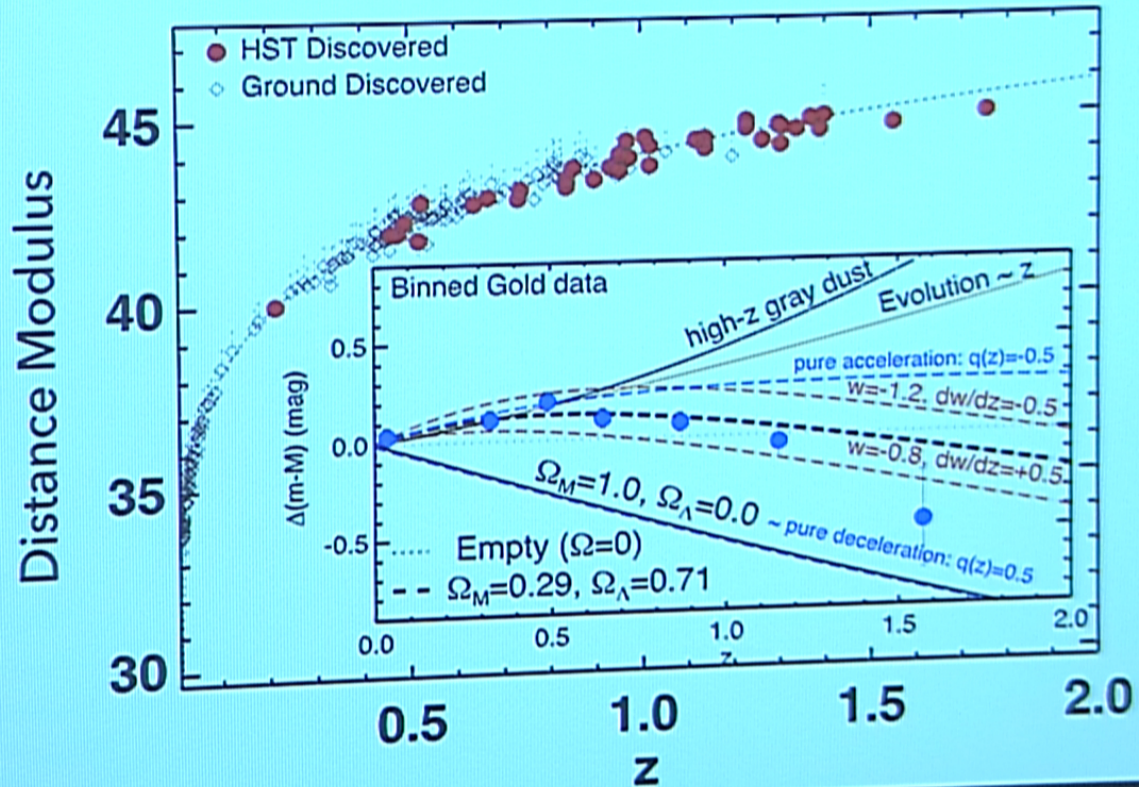
# Data lifecycle

- During acquisition
  - ▶ Real-time phase
- Post-acquisition
  - ▶ Basic results
  - ▶ Processing to likelihood function
- Science
  - ▶ Parameter estimation
  - ▶ Synthesis with other data



# Accelerating Universe

$$D_L = (L/4\pi F)^{1/2}$$



Riess et al. 2007



# Probing Dark Energy

- Acceleration 'explained' by dark energy (cosmological constant, quintessence ...)

- Expansion is governed by Friedmann equation

$$H(z)^2 \approx \Omega_m (1+z)^3 + \Omega_{DE} \exp \left[ \int_0^z (1+w(z)) \frac{dz}{1+z} \right]$$

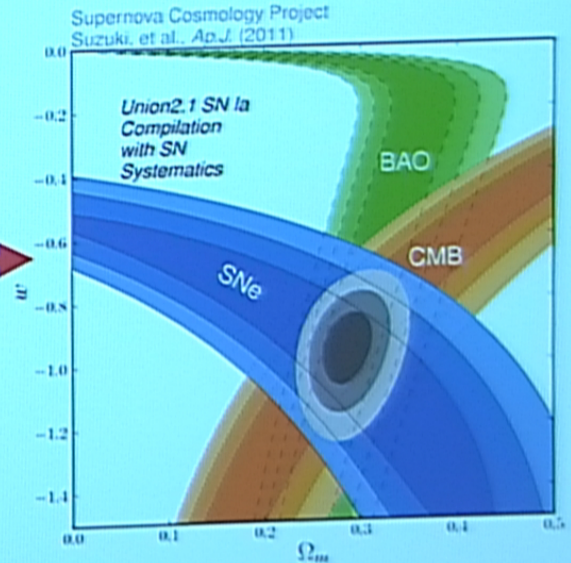
- Fundamental physics is contained within the equation of state  $w(\rho) = p/\rho < -1/3$



# Probing Dark Energy

- Construct likelihood function for dataset i.e.  $\text{Pr}(\text{supernovae}|\text{expansion})$
- Use Markov-Chain Monte-Carlo to infer distribution of relevant parameters

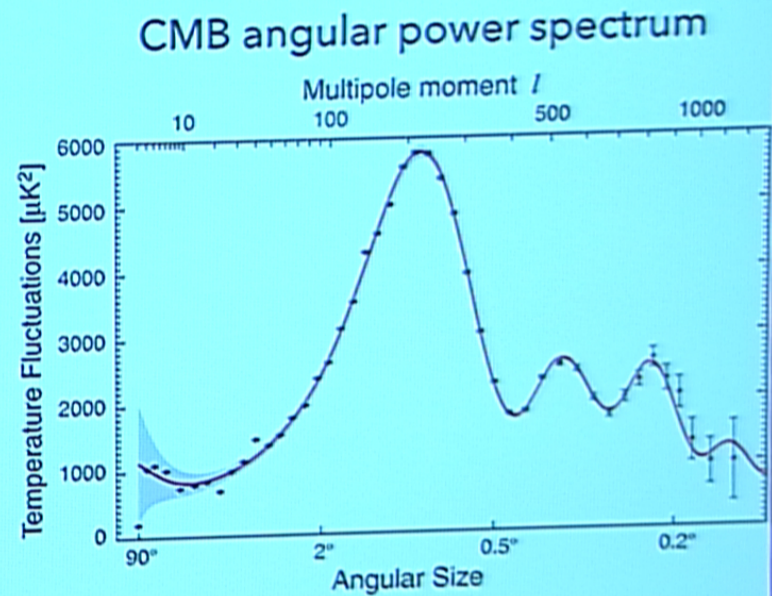
- Propose parameters ( $w_0, w_a, \Omega_m, \dots$ )
- Generate expansion history
- Compare to data (likelihood), accept parameters if likely
- Repeat until enough samples





# Baryon Acoustic Oscillations

- Sound waves propagating in the early Universe. Leave acoustic peaks in the CMB
- Weaker imprint left in the matter distribution
- Gives a standard (statistical) ruler

