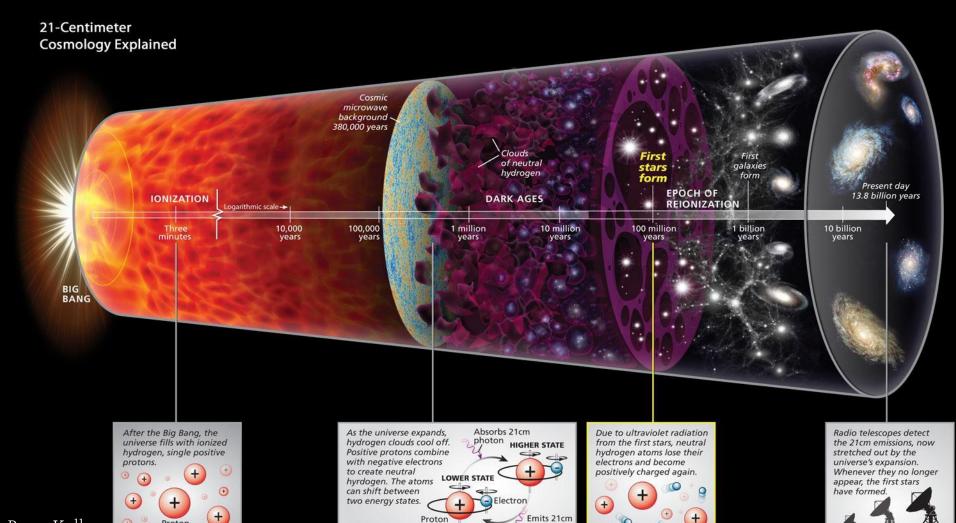


SUPERIORE

Forecasting upcoming 21cm SKA observations with 21cmFAST

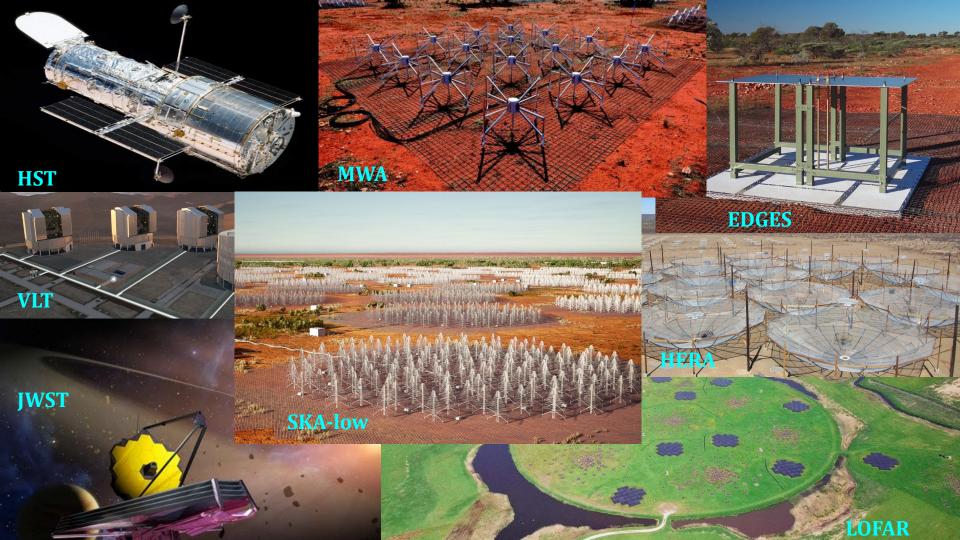
Daniela Breitman Scuola Normale Superiore In collaboration with Prof. Andrei Mesinger 5th SKA Workshop - Bologna November 25, 2025



