Constraining reionization with the global 21cm signal and kSZ

Joëlle-Marie Bégin

In collaboration with Adrian Liu and Adelie Gorce

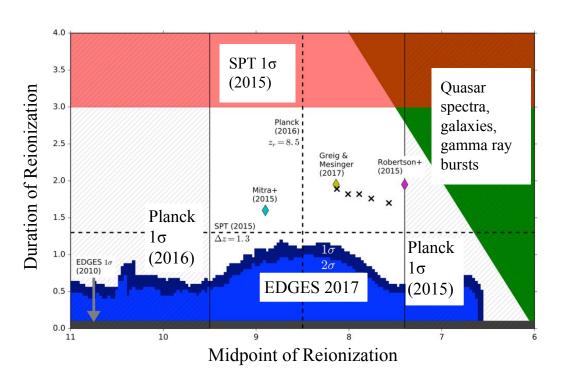


Overview

- The global 21cm signal and kinetic Sunyaev Zeldovich effect are **complementary** probes of reionization.
- The **Karhunen-Loeve** basis highlights this complementarity.
- Working in this basis facilitates the **detection and removal of systematics.**

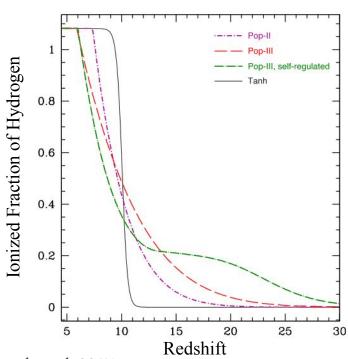
The ionization history

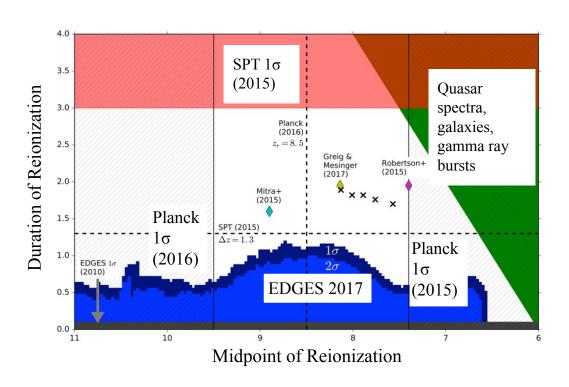
■ We have some bounds for its midpoint, end, and duration.



The ionization history

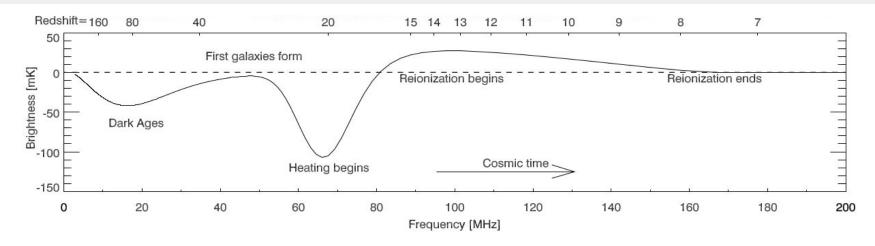
- We have some bounds for its midpoint, end, and duration.
- Few limits on precise shape.



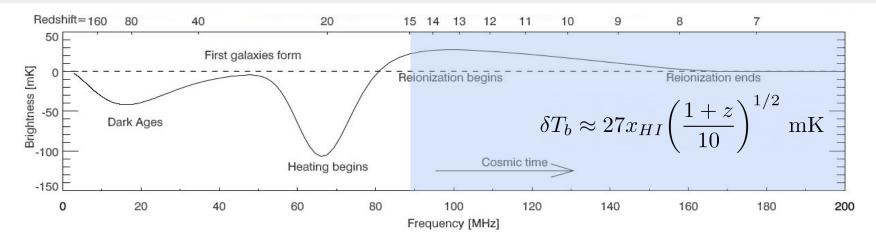


Miranda et al, 2017

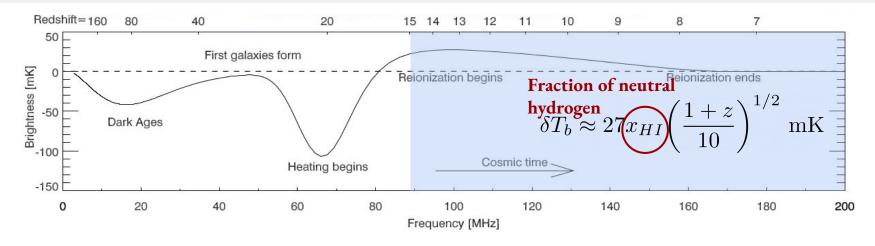
Monsalve et al, 2017



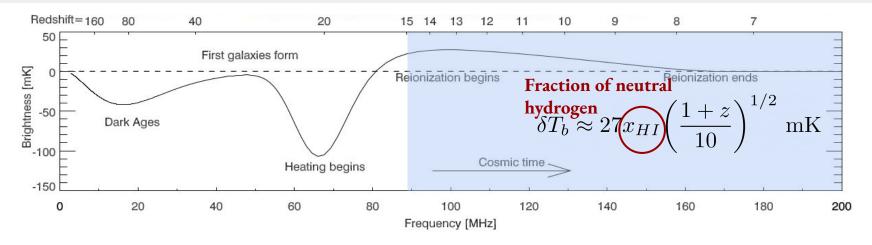
■ During reionization, the global signal closely tracks the ionization history



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- During reionization, the global signal closely tracks the ionization history
- The global signal is **most sensitive to rapidly evolving reionization histories** due to spectrally smooth foregrounds

The kinetic Sunyaev-Zeldovich effect (kSZ)