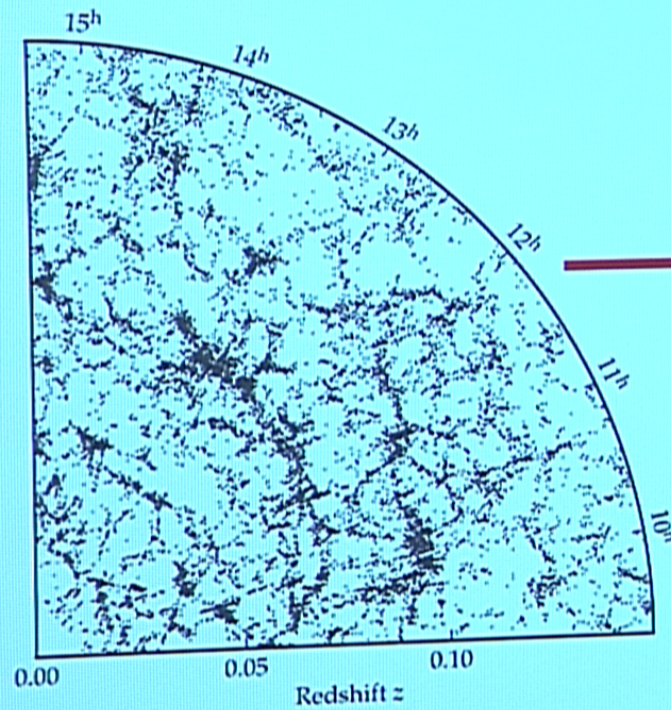


$$r_s = \int_0^{\tau_*} c_s d\tau \sim 100 h^{-1} \text{ Mpc}$$

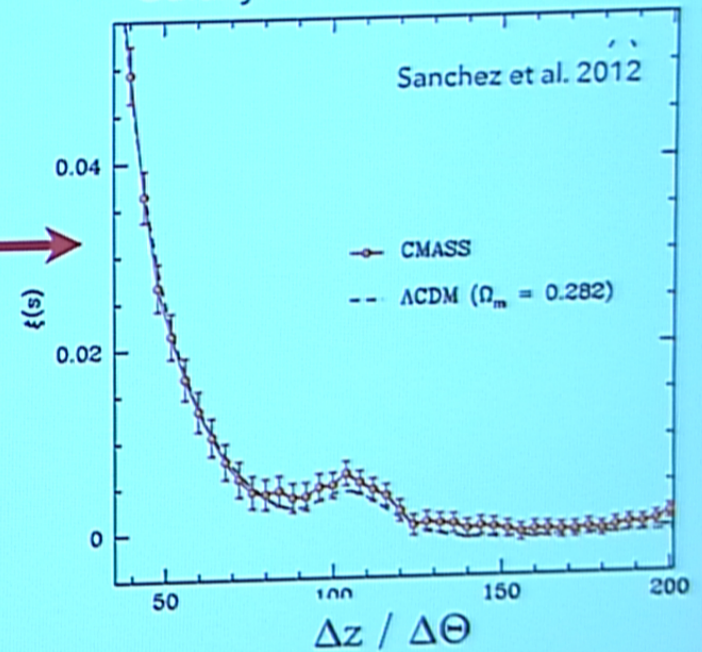
Known from CMB



# Galaxy redshift surveys



Galaxy Correlation Function



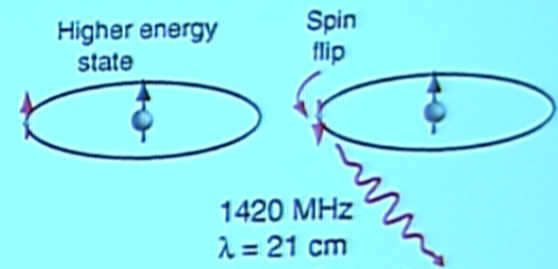
$$r_s = \Delta\theta d_A(z) \quad r_s = \frac{c\Delta z}{H(z)}$$



# 21cm Intensity Mapping



# Cosmological 21 cm



- 21cm line is the transition between parallel and anti-parallel spins of neutral Hydrogen
- The ratio between the two occupancies determines the spin temperature  $T_s$

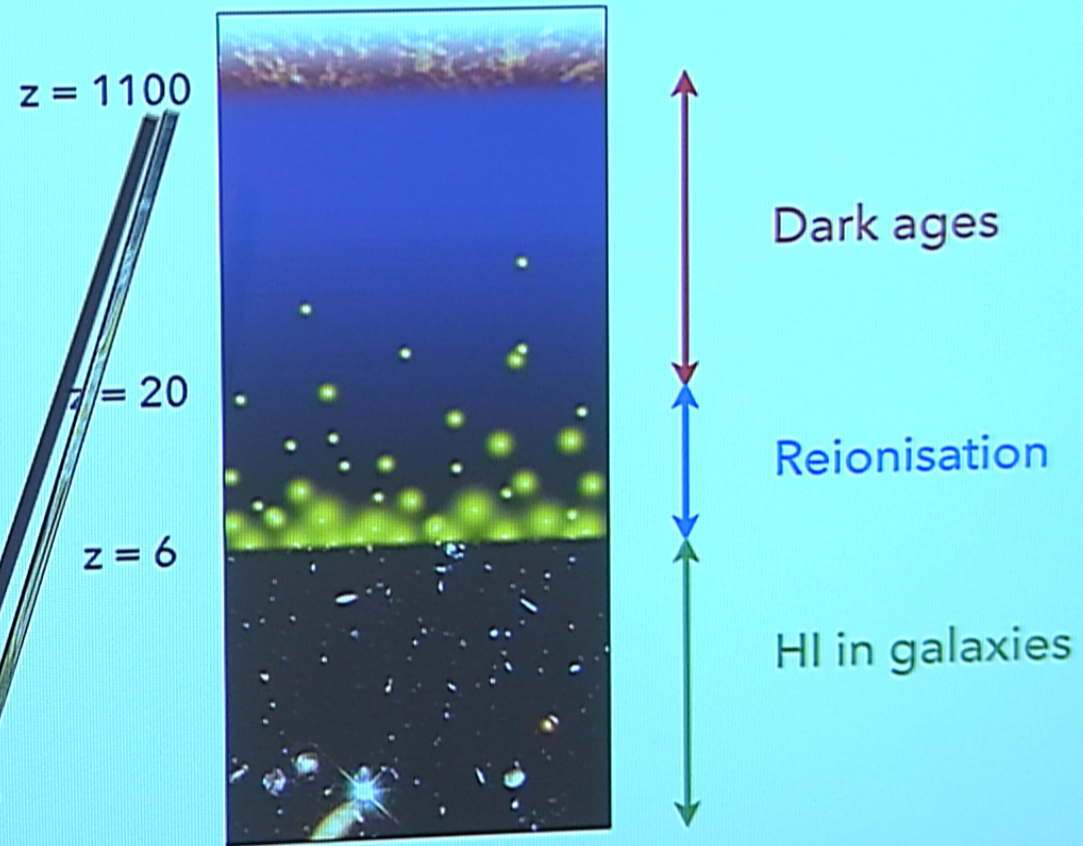
$$n_1/n_0 = (g_1/g_0) \exp(-T_*/T_s)$$

- We can observe the contrast relative to the CMB

$$\Delta T = 23.8 \left( \frac{1+z}{10} \right)^{1/2} [1 - \bar{x}(1 + \delta_x)] (1 + \delta_b)(1 - \delta_v) \left[ \frac{T_s - T_\gamma}{T_s} \right] \text{ mK}$$



# Hydrogen in the Universe



Dark ages

Reionisation

HI in galaxies

Djorgovski et al. (Caltech)