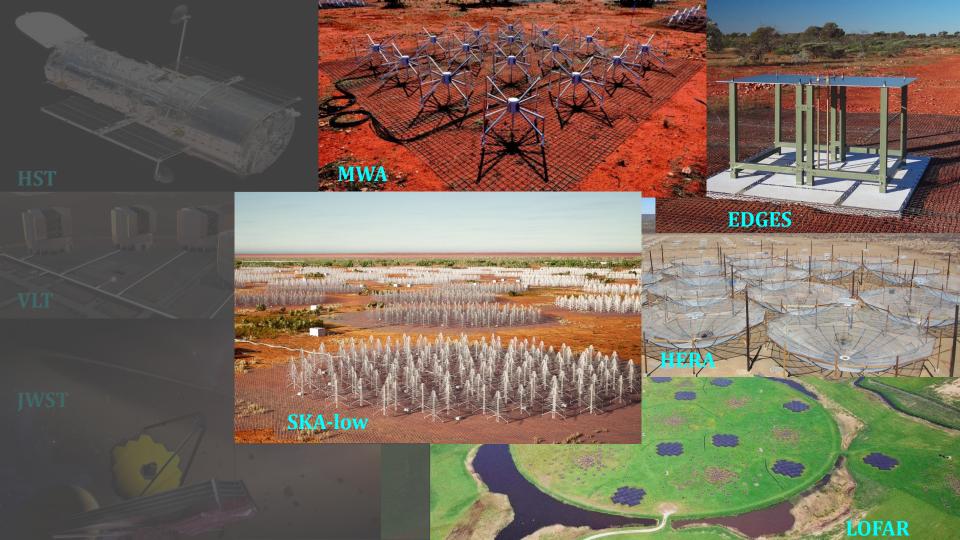
photon

Roen Kelly

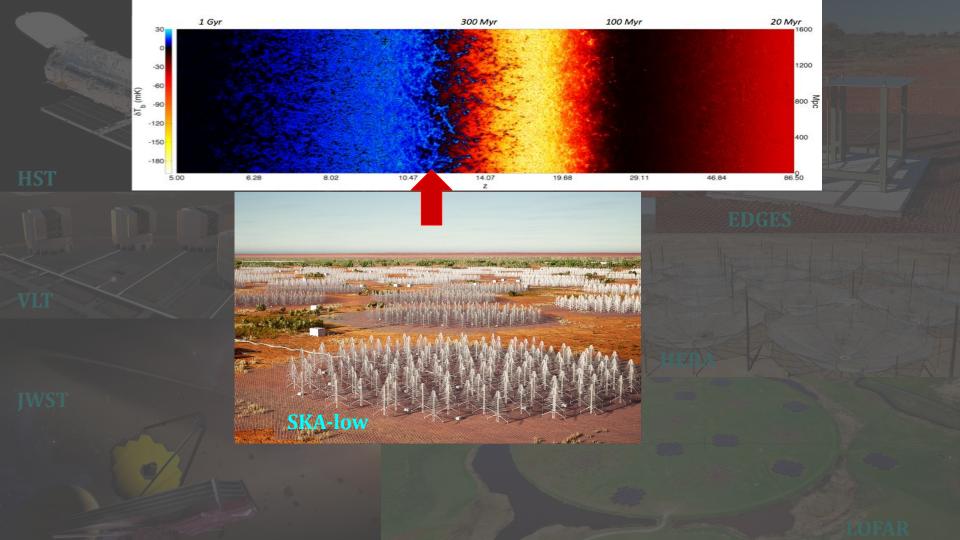
Proton



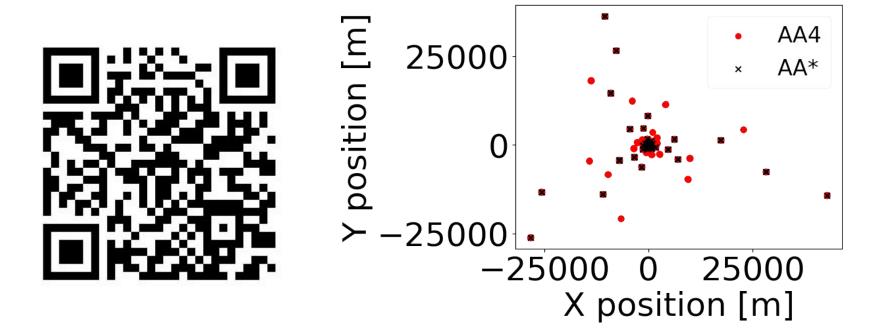


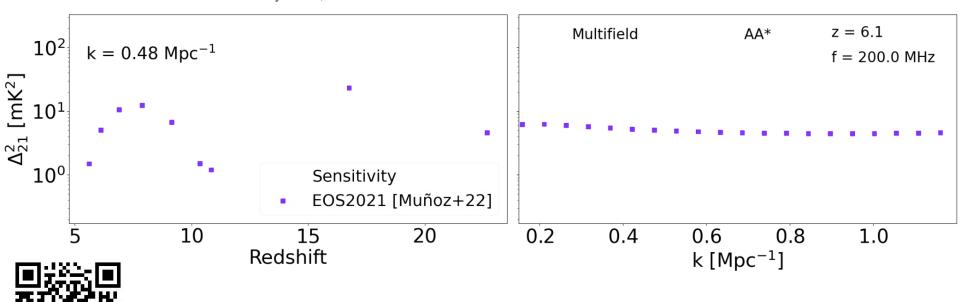


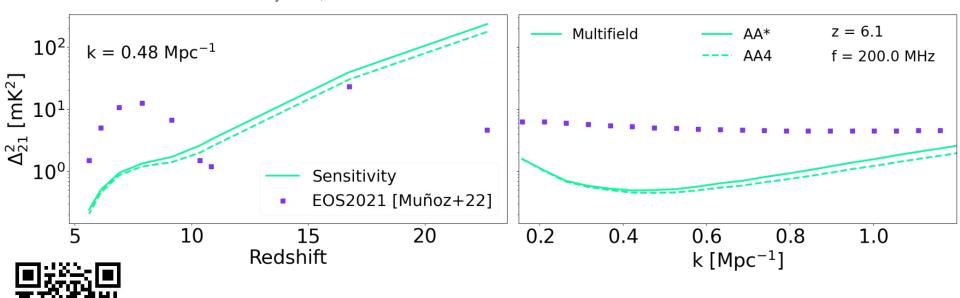


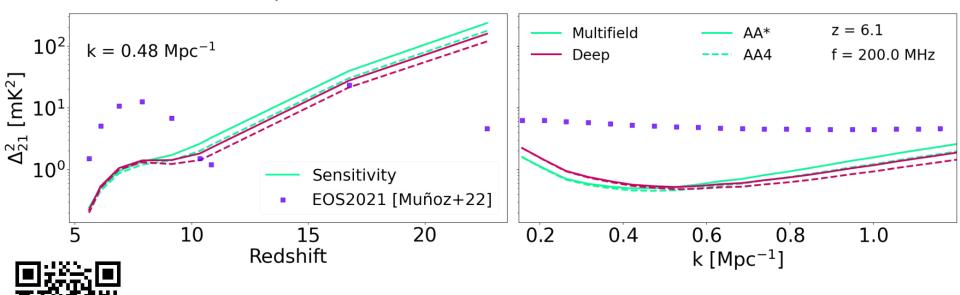


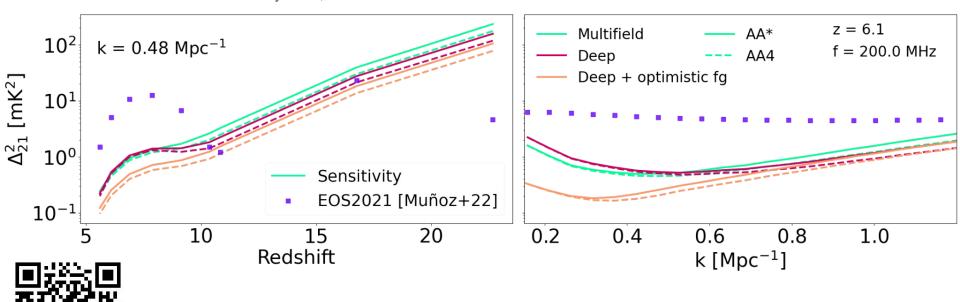




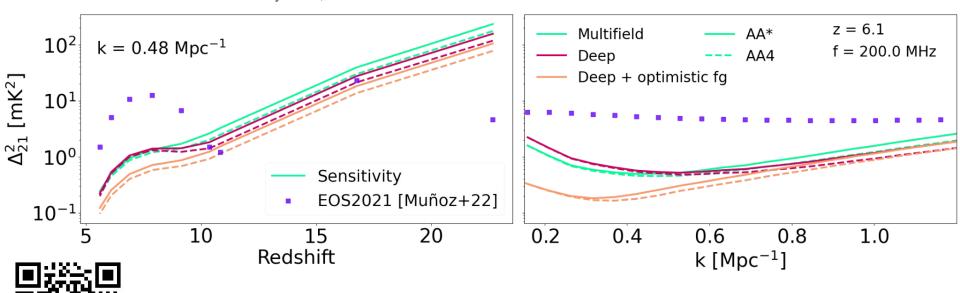






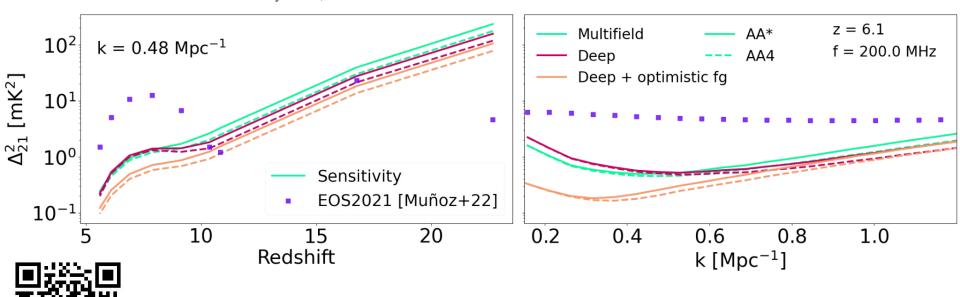


Use 21cmSense [Murray+18, Pober+14] to forecast SKA observations:



