

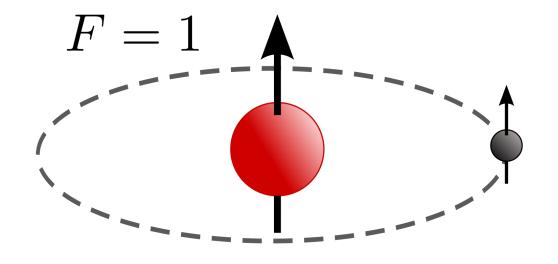
# Weighing neutrinos with 21cm intensity mapping at the SKAO

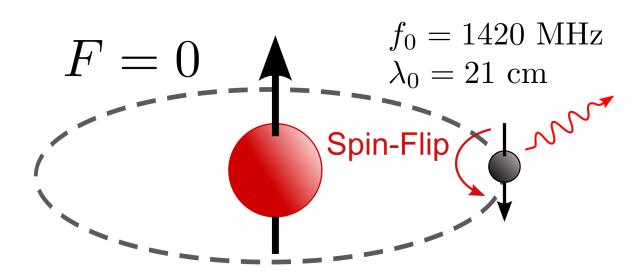
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Based on: arxiv 2504.18625, GA, M. Berti, B.S. Haridasu, M. Spinelli, M. Viel

## 21CM LINE - INTRODUCTION

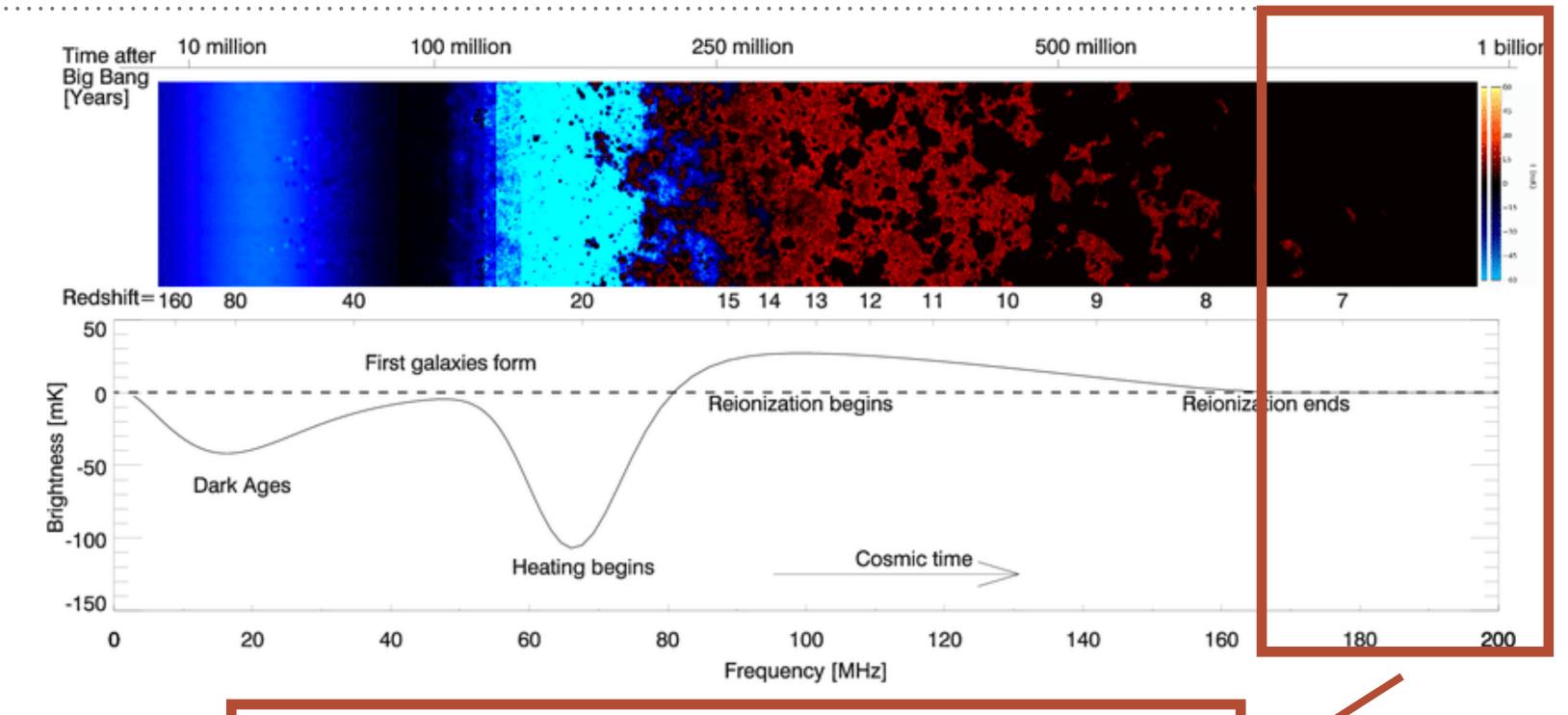
[Pritchard & Loeb, 2011]