 CMB photons scattering off of energetic electrons with bulk relative velocity

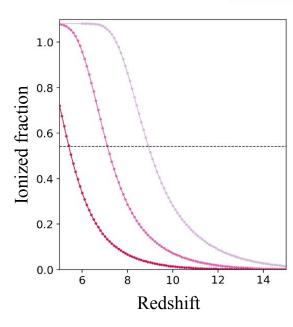


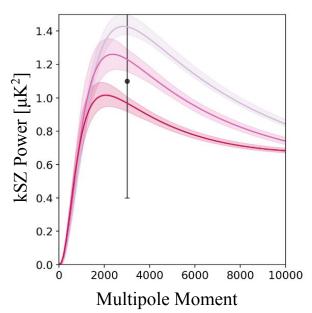
The kinetic Sunyaev-Zeldovich effect (kSZ)

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Power spectrum
changes with
midpoint, duration,
morphology of
reionization



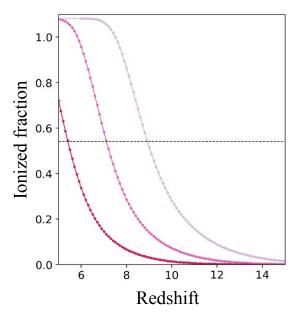


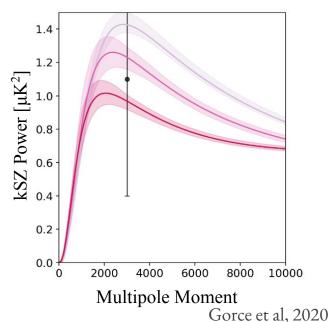
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- Power spectrum
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- Sensitive to extended ionization histories.

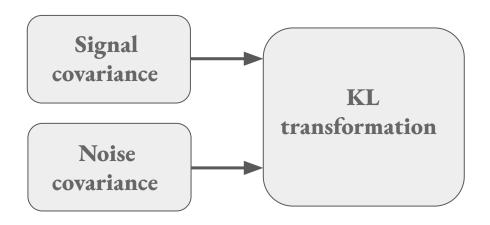




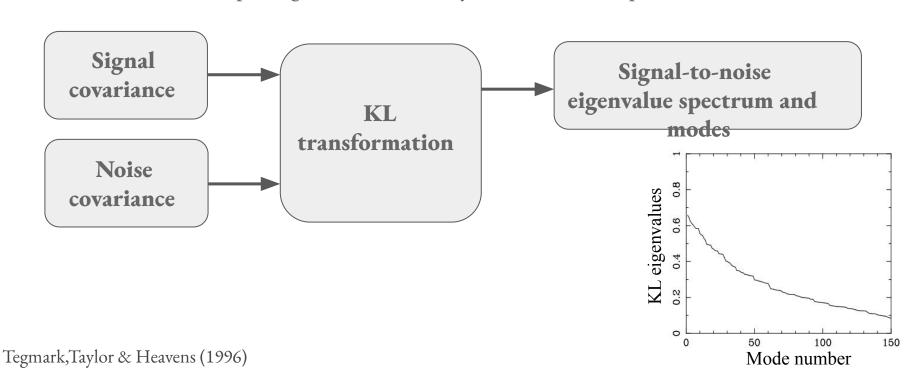
The global signal is sensitive to rapidly evolving ionization histories, the kSZ to extended ionization histories.

- A transformation whose eigenvalues represent the ratio of two signals.
- Familiar example: signal-to-noise analysis and data compression.

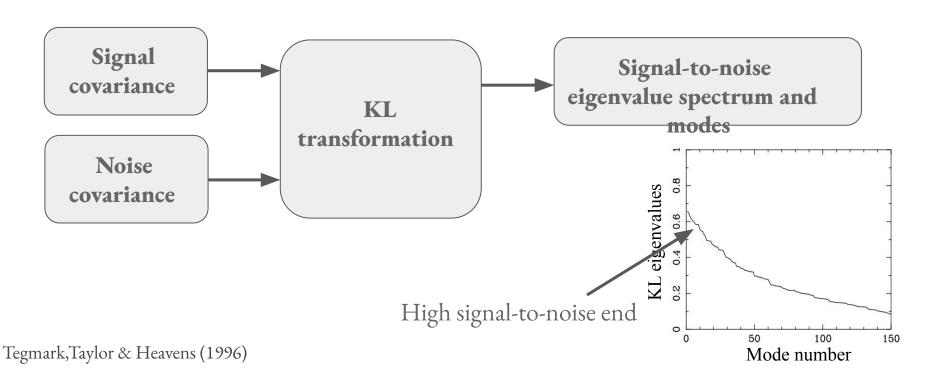
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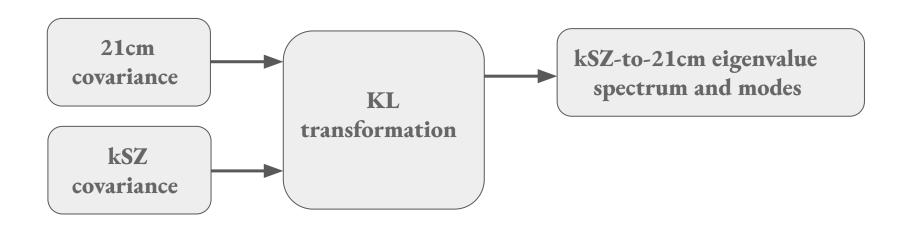
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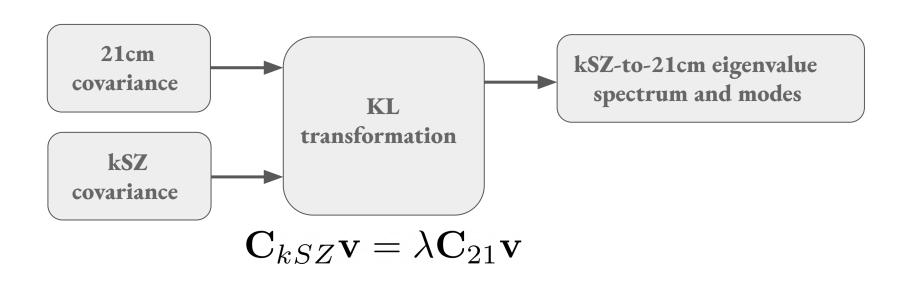
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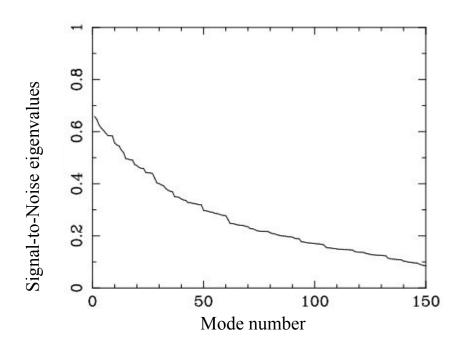