How can we learn from 21 cm SKA-low observations?

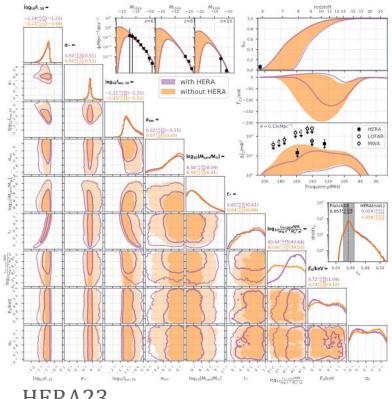
Use 21cmSense [Murray+18, Pober+14] to forecast SKA observations:



How can we learn from 21 cm SKA-low observations? Bayesian inference

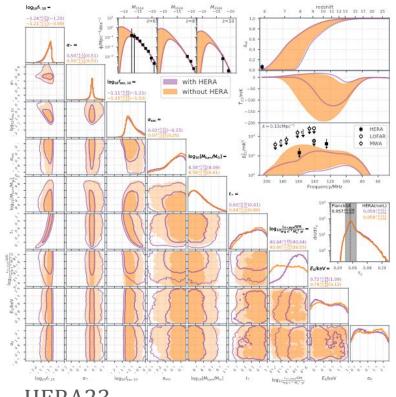
- $P(\theta)$ = prior knowledge on model parameters θ
- Forward model $P(d|\theta)$ = probability of reproducing observation d with model parameters θ $P(\theta|d) = \frac{P(d|\theta)P(\theta)}{P(d)}$
- What we want $P(\theta|d)$ = updated knowledge on θ provided an observation d

Expensive: requires ~ 300k forward model evaluations



HERA23

- Expensive: requires ~ 300k forward model evaluations
- Computationally impractical, even for semi-analytical simulators

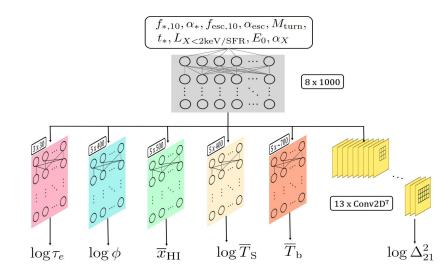


HERA23

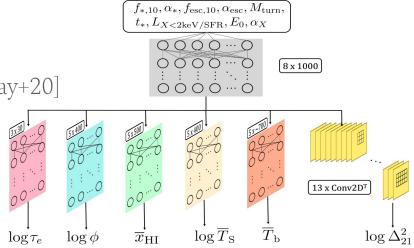
- **Expensive:** requires ~ **300k** forward model evaluations
- Computationally impractical, even for semi-analytical simulators