#### Neutral hydrogen (HI) line:

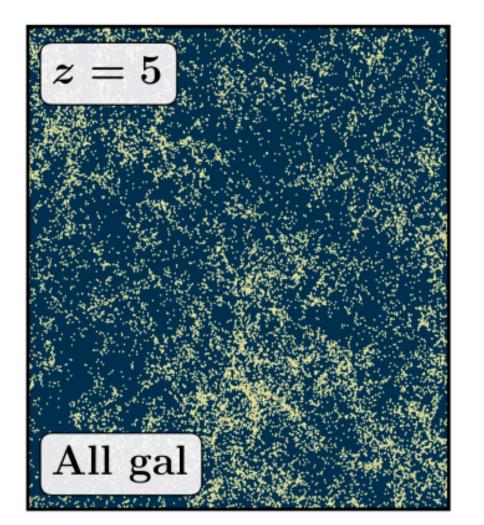
hyperfine splitting of the ground state due to interaction between the magnetic moments of proton and electron.

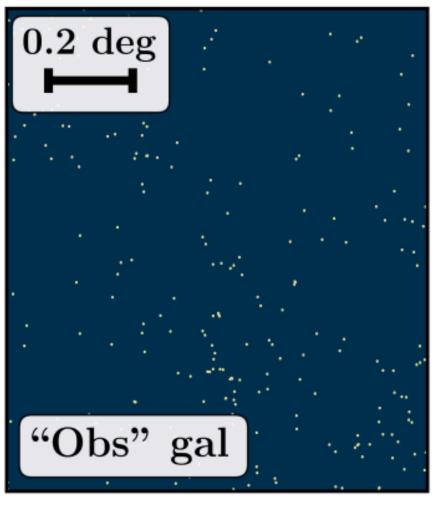


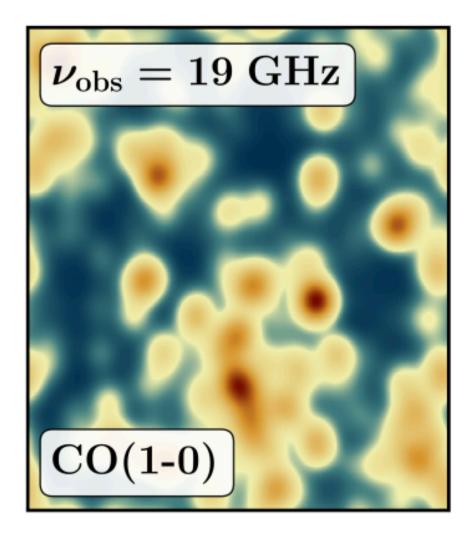
#### After reionization:

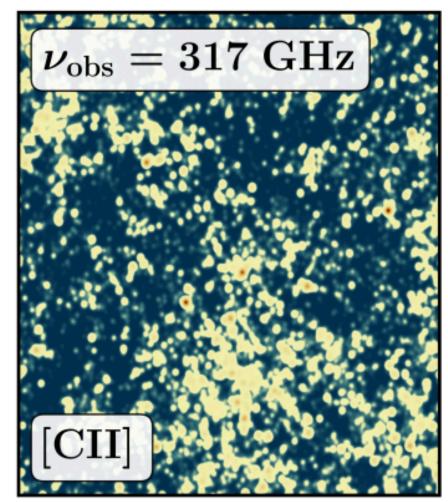
- \* Most of the HI is in galaxies and in the intergalactic medium (IGM).
- \* The 21cm signal is a biased tracer of the underlying matter field.