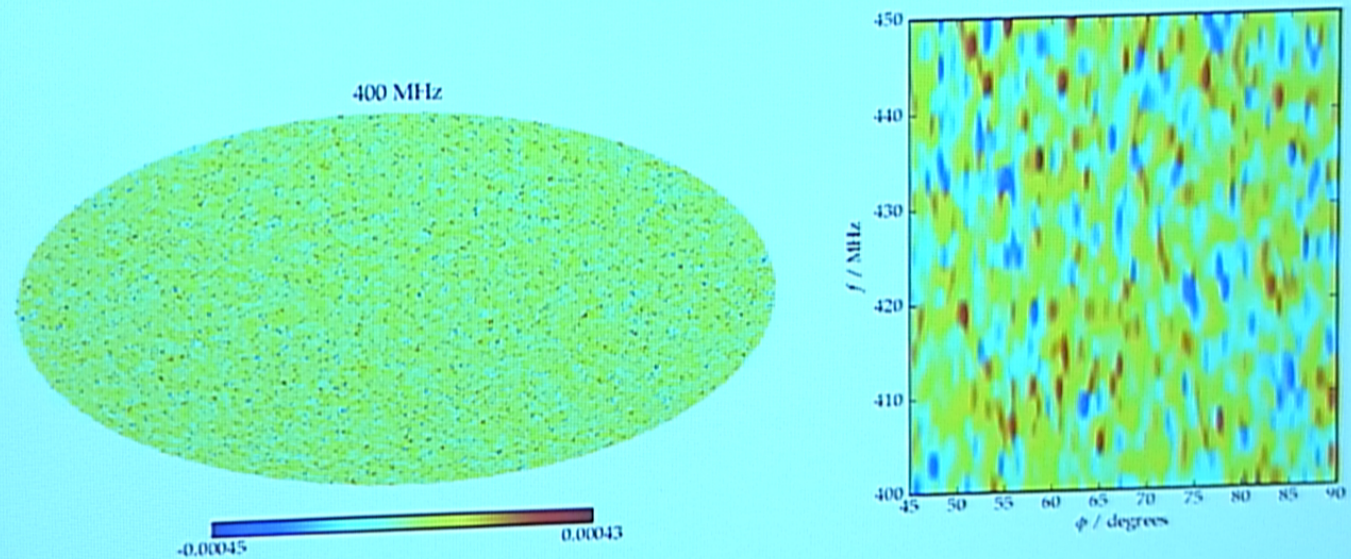
# 21 cm Intensity Mapping

- In 21 cm the frequency gives the redshift.
- Observe the diffuse emission from many unresolved galaxies
- Changes the game in telescope design:
  - ▶ Previously: large field of view, large collecting area, large angular resolution (SKA?)
  - ▶ Now: large field of view, large collecting area, modest angular resolution (compact arrays, single dishes).

Chang, Pen, Peterson and McDonald , 2008, <http://arxiv.org/pdf/0709.3672>



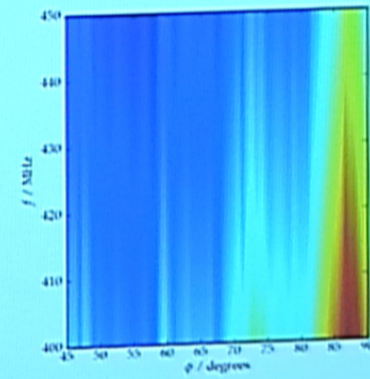
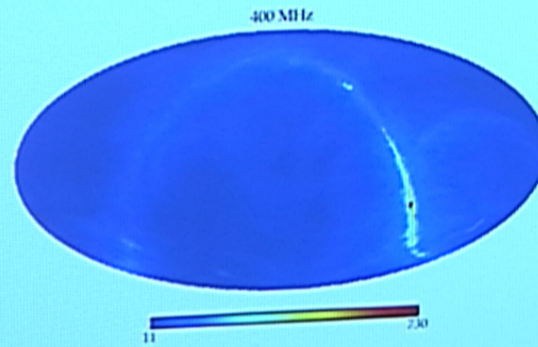
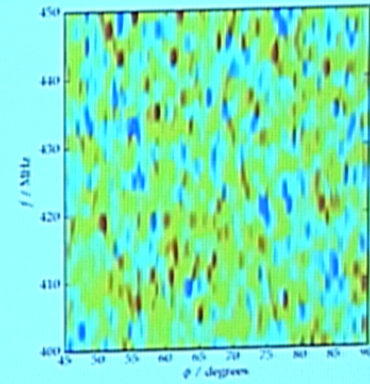
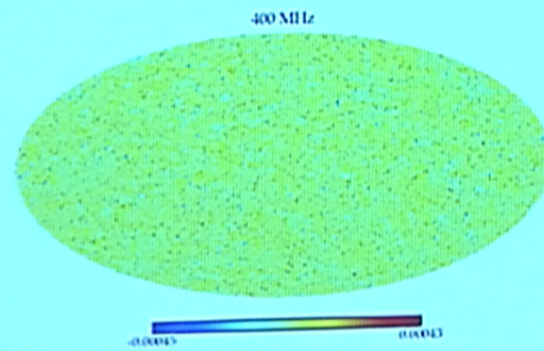
# Foreground Challenges