■ In our case: kSZ-to-21cm analysis

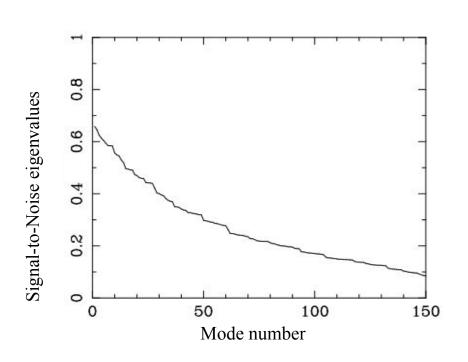


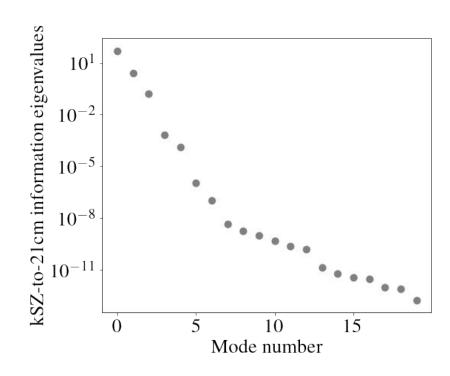
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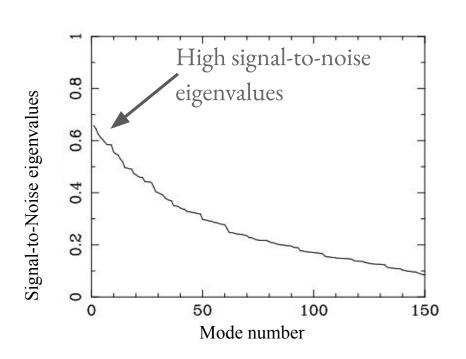
Tegmark, Taylor & Heavens (1996)

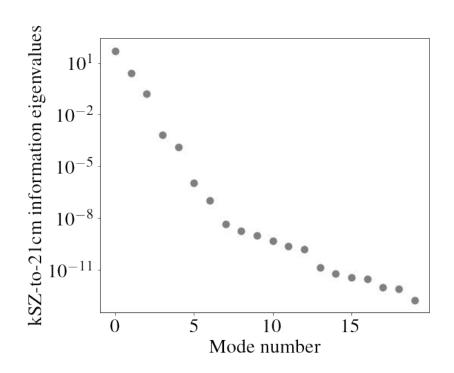




Tegmark, Taylor & Heavens (1996)

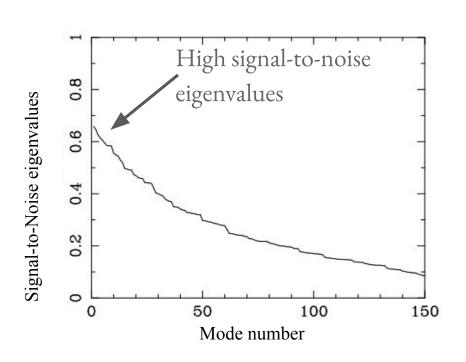
Begin, Liu & Gorce (in prep)

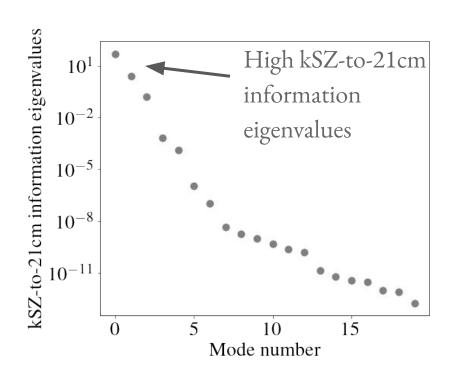




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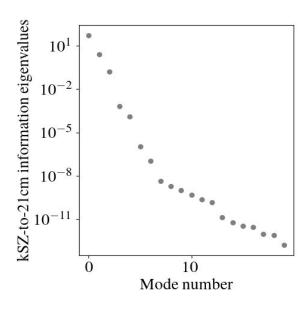
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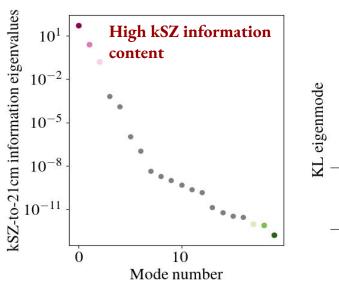