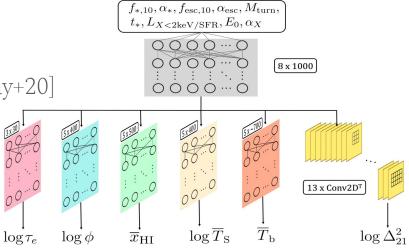
• Input: nine astrophysical parameters



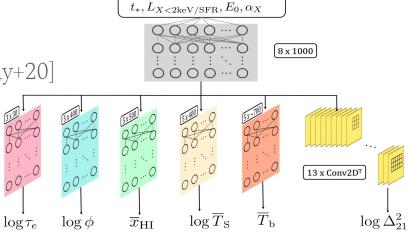
- Input: nine astrophysical parameters
- Model: 21cmFASTv3 [Mesinger+07,11, Murray+20]



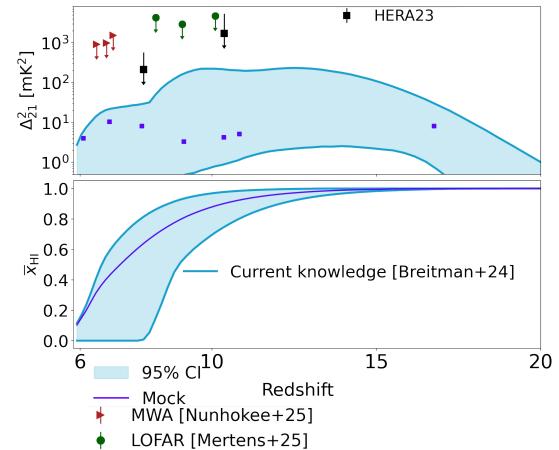
- Input: nine astrophysical parameters
- Model: 21cmFASTv3 [Mesinger+07,11, Murray+20]
- Output: six summary observables

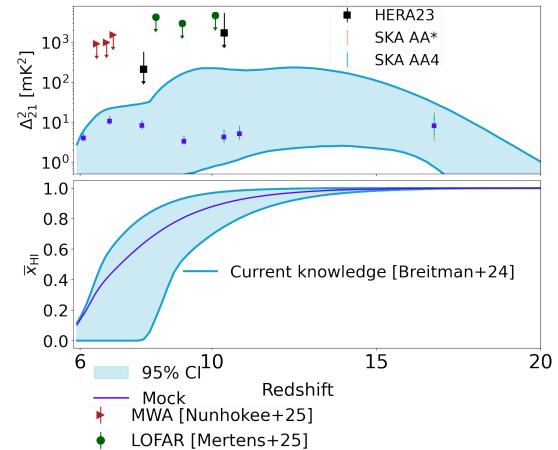


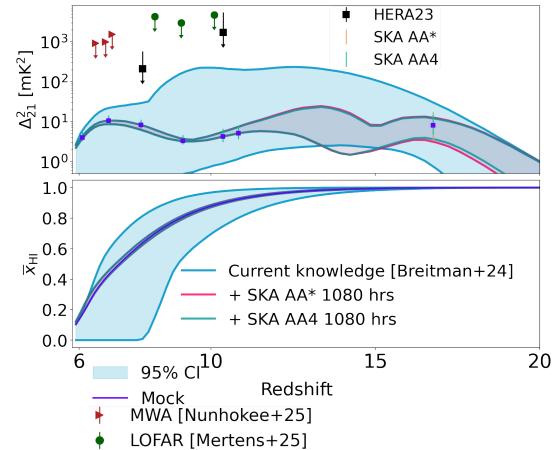
- Input: nine astrophysical parameters
- Model: 21cmFASTv3 [Mesinger+07,11, Murray+20]
- Output: six summary observables