Cosmological 21cm Signal  $\sim 1\text{mK}$

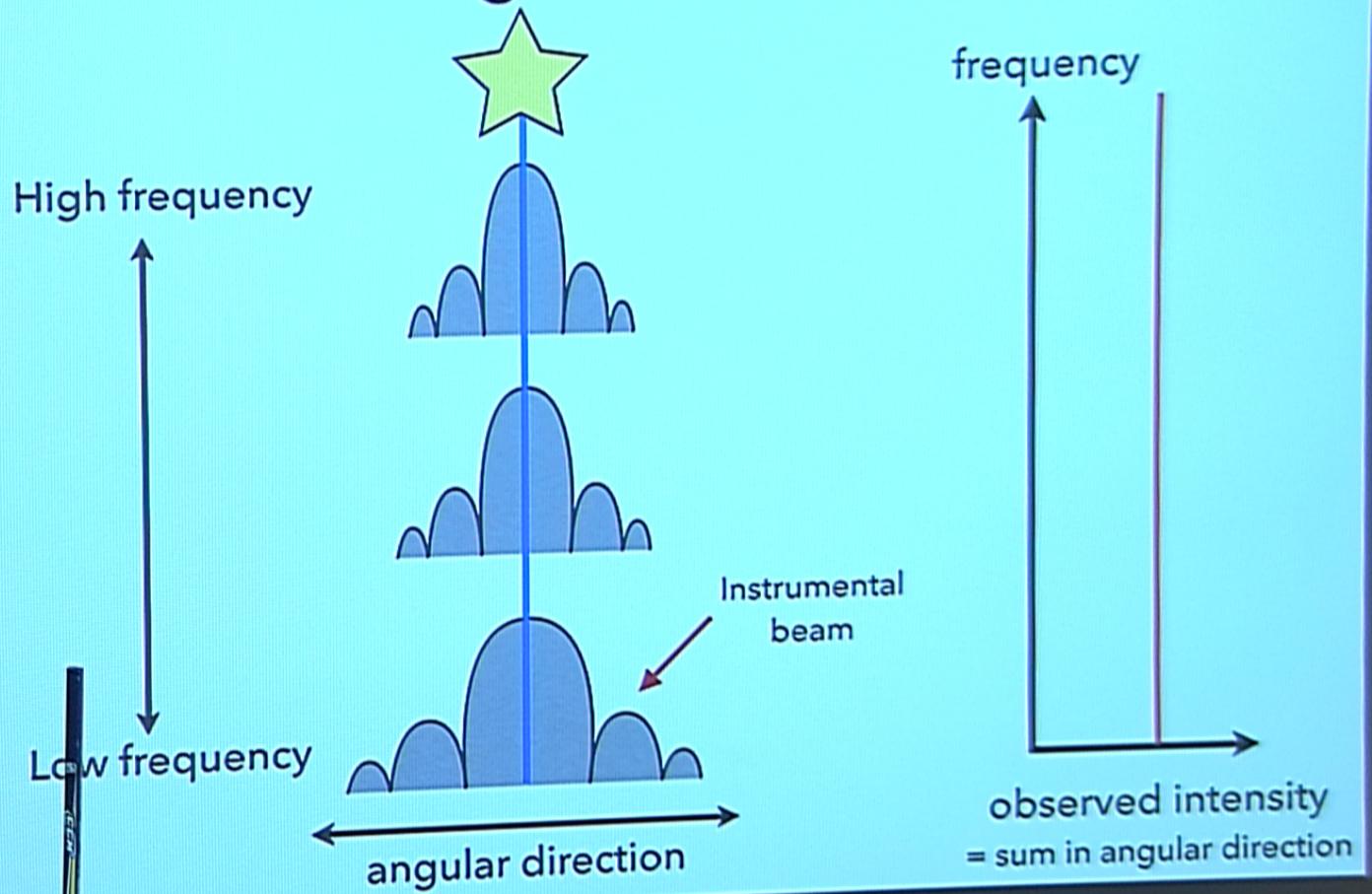


# A way out?



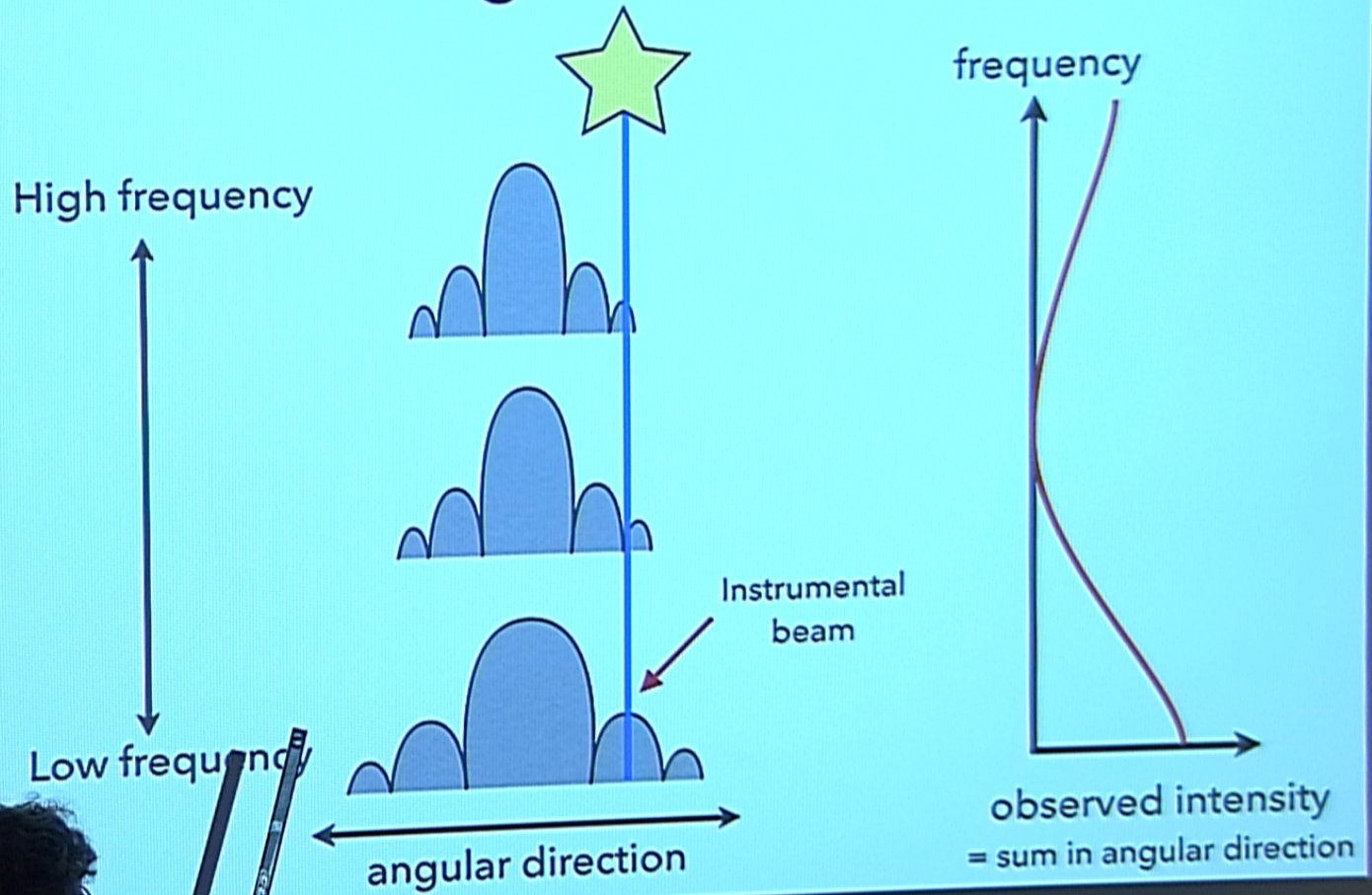


# Mode mixing





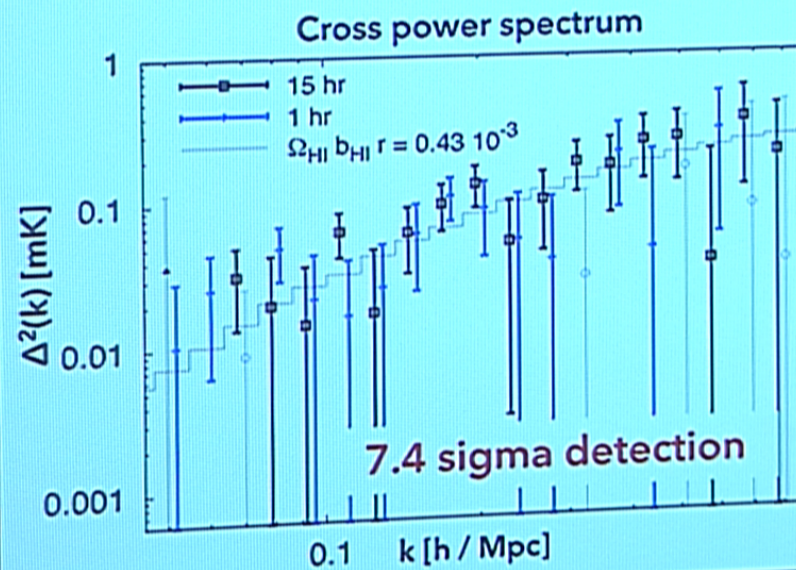
# Mode mixing



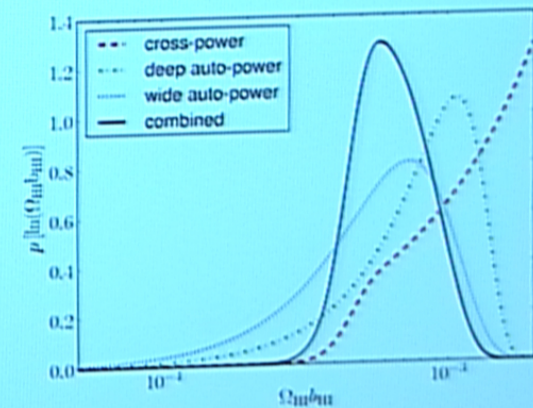


# Cross correlation detection

- Cross-correlation with of GBT data with DEEP2 Galaxy survey by Chang et al.(2010) - *avoids foreground problem!*
- Updated using WiggleZ survey (Masui et al. 2012)



$$\Omega_{\text{HI}} = [0.62^{+0.25}_{-0.15}] \times 10^{-3}$$





# Intensity Mapping at Green Bank



flickr.com/photos/sunlight/



# The Future?

- Work at GBT will continue with the aim of measuring the 21cm *autocorrelation*.
- However, observations like this are slow. To survey the whole sky to this depth ~ 20 years
  - ▶ Is there a better way to do this?