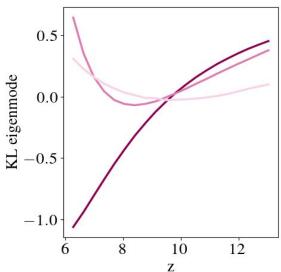


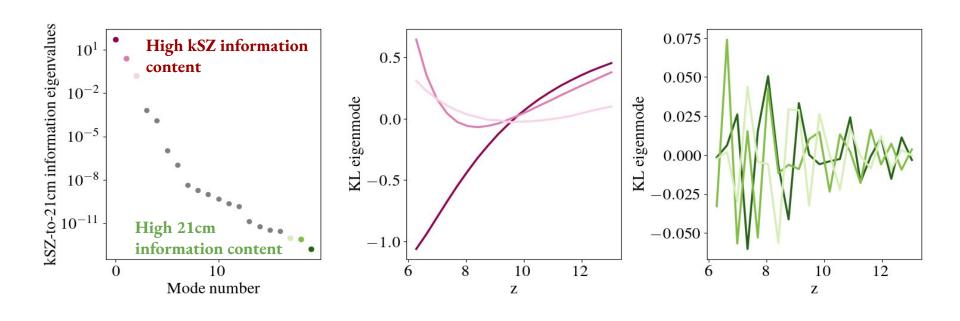
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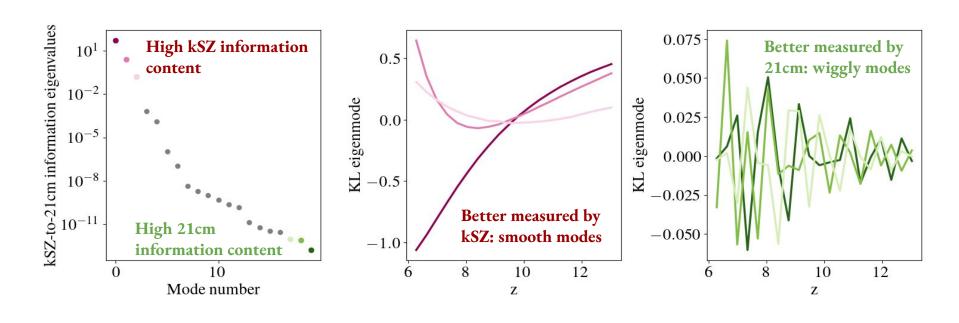
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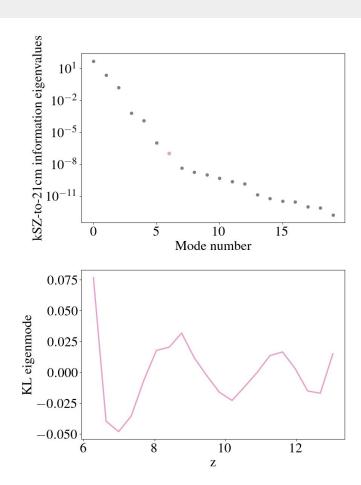


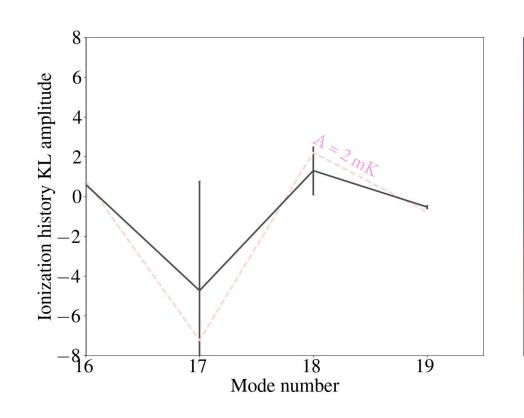


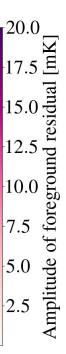


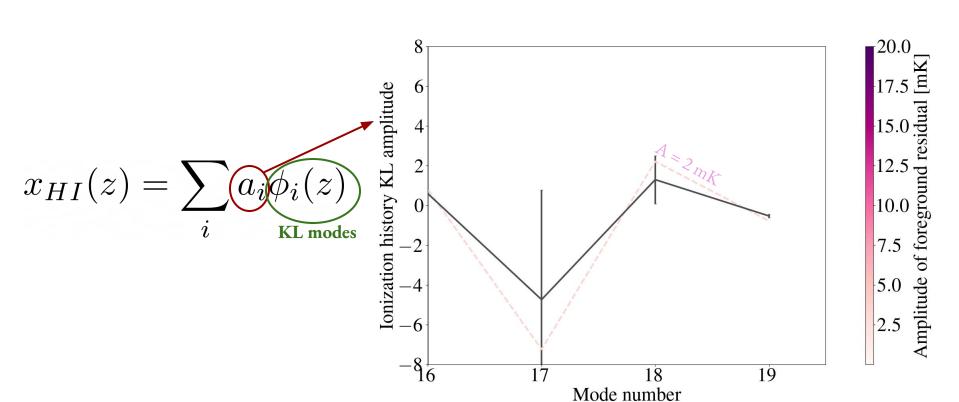
Overlap modes

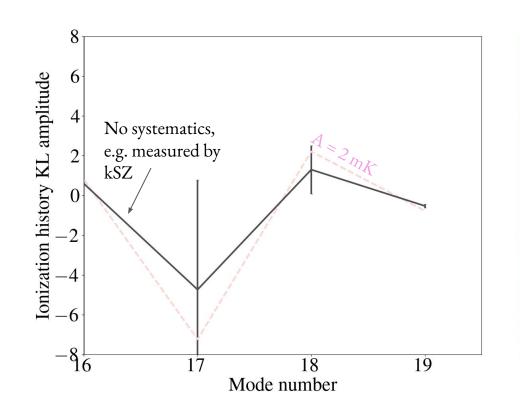
- Some modes are well measured by both signals
- We can use these common modes to perform consistency checks, allowing us detect and project out systematics