## LINE-INTENSITY MAPPING — THEORY





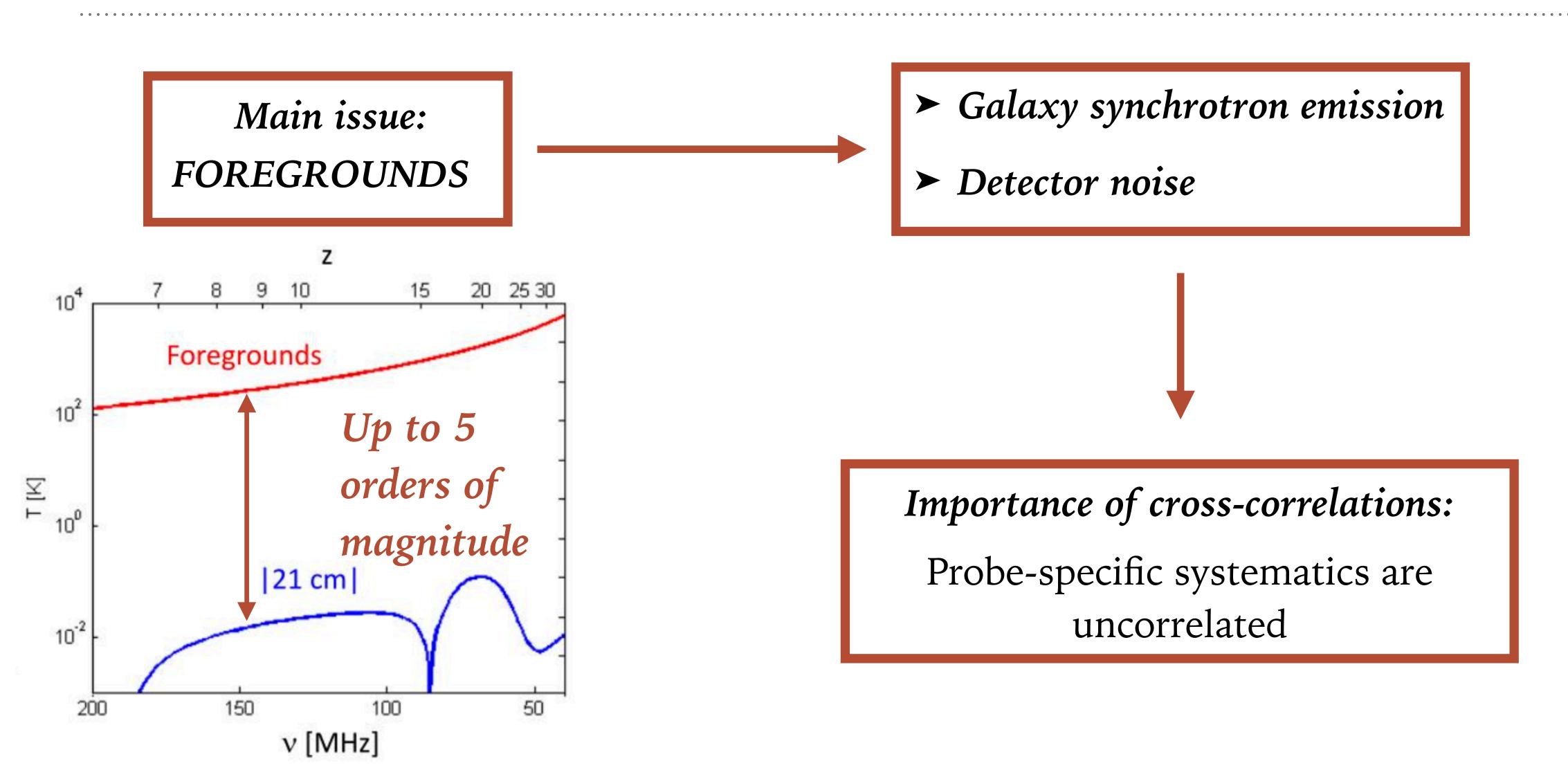




[Bernal + 2022]

- ➤ Intensity mapping (IM): measures the integrated emission from individually unresolved galaxies and intergalactic medium (IGM).
- > Probes large volumes quickly —— high potential of cross-correlations.