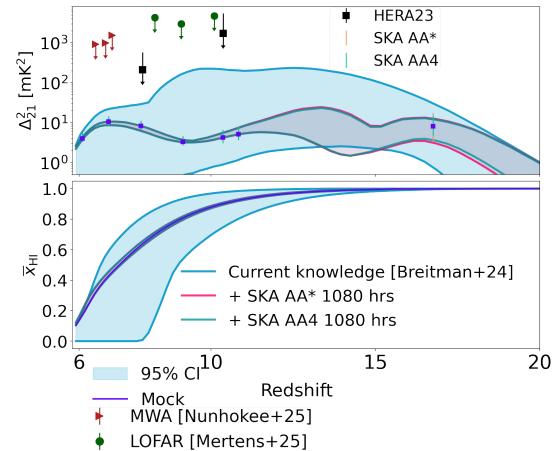


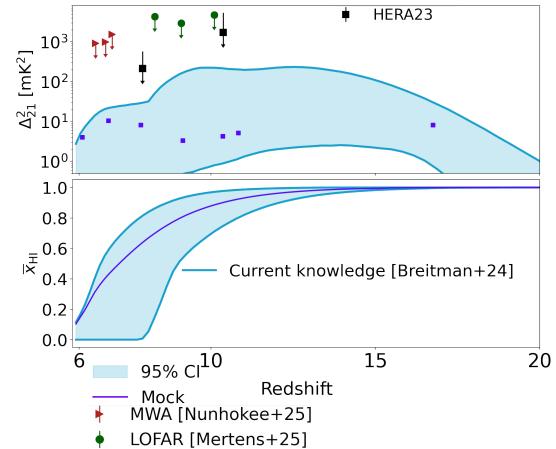
 $f_{*,10}, \alpha_*, f_{\mathrm{esc},10}, \alpha_{\overline{\mathrm{esc}}, M_{\mathrm{turn}}},$