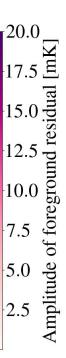










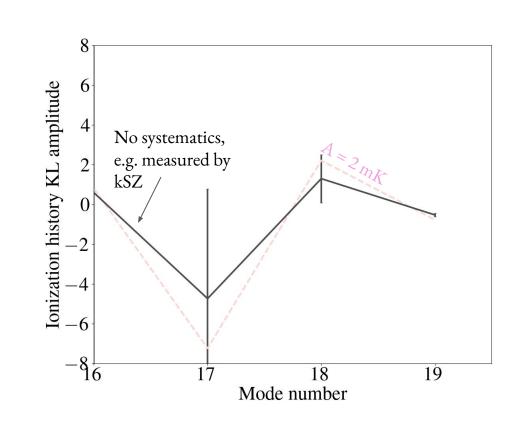


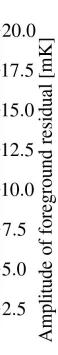
Model foregrounds as power law

$$T_{fg} = A \left(\frac{\nu}{\nu_{\star}}\right)^{\alpha}$$

Add to ionization history, do KL transform

$$KL(x_{HI} + x_{fg})$$



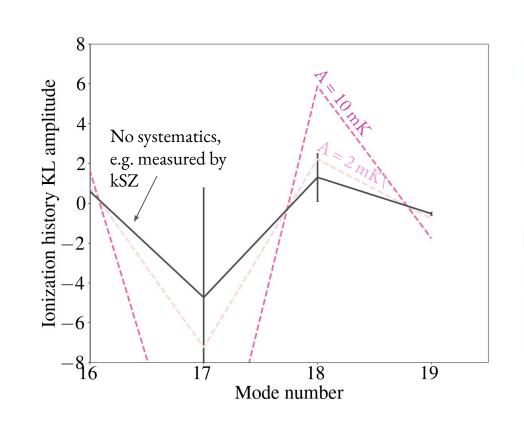


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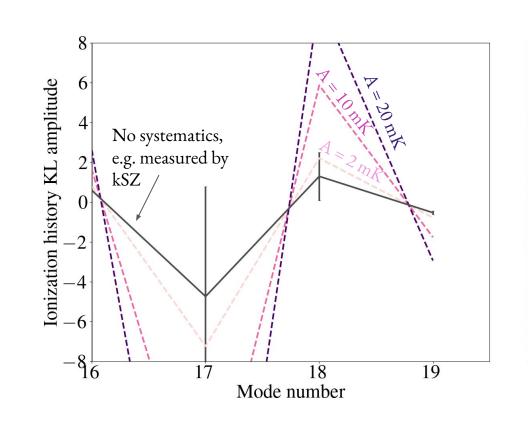
17.5 15.0 12.5 10.0 12.5 10.0 7.5 5.0 2.5 2.5 2.5

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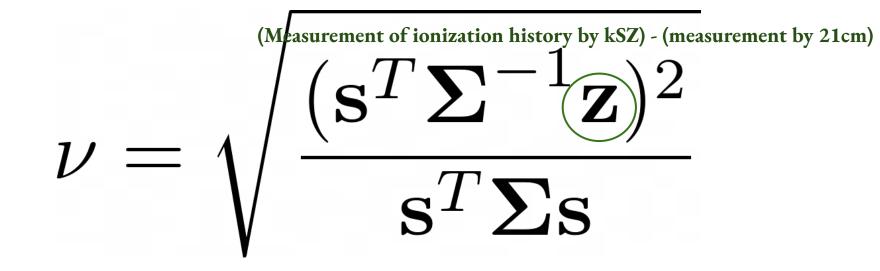
The Linear Matched Filter

Used to detect presence of template shape in data

$$\nu = \sqrt{\frac{(\mathbf{s}^T \mathbf{\Sigma}^{-1} \mathbf{z})^2}{\mathbf{s}^T \mathbf{\Sigma} \mathbf{s}}}$$