## 21CM LINE - FOREGROUNDS



## METHODOLOGY

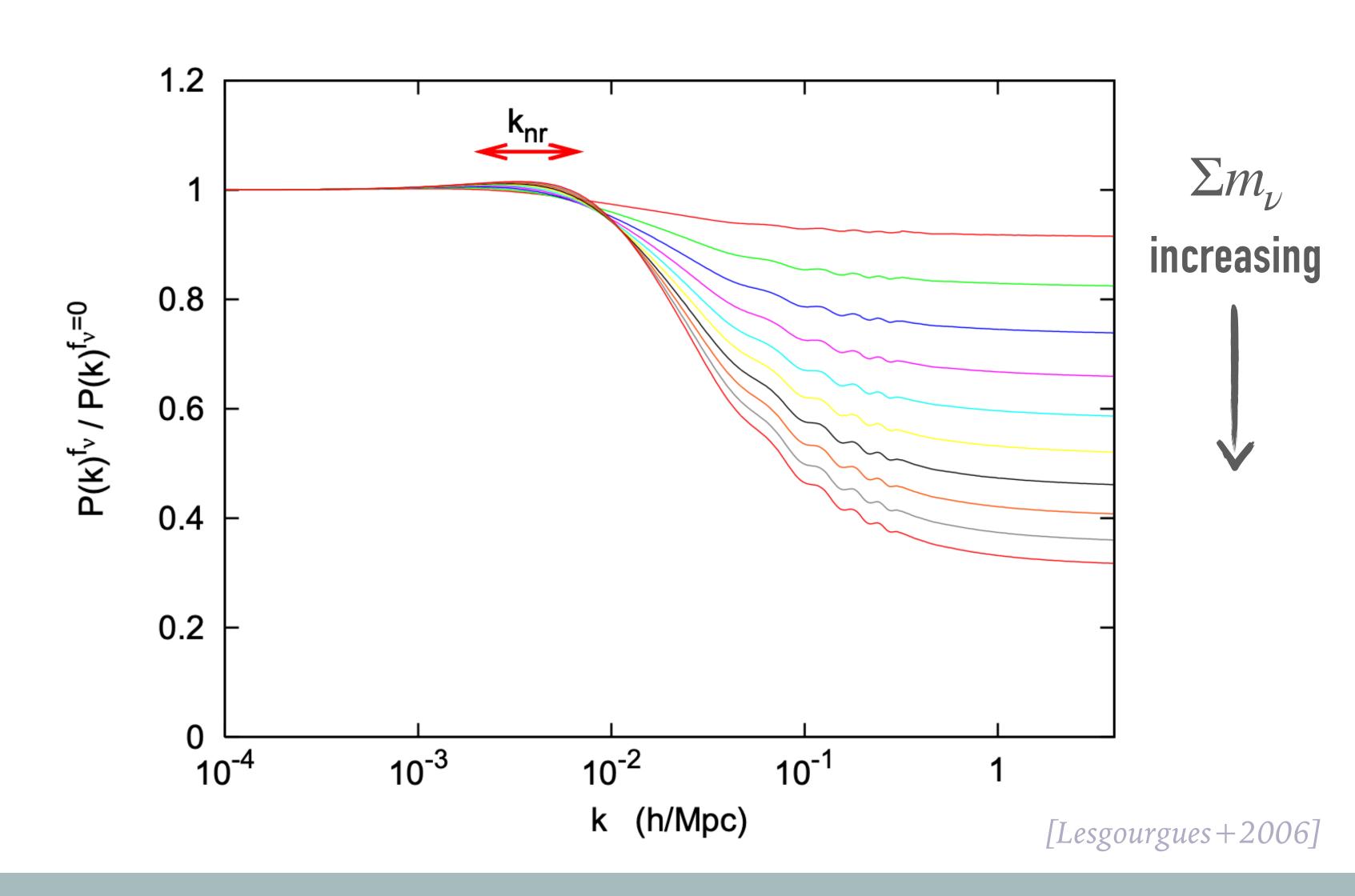
The goal is to forecast the constraining power of future 21cm intensity mapping with the SKAO on the sum of neutrino masses,  $\Sigma m_{\nu}$ .

➤ Build **synthetic data sets** that mimic realistic observations that will be possible with the SKAO and forecast the constraining power with a Bayesian analysis.

➤ We focus on 21cm IM auto-power spectrum and 21cm IM - galaxies cross-correlation power spectrum measurements.