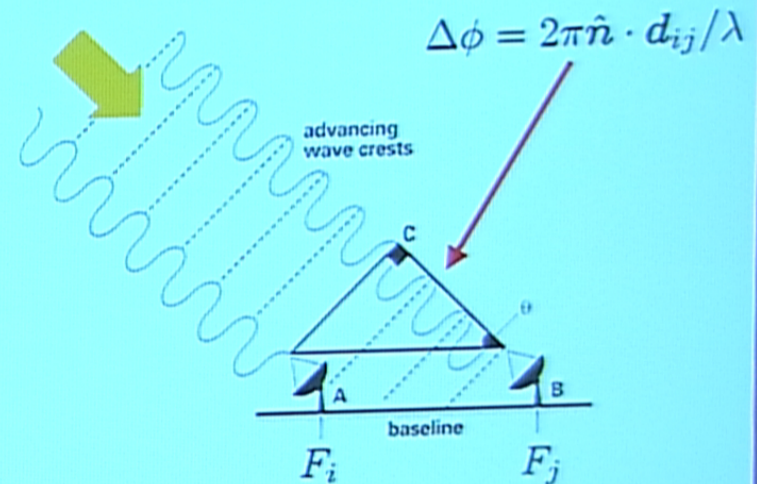
# Interferometers

- Visibility is instantaneous correlation of 2 antennas

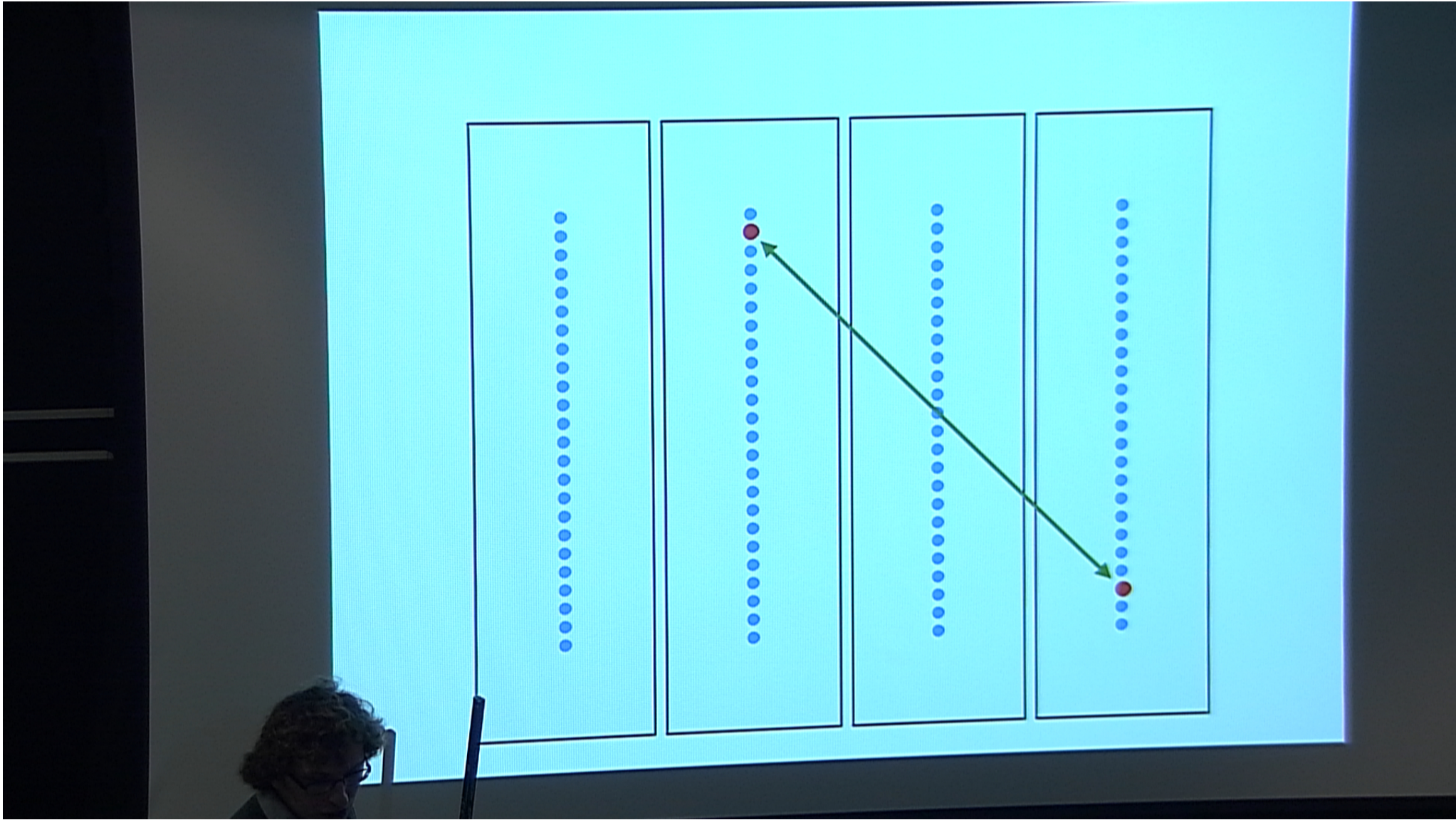
$$V_{ij} = \langle F_i F_j^* \rangle$$

- Each pair measures a Fourier mode of the sky
- Written explicitly:

$$V_{ij}(t) = \frac{1}{\Omega_{ij}} \int d^2 \hat{n} A_i(\hat{n}; t) A_j^*(\hat{n}; t) e^{2\pi i \hat{n} \cdot \mathbf{u}_{ij}(t)} T(\hat{n})$$







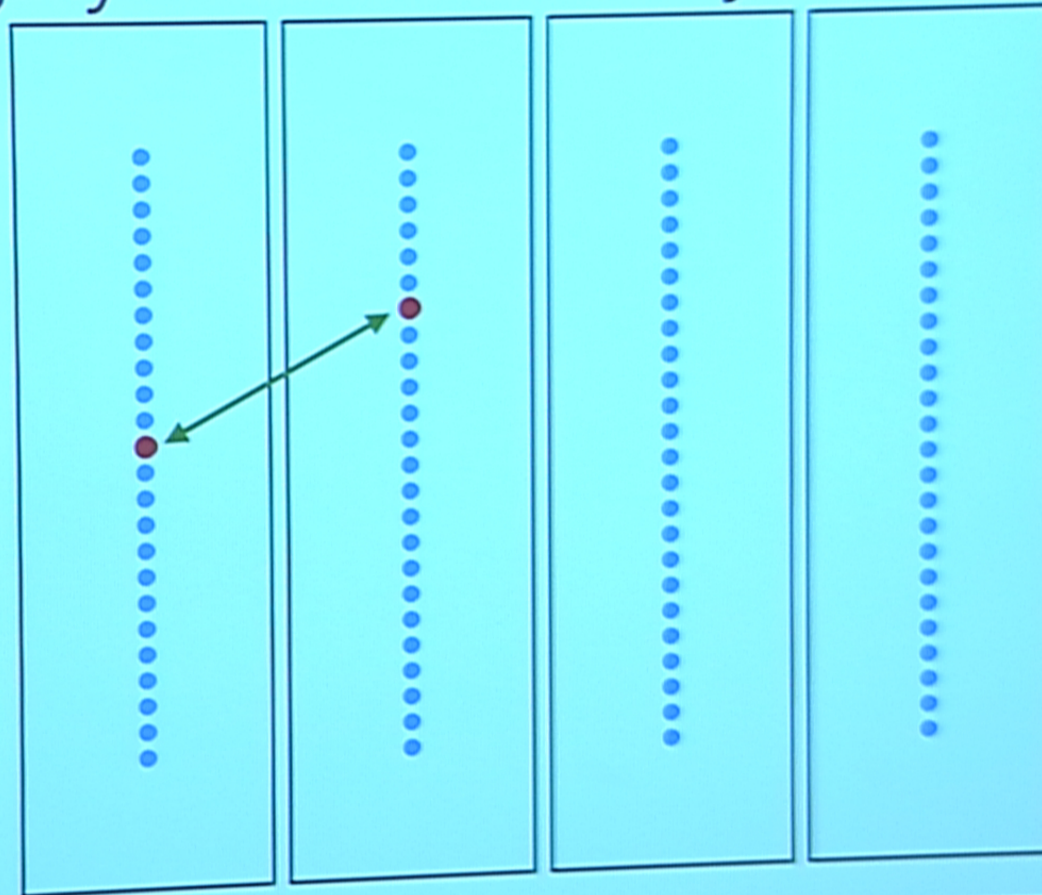


# Data rate

- For full  $N^2$  correlation
  - ▶ ~5 GB/s
  - ▶ ~400 TB/day
  - ▶ ~140 PB/year
- Need a way to significantly compress the data!