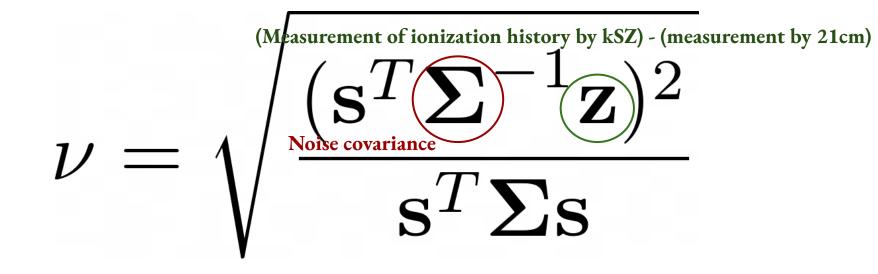
The Linear Matched Filter

■ Used to detect presence of template shape in data



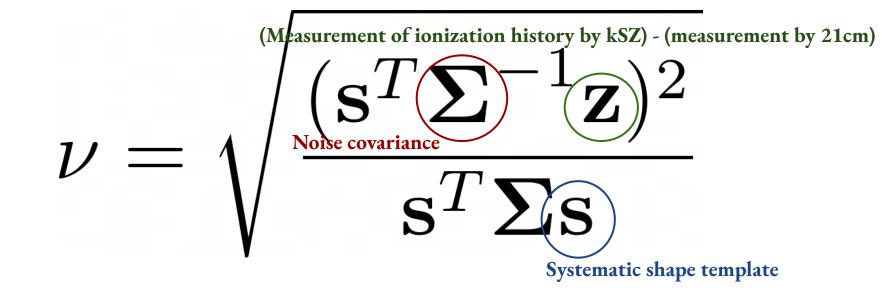
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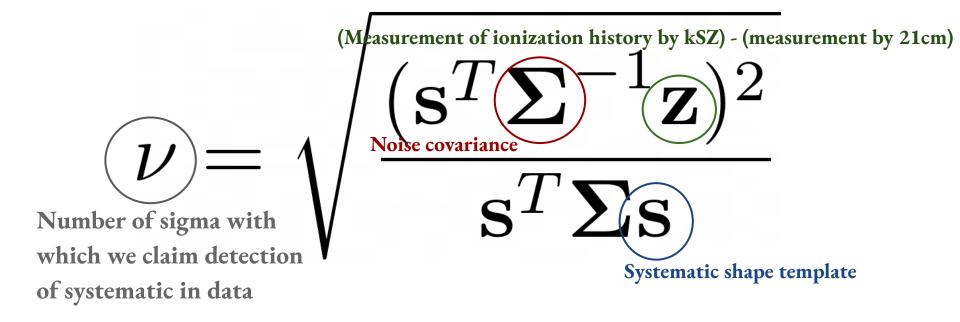
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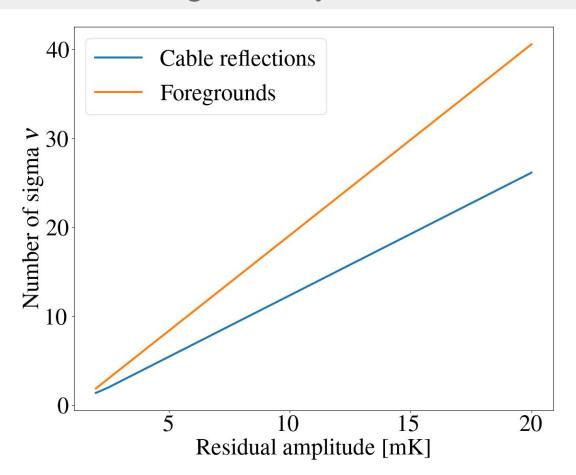


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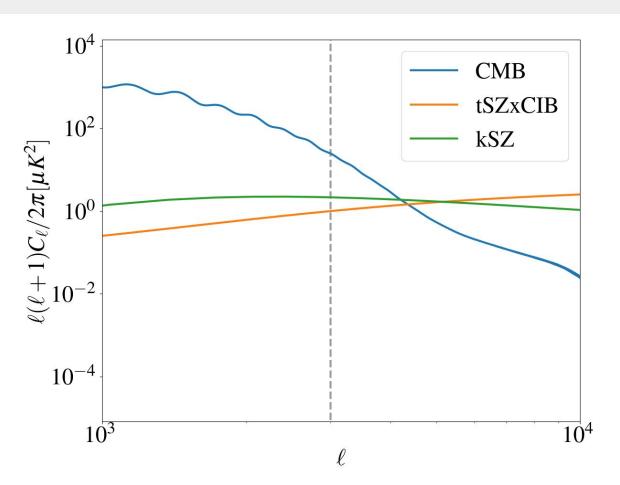
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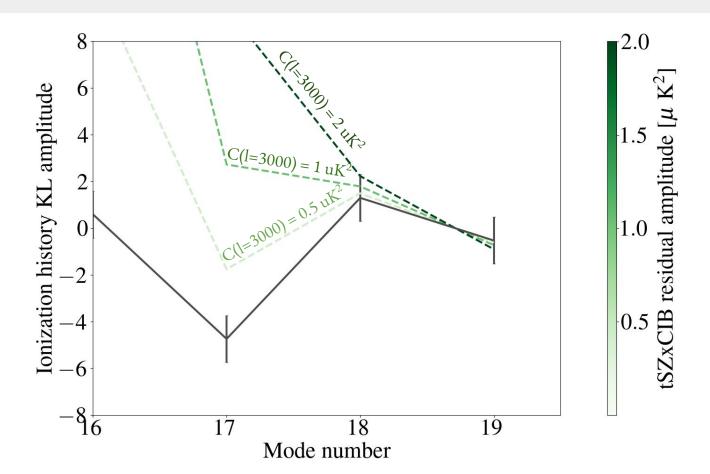
Cable reflection and foreground systematics



CMB and tSZxCIB residuals



CMB and tSZxCIB residuals



Conclusions

- The Karhunen-Loève basis highlights the complementary relation between the kSZ and 21cm global signal.
- By performing statistical tests on modes that are well measured by both probes, we can detect to the presence of systematics.
- This is a general framework that can be **generalized to any two probes**.

BACKUP SLIDES

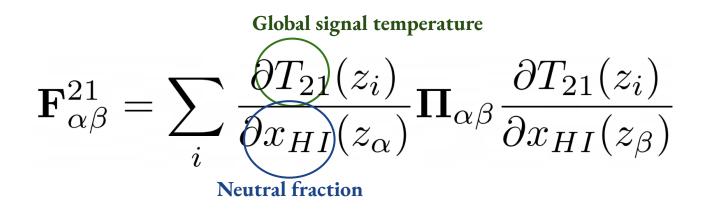
■ We use the Fisher information matrix formalism

$$\mathbf{F}_{\alpha\beta}^{21} = \sum_{i} \frac{\partial T_{21}(z_i)}{\partial x_{HI}(z_\alpha)} \mathbf{\Pi}_{\alpha\beta} \frac{\partial T_{21}(z_i)}{\partial x_{HI}(z_\beta)}$$