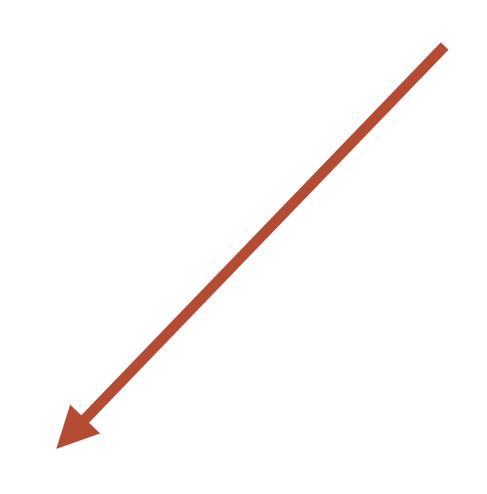
## MASSIVE NEUTRINOS IN COSMOLOGY

Free-streaming: neutrinos freestream out of high-density regions suppressing perturbations on small scales. The resulting suppression in the matter power spectrum provides a way to constrain neutrino masses.



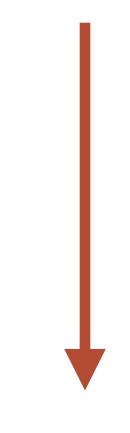
## FORECASTING - SURVEYS TO BUILD SYNTHETIC DATA SETS

Surveys to build synthetic data sets



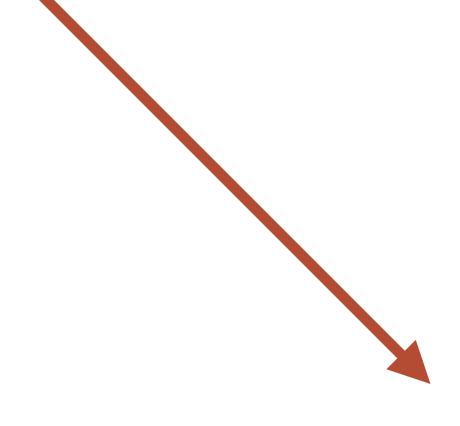
IM surveys

SKA1-Mid surveys in redshift range 0 < z < 3



Euclid-like survey

Euclid-like survey in redshift range



DESI-like survey

DESI ELG- like survey in redshift range 0.7 < z < 1.7

## FORECASTING - BUILDING THE SYNTHETIC DATA SETS

For each redshift bin, the survey specifications fix the range of accessible scales

#### z-bin volume

$$V_{\text{bin}}(z_c) = \Omega_{\text{sur}} \int_{z-\Delta z/2}^{z+\Delta z/2} dz' \frac{cr(z')^2}{H(z')}$$



$$k_{\min}(z_c) = \frac{2\pi}{\sqrt[3]{V_{\min}(z_c)}}$$

#### Dimension of the telescope beam

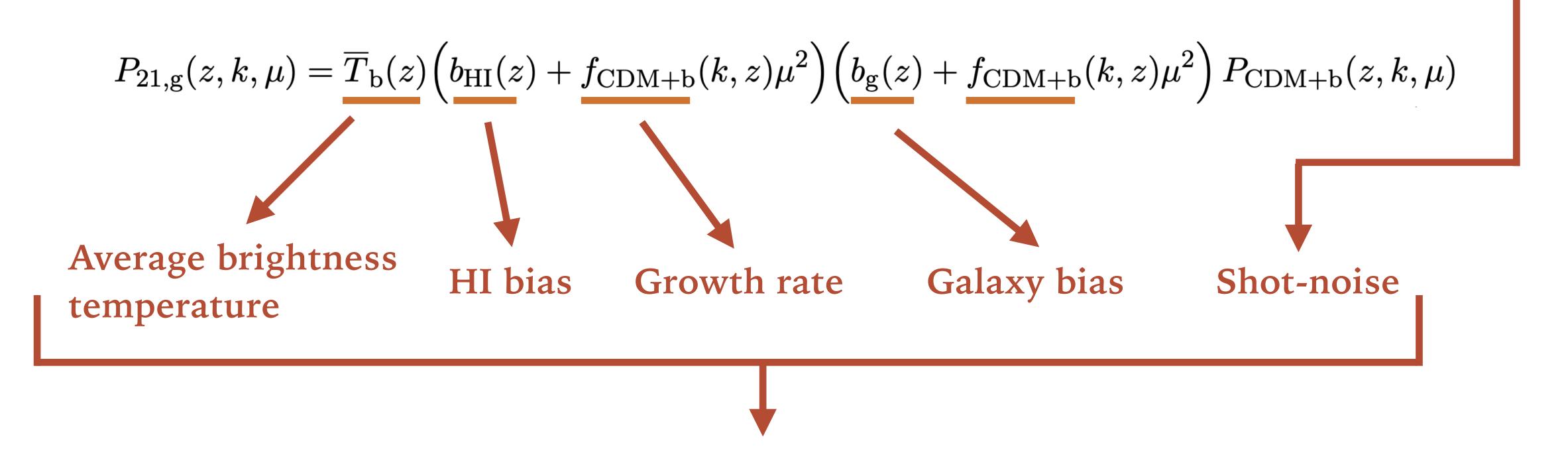
$$R_{\text{beam}}(z_c) = \frac{\theta_{\text{FWHM}}}{2\sqrt{2\ln 2}} r(z_c)$$