# Highly redundant array





# Not so fast! Calibration

- Each feed has an unknown, time-variable, complex gain, from amplifier and cable behaviour
- Must correct for this, or the baselines will average incoherently
- This process is *calibration*, and must be done in **real-time**
  - ▶ Nearly optimal solution via eigenvalue decomposition
  - ▶ Use injected calibration signal
  - ▶ Sky signal pulsars

Newburgh + CHIME, arXiv:1406.2267

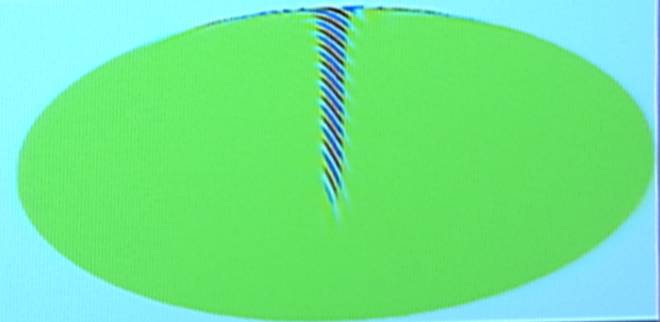


# Interferometers

$$V_{ij}(t) = \frac{1}{\Omega_{ij}} \int d^2 \hat{\mathbf{n}} A_i(\hat{\mathbf{n}}; t) A_j^*(\hat{\mathbf{n}}; t) e^{2\pi i \hat{\mathbf{n}} \cdot \mathbf{u}_{ij}(t)} T(\hat{\mathbf{n}})$$

- Write in terms of a beam transfer function:

$$V_{ij}(t) = \int d^2 \hat{\mathbf{n}} B_{ij}(\hat{\mathbf{n}}; t) T(\hat{\mathbf{n}}) + n_{ij}(t)$$





# Transit Interferometers

- Timeseries is periodic on the sidereal day  $t \rightarrow \phi$ 
  - ▶ Apply this restriction and see how the analysis goes.

$$V_{ij}(\phi) = \int d^2 \hat{\mathbf{n}} B_{ij}(\hat{\mathbf{n}}; \phi) T(\hat{\mathbf{n}}) + n_{ij}(\phi)$$

**Spherical Harmonic Transform**

$$V^{ij}(\phi) = \sum_{lm} B_{lm}^{ij}(\phi) a_{lm}^T + n^{ij}(\phi)$$

**Fourier Transform**

$$V_m^{ij} = \sum_l B_{lm}^{ij} a_{lm}^T + n_m^{ij}$$



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**Fourier Transform**

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# m-mode transform

- Mapping does not mix m's (each is independent)

$$V_m^\alpha = \sum_l B_{lm}^\alpha a_{lm}^T + n_m^\alpha$$

- Write in vector form

$$\mathbf{v} = \mathbf{B} \mathbf{a} + \mathbf{n} .$$

- Simple, linear mapping from the information on the sky, to the measured degrees of freedom
- Discrete relation, with finite number of degrees, can apply all the standard statistical, signal processing techniques.
- Computationally efficient: For 1000 m's an  $O(N^3)$  matrix operation becomes  $10^6$  times faster



# Interferometric Imaging

- Traditional imaging is based around the 2D Fourier Transform approximation to the interferometry equation (only valid on small patches instantaneously)
- Use a series of steps to relax this approximation and increase field of view (w-projection, mosaicking, A-projection)
  - ▶ eg. w-term. From non coplanarity of array and sky. Solve by iteratively deconvolving the effects

$$V = \int dx dy A^2(x, y) e^{2\pi i (ux + vy + w \sqrt{1-x^2-y^2})} I(x, y)$$



# m-mode Imaging

- For our restricted domain (transit telescopes), we can solve the problem exactly.

- Measurement is linear mapping:

$$\mathbf{v} = \mathbf{B} \mathbf{a} + \mathbf{n} .$$

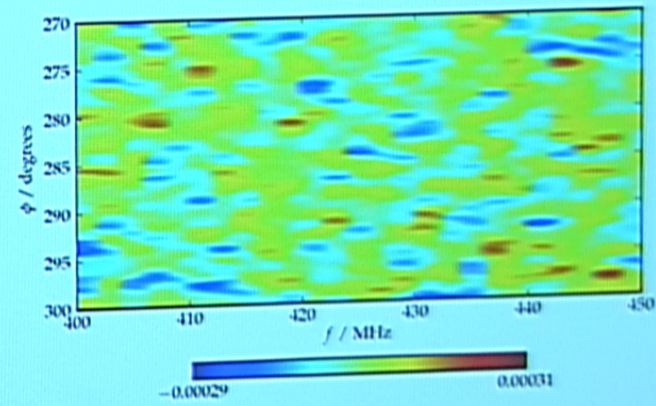
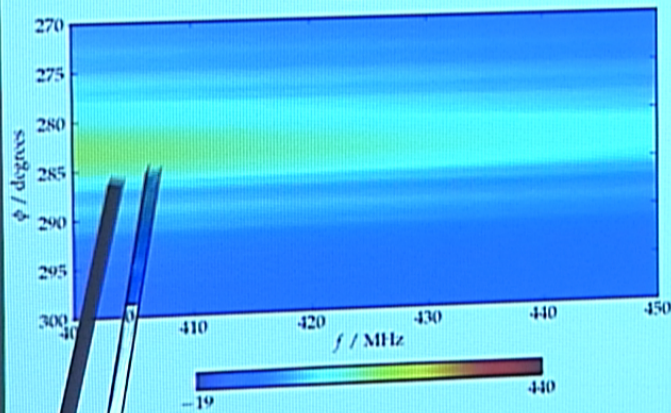
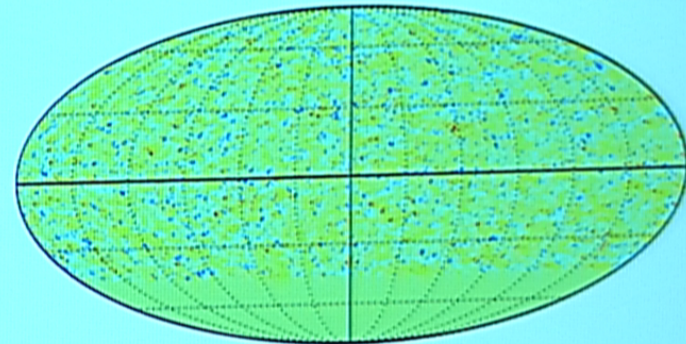
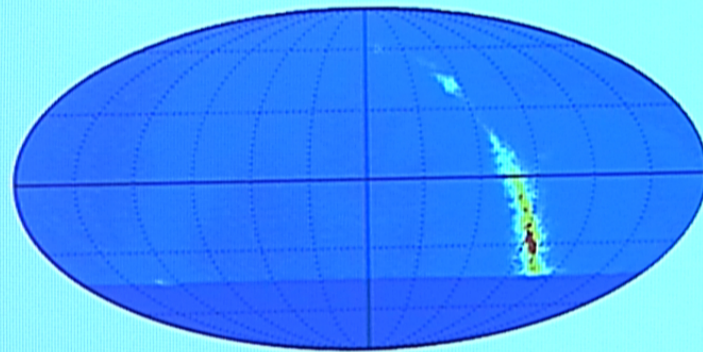
- How do we make an image of the sky? Use standard tools of signal processing:

- ▶ Pseudo-inverse to solve and regularize (*Maximum likelihood*)
- ▶ Wiener Filter (*Bayesian expectation*)

- Conceptually straightforward. Deals naturally with all full sky effects, polarisation etc.