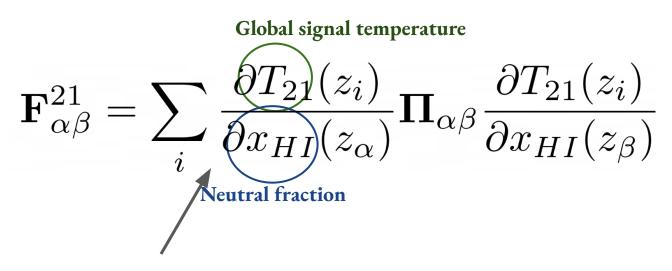
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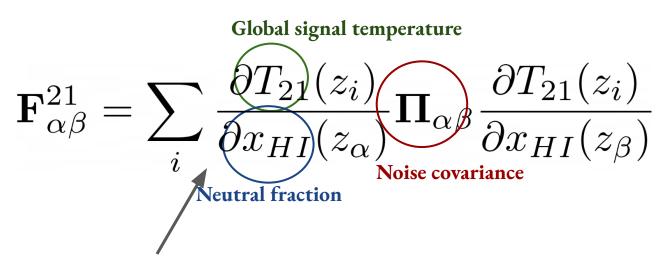
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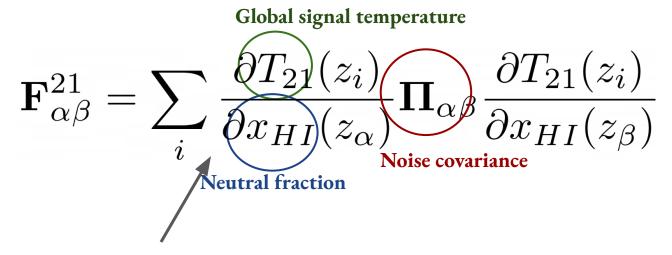
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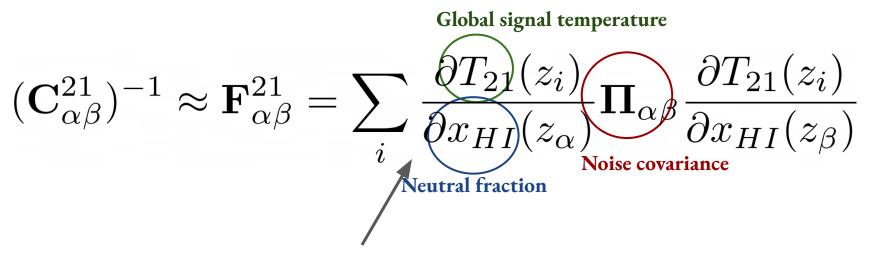
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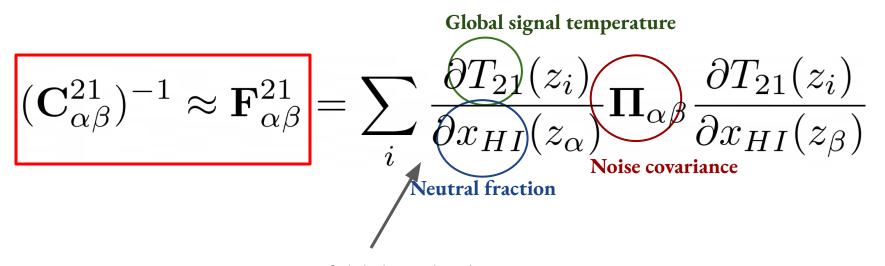
- We use the Fisher information matrix formalism
- Entries of Fisher matrix tell us how good 21cm signal is at constraining the ionized fraction in a redshift bin



■ We use the Fisher information matrix formalism



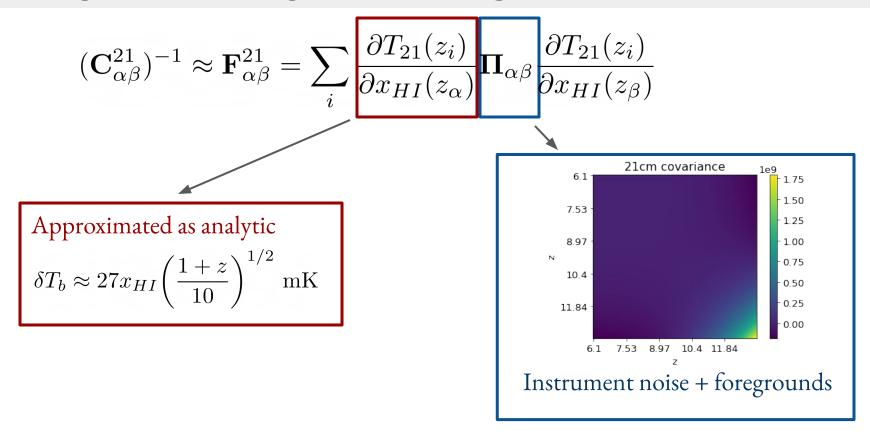
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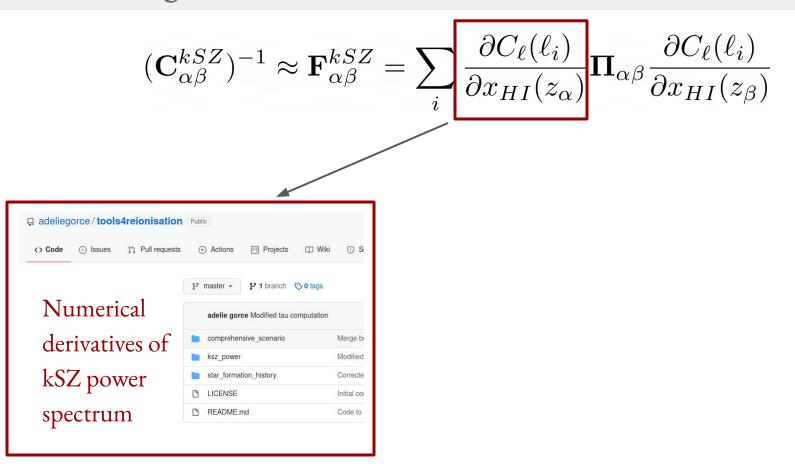
$$(\mathbf{C}_{\alpha\beta}^{21})^{-1} \approx \mathbf{F}_{\alpha\beta}^{21} = \sum_{i} \frac{\partial T_{21}(z_i)}{\partial x_{HI}(z_\alpha)} \mathbf{\Pi}_{\alpha\beta} \frac{\partial T_{21}(z_i)}{\partial x_{HI}(z_\beta)}$$