$$k_{\max}(z_c) = \frac{2\pi}{R_{\text{beam}}(z_c)}$$

## FORECASTING - POWER SPECTRUM MODEL

$$P_{21}(z, k, \mu) = \overline{T}_{b}^{2}(z) \left[ b_{HI}(z) + f_{CDM+b}(z, k) \mu^{2} \right]^{2} P_{CDM+b}(z, k) + P_{SN}(z)$$



Nuisance parameters (either computed from theory or sampled)

## FORECASTING - MIMICKING REALISTIC OBSERVATIONS

To mimic realistic observations we add two factors:

#### Gaussian beam smoothing

$$\tilde{B}(z, k, \mu) = \exp\left[\frac{-k^2 R_{\text{beam}}^2(z)(1 - \mu^2)}{2}\right]$$

Suppresses the power on small scales

#### Alcock-Paczynski (AP) effect

$$\alpha_{\perp}(z) = \frac{D_A(z)}{D_A^{\text{fid}}(z)}$$
 and  $\alpha_{\parallel}(z) = \frac{H^{\text{fid}}(z)}{H(z)}$ 

$$q = \frac{k}{\alpha_{\perp}} \sqrt{1 + \mu^2 \left(\frac{\alpha_{\perp}^2}{\alpha_{\parallel}^2} - 1\right)} \qquad \nu = \frac{\alpha_{\perp} \mu}{\alpha_{\parallel} \sqrt{1 + \mu^2 \left(\frac{\alpha_{\perp}^2}{\alpha_{\parallel}^2} - 1\right)}}$$

## FORECASTING - BUILDING THE SYNTHETIC DATA SETS - ERRORS

#### 21cm IM variance

#### Instrument noise

$$\sigma_{21}(z,k,\mu) = \frac{\hat{P}_{21}(z,k,\mu) + P_{N}(z)}{N_{\text{modes}}(z,k,\mu)} \qquad P_{N}(z) = \frac{T_{\text{sys}}^{2} 4\pi f_{\text{sky}}}{N_{\text{dish}} t_{\text{obs}} \delta \nu} \frac{V_{\text{bin}}(z)}{\Omega_{\text{sur}}}$$

#### cross-correlation variance

$$\sigma_{21,g}(z,k) = \frac{1}{\sqrt{2N_{\text{modes}}(z,k)}} \sqrt{\hat{P}_{21,g}^2(z,k) + \left(\hat{P}_{21}(z,k) + P_{\text{N}}(z)\right) \left(\hat{P}_{\text{g}}(z,k) + \frac{1}{\overline{n}_{\text{g}}}(z)\right)}$$

#### Number of modes per k bin

$$N_{
m modes}(z,k) = rac{k^2 \Delta k(z_c)}{4\pi^2} V_{
m bin}(z_c)$$

## FORECASTING - FULL POWER SPECTRUM MODEL

Adding everything you get the **21cm** (observed) power spectrum model and similar for the cross

$$\hat{P}_{21}(z, k, \mu) = \frac{1}{\alpha_{\perp}^2 \alpha_{\parallel}} \tilde{B}^2(z, q, \nu) P_{21}(z, q, \nu) + P_{N}(z)$$