Observed sky (from time stream)  $\hat{\mathbf{a}} = (\mathbf{N}^{-\frac{1}{2}}\mathbf{B})^+ \mathbf{N}^{-\frac{1}{2}}\mathbf{v}$



2x15m wide cylinders, 60 feeds, 0.25m spacing 400-600 MHz



# Foreground Removal

- Spectral smoothness allows separation of 21cm
  - ▶ Measure components and model (Liu, Dillon etc.)
  - ▶ Power spectrum removal (Foreground wedge)
  - ▶ Delay-space filtering (Parsons et al. 2012)
- Most methods have difficulties:
  - ▶ *Mode mixing* of angular and frequency fluctuations by frequency-dependent beams (esp. interferometers)
  - ▶ *Robustness* Biasing introduced if foreground model poorly understood (esp. non-gaussianities)
  - ▶ *Statistical Optimality* Need to keep track of transformations on statistics, for optimal PS estimation
  - ▶ *Polarisation leakage* mixes fluctuations from polarised foreground



# Karhunen-Loeve Transform

- Old CMB idea - E/B mode separation (Bunn et al. 2003)
- An 'optimal' treatment - m-modes makes it feasible.
- Construct the covariances of the signal and foregrounds in the measured basis

$$\mathbf{S} = \langle \mathbf{s} \mathbf{s}^\dagger \rangle = \mathbf{B} \langle \mathbf{a}_s^* \mathbf{a}_s^T \rangle \mathbf{B}^\dagger \quad \mathbf{F} = \mathbf{B} \langle \mathbf{a}_f \mathbf{a}_f^\dagger \rangle \mathbf{B}^\dagger$$

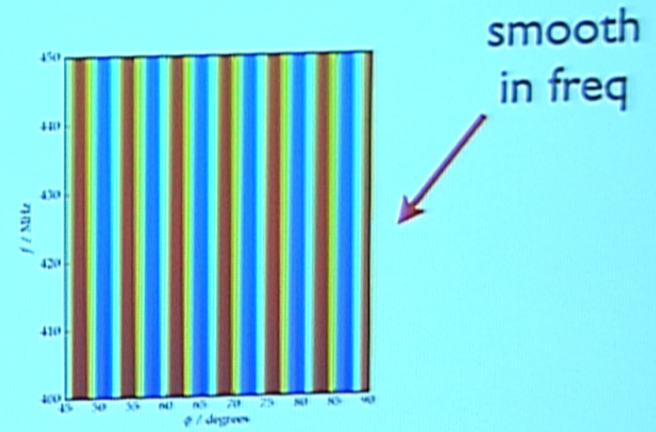
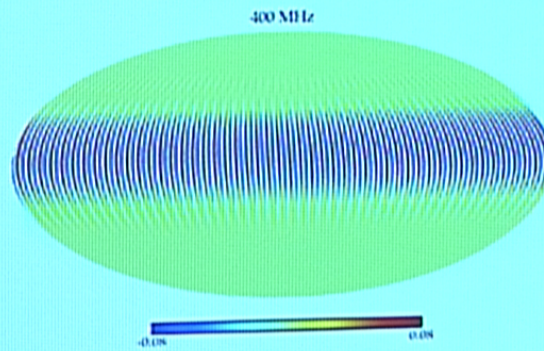
- Jointly diagonalise both (eigenvalue problem)

$$\mathbf{S} \mathbf{x} = \lambda \mathbf{F} \mathbf{x}$$

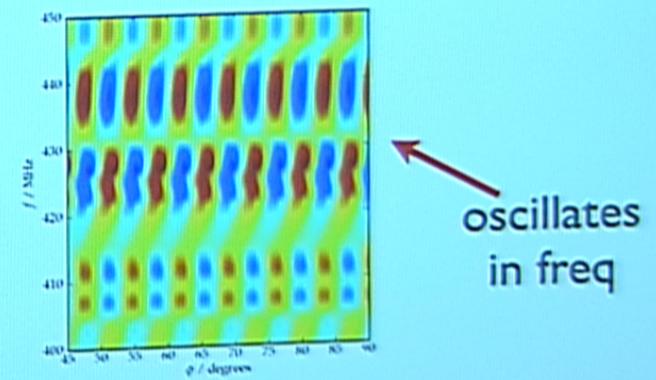
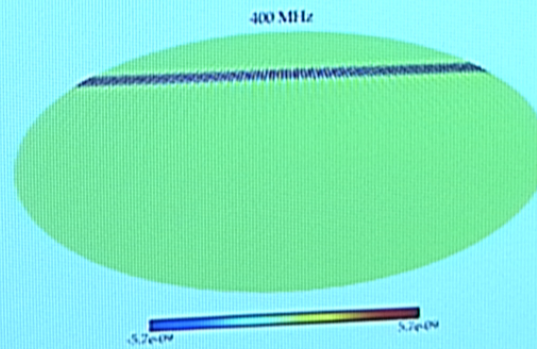
- Gives a new, uncorrelated basis. Corresponding eigenvalue gives the expected signal to foreground power ratio.



## Most foreground



## Most signal





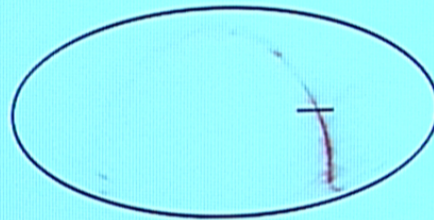
# Foreground Removal with KLT

- Foreground removal is performed by projecting out modes with low signal-to-foreground ratio.
- Robustness to model uncertainties by choosing a conservatively large threshold; we would prefer to increase our errors bars in order to remove bias.
- Addresses the previous problems
  - ▶ Analysis uses all measured data to avoid mode mixing.
  - ▶ Can be made arbitrarily robust - increase threshold for removal
  - ▶ Linear transformation in the data space, keeps track of statistics

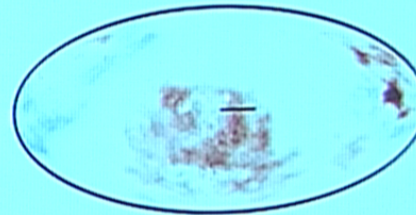


# Foreground Cleaning

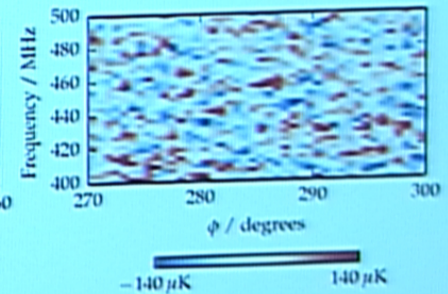
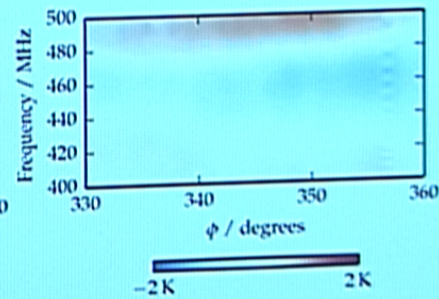
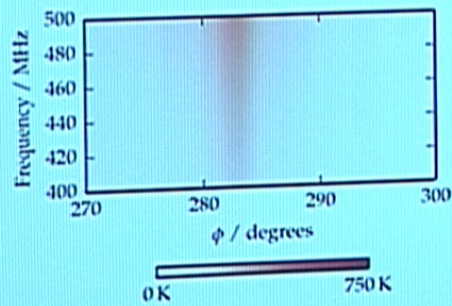
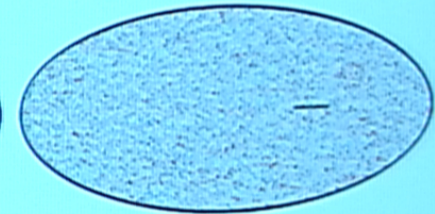
Unpolarised Foreground



Polarised Foreground (Q)