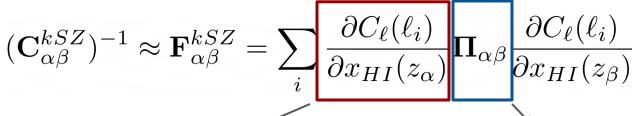
$$(\mathbf{C}_{\alpha\beta}^{21})^{-1} \approx \mathbf{F}_{\alpha\beta}^{21} = \sum_{i} \frac{\partial T_{21}(z_{i})}{\partial x_{HI}(z_{\alpha})} \mathbf{\Pi}_{\alpha\beta} \frac{\partial T_{21}(z_{i})}{\partial x_{HI}(z_{\beta})}$$

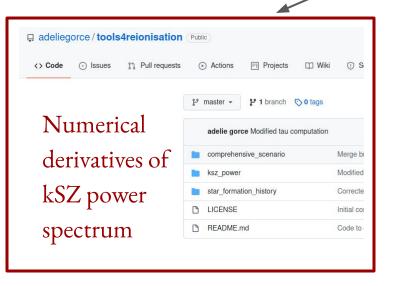
Approximated as analytic
$$\delta T_b \approx 27 x_{HI} \left(\frac{1+z}{10}\right)^{1/2} \, \mathrm{mK}$$

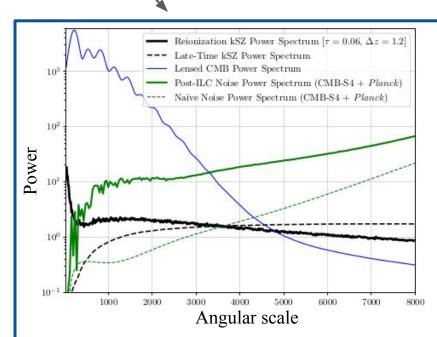


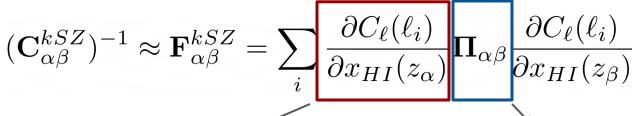
$$(\mathbf{C}_{\alpha\beta}^{kSZ})^{-1} \approx \mathbf{F}_{\alpha\beta}^{kSZ} = \sum_{i} \frac{\partial C_{\ell}(\ell_{i})}{\partial x_{HI}(z_{\alpha})} \mathbf{\Pi}_{\alpha\beta} \frac{\partial C_{\ell}(\ell_{i})}{\partial x_{HI}(z_{\beta})}$$

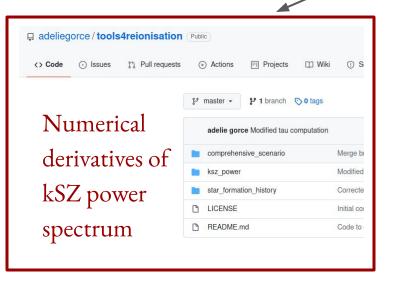


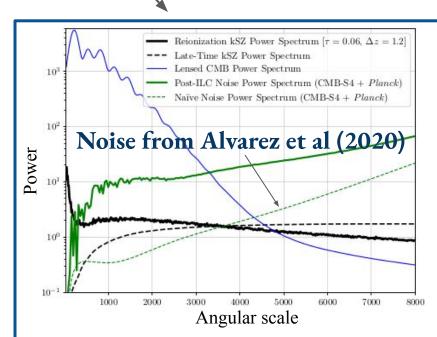












$$egin{aligned} \overline{\mathbf{C}}_{kSZ} &= \mathbf{\Psi}^T \mathbf{L}_{21}^{-1} \mathbf{C}_{kSZ} \mathbf{L}_{21}^{-T} \mathbf{\Psi} = \mathbf{\Psi}^T \mathbf{G} \mathbf{\Psi} = \mathbf{\Lambda} \ \overline{\mathbf{C}}_{21} &= \mathbf{\Psi}^T \mathbf{L}_{21}^{-1} \mathbf{C}_{21} \mathbf{L}_{21}^{-T} \mathbf{\Psi} = \mathbf{\Psi}^T \mathbf{\Psi} = \mathbf{I} \end{aligned}$$

$$\overline{\mathbf{C}}_{21} = \mathbf{\Psi}^T \mathbf{L}_{21}^{-1} \mathbf{C}_{21} \mathbf{L}_{21}^{-T} \mathbf{\Psi} = \mathbf{\Psi}^T \mathbf{\Psi} = \mathbf{I}$$

$$\mathbf{L}_{21}^{-1}\mathbf{C}_{kSZ}\mathbf{L}_{21}^T\mathbf{v} = \lambda \mathbf{L}_{21}^T\mathbf{v}.$$

 $\mathbf{C}_{kSZ}\mathbf{v} = \lambda \mathbf{C}_{21}\mathbf{v}.$