> We can decompose the power spectrum using Legendre polynomials

$$\hat{P}_{X,\ell}(z,k) = \frac{(2\ell+1)}{2} \int_{-1}^{1} d\mu \, \mathcal{L}_{\ell}(\mu) \hat{P}_{X}(z,k,\mu) \qquad \begin{array}{c} \text{Power spectrum} \\ \text{multipoles} \end{array}$$

$$C_{\ell,\ell'}(z,k) = \frac{(2\ell+1)(2\ell'+1)}{2} \int_{-1}^{1} d\mu \, \mathcal{L}_{\ell}(\mu) \, \mathcal{L}_{\ell'}(\mu) \, \sigma^2(z,k,\mu)$$
 Multipoles covariance

## FORECASTING - DATA ANALYSIS

We consider three values for  $\Sigma m_{\nu} = 0.06$ , 0.1, 0.4 eV, meaning that in total we have 9 data sets (3 data sets for each of the 21cm IM auto-power spectrum, the SKAO×DESI cross-correlation and the SKAO×Euclid cross-correlation)

➤ Additionally, we combine our mock data sets with Planck 2018.

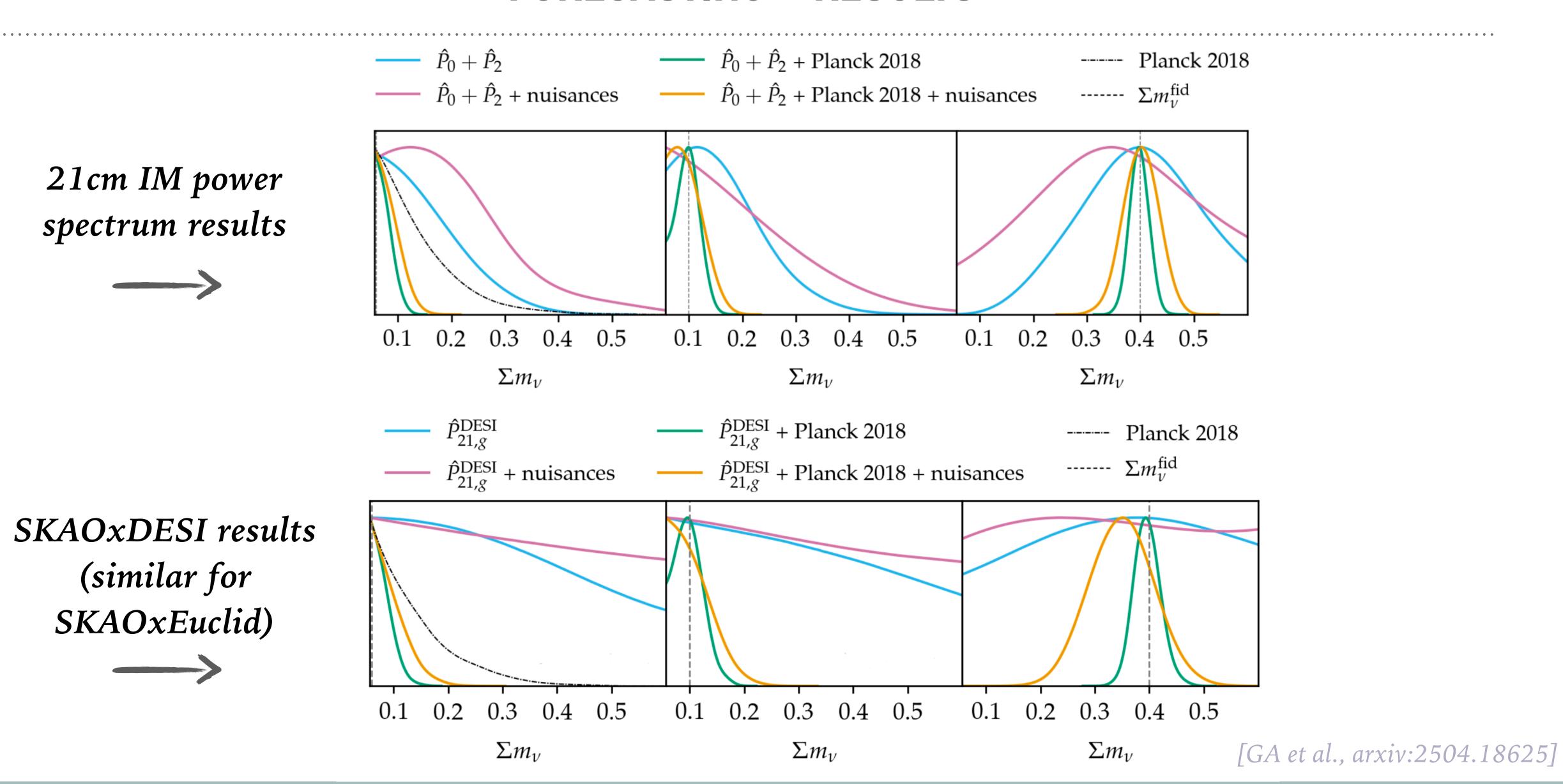
#### Gaussian likelihood

$$-\ln\left[\mathcal{L}\right] = rac{1}{2}\sum_{z}\Delta\Theta(z)^{\mathrm{T}}C^{-1}(z)\Delta\Theta(z)$$
 
$$\Delta\Theta(z) = \Theta^{\mathrm{th}}(z) - \Theta^{\mathrm{obs}}(z)$$

## LIKELIHOOD FUNCTION

- ➤ We built on Maria's works [Berti+2021, Berti+2022, Berti+2023] and built a *Gaussian likelihood code* fully integrated with the MCMC sampler **Cobaya**.
- ➤ This likelihood code, called **topk**, will be made *publicly available* on Maria's GitHub page at **github.com/mberti94**
- The likelihood code can handle the computation of:
- 21cm IM power spectrum multipoles
- 21cm and galaxies cross-correlation power spectrum multipoles

## FORECASTING - RESULTS



## FORECASTING - RESULTS

Likelihoods	$\Sigma m_{ u}^{ m fid} = 0.06{ m eV}$	$\Sigma m_{ u}^{ m fid} = 0.1{ m eV}$	$\Sigma m_{ u}^{ m fid} = 0.4{ m eV}$
$\hat{P}_0 + \hat{P}_2$	< 0.287	< 0.317	$0.41^{+0.11}_{-0.14}$