21cm Signal

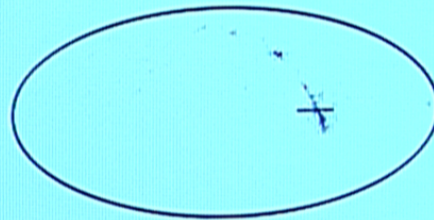


Foregrounds  $10^6$  times larger than signal

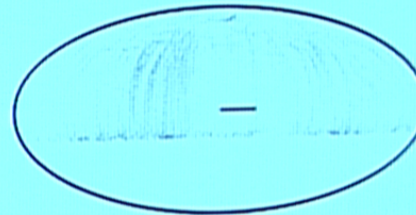


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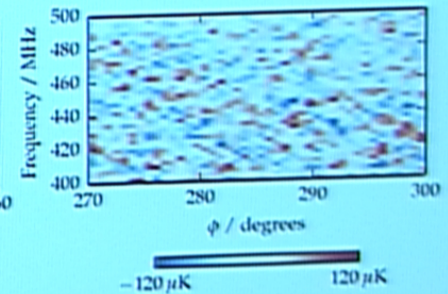
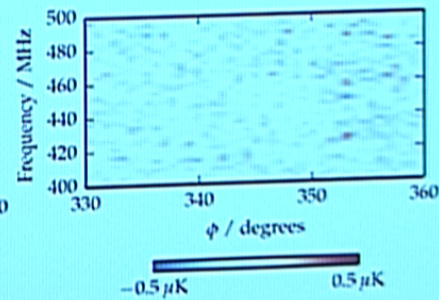
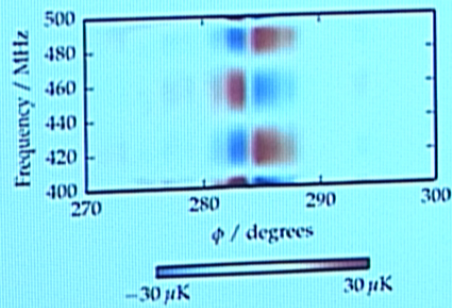
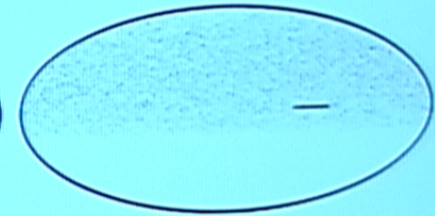
Unpolarised Foreground



Polarised Foreground (Q)



21cm Signal

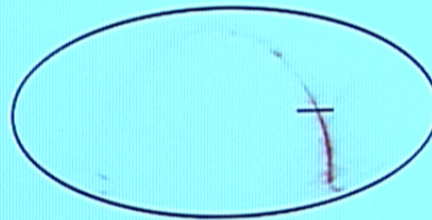


Foreground residuals significantly smaller than signal

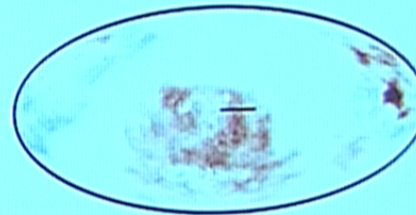


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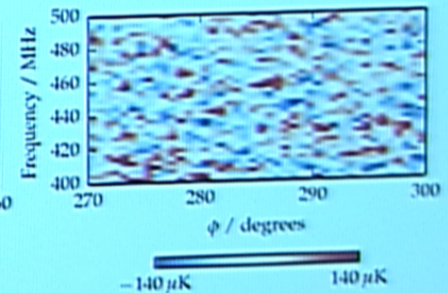
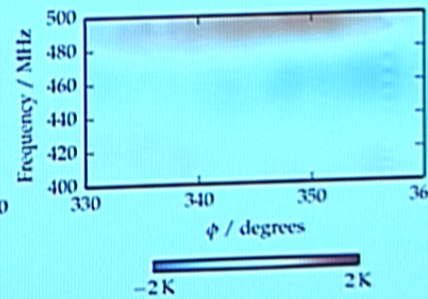
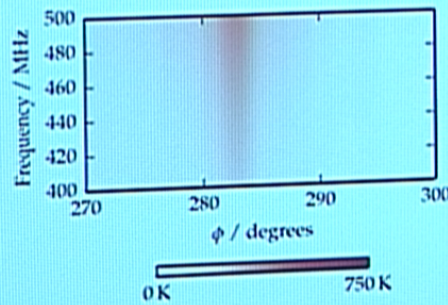
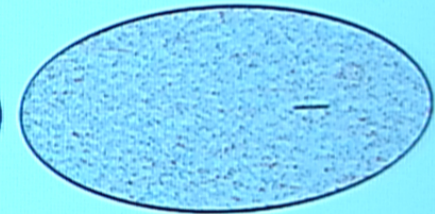
Unpolarised Foreground



Polarised Foreground (Q)



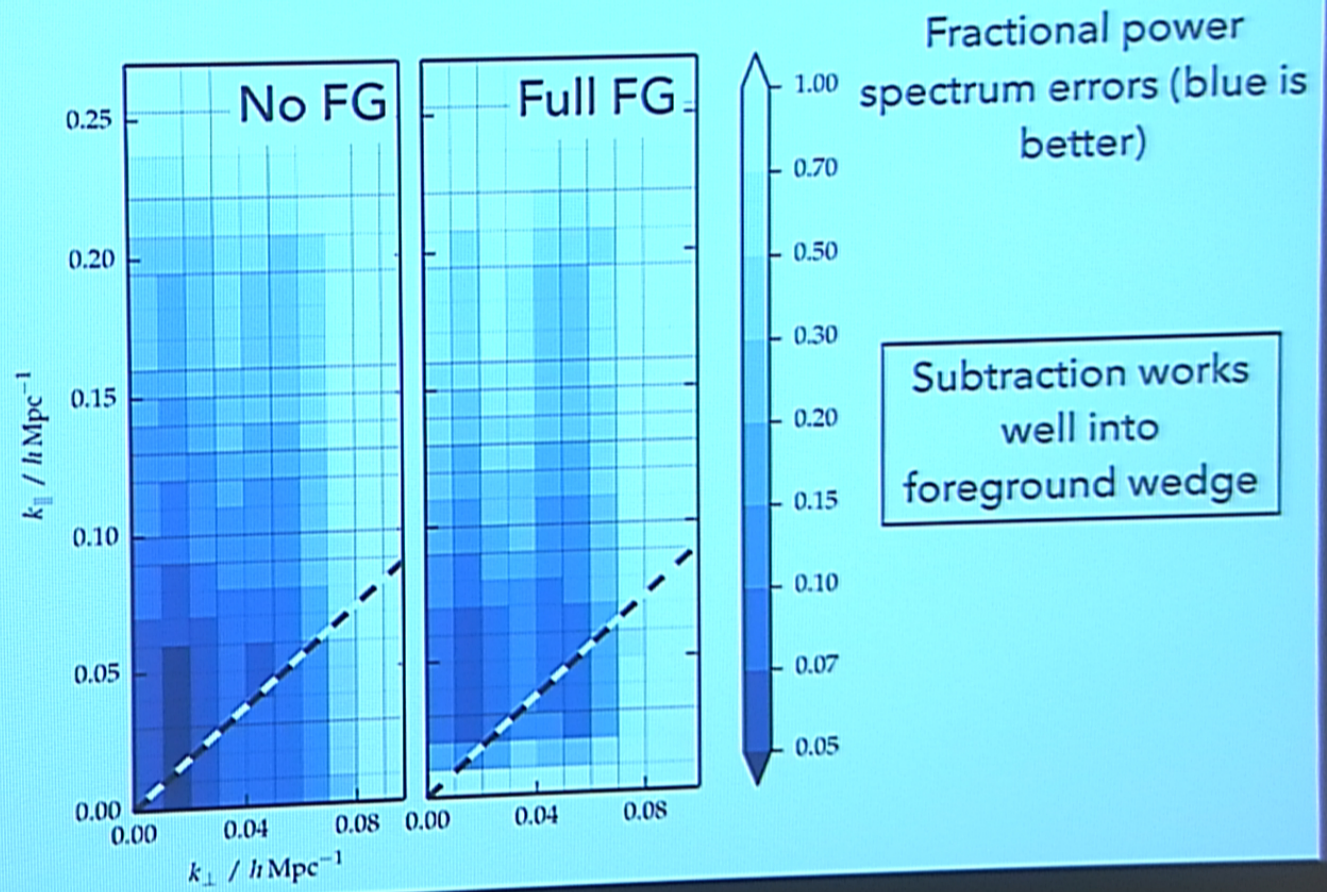
21cm Signal



Foregrounds  $10^6$  times larger than signal



# 2D Power spectrum Estimation





# Summary

- Data Science is becoming an increasingly large and distinct part of physics
- 21cm Intensity Mapping is a promising technique for mapping the Universe and measuring BAOs.
- Data volume and foregrounds are challenging
- New techniques, like the m-mode formalism show promise for surmounting them