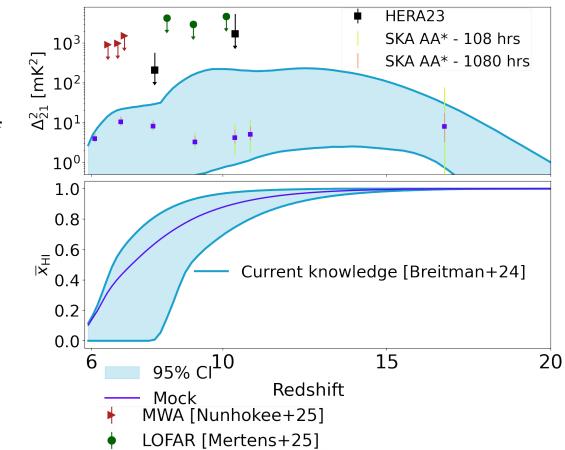


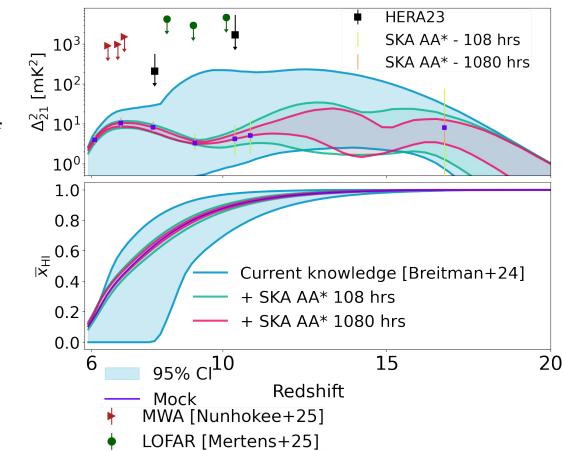




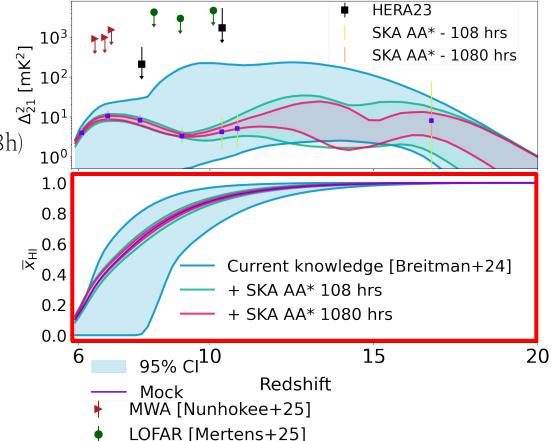




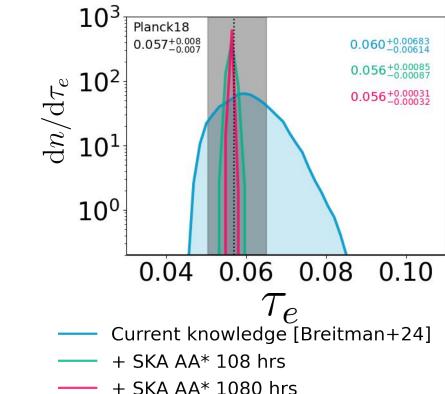




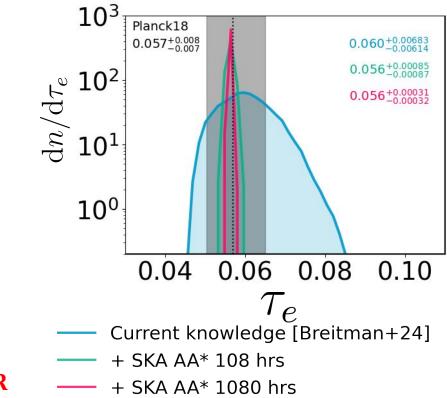
- For CD/EoR science, **AA*** is just as constraining as **AA4**
- **Percent-level EoR**: 1080h (108h) midpoint within ±0.04 (0.02)



- For CD/EoR science, AA* is just as constraining as AA4
- **Percent-level EoR**: 1080h (108h) midpoint within ±0.04 (0.02)
- τ_{e} 1 σ over 20x tighter than Planck18

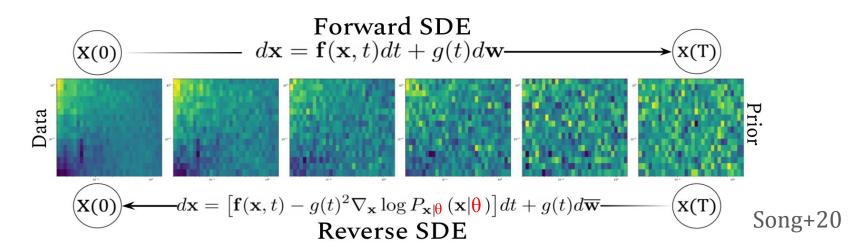


- For CD/EoR science, AA* is just as constraining as AA4
- **Percent-level EoR**: 1080h (108h) midpoint within ±0.04 (0.02)
- τ 1σ over 20x tighter than Planck18



AA* is extremely informative for CD/EoR — science even after only ~100 hrs of observation!

- First emulator of CD/EoR summary observables with **eleven** input parameters
- First emulator of the 2D 21-cm PS
- First score-based generative emulator of the PS



Breitman+25 in prep

Now available!!

pip install py21cmemu

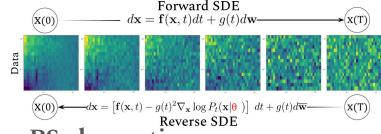


https://github.com/21cmfast/21cmemu





Conclusions



Let's EMUlate!

- Goal: Learn from 21-cm PS observations
- Method: Multi-tracer inference with 21-cm and other EoR/CD probes
 - Emulation makes these inferences computationally cheap
- New 21cmEMUv3:
 - Astrophysical model including two galaxy populations: ACGs and MCGs
 - First emulator of CD/EoR summary observables with eleven input parameters
 - First emulator of the 2D 21-cm PS
 - First score-based generative emulator of the PS
 - FE 68% CL ≤ 5% for all quantities
- Forecasting SKA with 21cmSense:
 - All SKA station and sub-station layouts available
 - For CD/EoR, AA* is just as constraining as AA4*
 - Percent-level EoR
 - τ 1σ 20x tighter than Planck18
 - AA* is very constraining even after ~100 hrs