+ nuisances	< 0.425	< 0.452	$0.34^{+0.16}_{-0.14}$
Planck 2018			
$+\;\hat{P}_0+\hat{P}_2$	< 0.105	$0.098 \pm 0.022$	$0.398 \pm 0.018$
+ nuisances	< 0.126	< 0.151	$0.401 \pm 0.034$
Planck 2018			
$+~\hat{P}_{21,\mathrm{g}}^{\mathrm{DESI}}$	< 0.116	$0.099^{+0.020}_{-0.033}$	$0.396^{+0.023}_{-0.026}$
+ nuisances	< 0.155	< 0.177	$0.349 \pm 0.060$
Planck 2018			
$+~\hat{P}^{ m Euclid}_{21,{ m g}}$	< 0.117	$0.100^{+0.021}_{-0.032}$	$0.397^{+0.023}_{-0.026}$
+ nuisances	< 0.156	< 0.180	$0.343 \pm 0.062$

- rianglerighter Cross-correlation data alone doesn't hold enough constraining power to improve the state of the art  $\sum m_{\nu}$ .
- When combined with complementary
  CMB data, gives constraints
  comparable to the ones obtained with
  auto-power.

[GA et al., arxiv:2504.18625]

# Thank you for your attention!

## Extra slides

## THE 21CM POWER SPECTRUM

 $ightharpoonup \overline{T}_b$  is the averaged brightness temperature of HI, can be computed as

$$\bar{T}_{b}(z) = 180 \ \Omega_{HI}(z) \frac{h H_0}{H(z)} (1+z)^2 \text{mK}$$
 [Battye+2013]

 $\blacktriangleright$  The HI density  $\Omega_{\rm HI}$  has mild redshift evolution

$$\Omega_{\rm HI}(z) = 4. \times 10^{-4} (1+z)^{0.6}$$

[Chrighton + 2015]

## THE 21CM POWER SPECTRUM

➤ We have no analytical models for HI bias and shot-noise, therefore we interpolate results from *hydrodynamical simulations* 

z	0	1	2	3	4	5
$b_{ m HI}$	0.84	1.49	2.03	2.56	2.82	3.18
$ \begin{array}{ c c }\hline P_{\rm HI}^{\rm SN} \\ [(h^{-1}{\rm Mpc})^3] \end{array} $	104	124	65	39	14	7

[Villaescusa-Navarro+2018]

The growth rate f(z) and the matter power spectrum  $P_m(z,k)$  are computed with a Boltzmann solver (CAMB) or an emulator. For massive neutrinos they are computed neglecting the contributions of neutrinos.