In the beginning of the Dark Ages, electrically neutral hydrogen gas filled the universe. As stars formed, they ionized the regions immediately around them, creating bubbles here and there. Eventually these bubbles merged together, and intergalactic gas became entirely ionized.

AST(RON

I he epoch of reionization from 21 cm observations





Existing observations leaves much unanswered:

- I) Lyman-alpha forest: end point z>6.5
- 2) CMB optical depth: mid point $z \sim 9 \pm 1.5$
- 3) kSZ amplitude: duration z<3 ?

HST probes skewer much smaller than scale of ionized regions + only brightest sources

LAE surveys (e.g. HSC) may eventually show enhanced clustering from reionization LBG/LAE surveys with Euclid possible to $z\sim8$ in deep field

Fundamental need for new types of observation to understand details of reionization

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Evolving optical depth



tau has fallen consistently as systematics better accounted for

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 $\left(\right)$

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HST

LBG detected up to $z\sim10$. Small numbers at z>7. JWST launches 2018

Decline in LAE/LBG fraction at z~7 possible signature of reionization



Schenker+ 2014

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HSC/LAE

Hypersuprime Cam (HSC) early data on Subaru ~14 deg² (21.2 deg²) 959 (z=5.7), 873 (z=6.6) LAE

No indication of reionization seen in clustering of LAEs

Evolution of number counts also consistent with evolution for 10¹¹ MSol halos



x_{HI}<0.3 at z=6.6

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Ultimately need to combine all reionization constraints e.g. Robertson+ (2015), Greig & Mesinger (2016), ... many more

Reionization histories typically very model dependent



Ota+ 2017

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21 cm & EoR





21 cm basics

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Hyperfine transition of neutral hydrogen



Spin temperature describes relative occupation of levels

$$n_1/n_0 = 3 \exp(-h\nu_{21cm}/kT_s)$$

Useful numbers:

 $\begin{array}{l} 200 \, \mathrm{MHz} \rightarrow z = 6 \\ 100 \, \mathrm{MHz} \rightarrow z = 13 \\ 70 \, \mathrm{MHz} \rightarrow z \approx 20 \\ \textbf{50 MHz} => \textbf{z} \sim \textbf{27} \end{array}$

 $t_{Age}(z = 6) \approx 1 \,\mathrm{Gyr}$ $t_{Age}(z = 10) \approx 500 \,\mathrm{Myr}$ $t_{Age}(z = 20) \approx 150 \,\mathrm{Myr}$ $t_{Age}(z=27) \sim 100 \,\mathrm{Myr}$

 $t_{\rm Gal}(z=8) \approx 100 \,\mathrm{Myr}$



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21 cm astrophysics

Pritchard & Loeb 2010



Systematic path to probing different epochs

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LOFAR







(completed)







Calibration - sky vs redundant (vs hybrid) calibration

Foregrounds ~ 10³-10⁵ signal. Diffuse & pt sources



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Upper limits beginning to make contact with possible models - rule out unheated IGM

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(I) Imaging

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3D maps of topology of reionization - ~ 10 arcmin resolution

Directly image HII region around AGN/bright sources

Environmental information for other probes of reionization

HII bubble catalogs and statistics

 δT (mK) at z=7.02 (117 MHz) with [5',0.8 MHz]



 δT (mK) at z=7.02 (117 MHz) with [5',0.8 MHz]





Granulometry

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Size distribution of ionised regions is an interesting probe of reionization





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21 cm cold spots arise two ways- ionised regions or underdense voids

SKA likely needed to study size distributions

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(2) Evolution of the power spectrum



Mesinger+ 2010

Rich science contained in spatial and redshift evolution of 21 cm power spectrum - observe z = 6 - 27

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Emulation

Reionization simulations often computationally expensive, but vital for 21 cm analysis - approximate schemes needed

Given a computationally expensive model y(x), for parameters x, build an "emulator" $\overline{y}(x)$ that is fast to evaluate

- 1) Form training set of well chosen samples $\{x_{train}, y_{train}=y(x_{train})\}$
- 2) Establish basis for approximation
- 3) Establish means of interpolating from training set to arbitrary points

Used extensively in climate change and also in cosmology

e.g. Coyote universe - emulation of non-linear matter power spectrum P(k | parameters) Heitmann+ 2006, Habib+ 2007

- Latin hyper-cube used to build training set
- PCA basis
- Gaussian processes for interpolation

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Curse of dimensionality: number of points required to sample in a regular grid scales as $(L/\Delta)^N$ - prohibitive for even modest N

Latin hypercube sampling - distribute M points such that each parameter is divided into M bins with each bin occupied only once.

More efficiently uses samples, especially if tweaked to reduce clustering of points e.g. minimax nearest neighbour distances



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Schmit+ 2017

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Different interpolation schemes - Gaussian Processes Kern+ 2017 - Neural networks Schmit+ 2017

Sampling density in MCMC proportional to target density => implies many samples are closely spaced in parameter space

Significant increase in speed without loss of accuracy e.g. training set ~ 10^3 samples c.f. 10^5 - 10^6 in MCMC



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21 cm bispectrum

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Need fast estimator for bispectrum, since large range of possible configurations

Expanding Dirac delta function in terms of exponentials allows factorisation

$$\delta^{\mathrm{K}}(\boldsymbol{m}_{1} + \boldsymbol{m}_{2} \dots + \boldsymbol{m}_{p}) = \frac{1}{N_{\mathrm{pix}}} \sum_{\boldsymbol{n}}^{N_{\mathrm{pix}}} \mathrm{e}^{i2\pi \boldsymbol{n} \cdot (\boldsymbol{m}_{1} + \boldsymbol{m}_{2} \dots + \boldsymbol{m}_{p})/N_{\mathrm{side}}}$$
$$= \frac{1}{N_{\mathrm{pix}}} \sum_{\boldsymbol{n}}^{N_{\mathrm{pix}}} \prod_{i}^{p} \mathrm{e}^{i2\pi \boldsymbol{n} \cdot \boldsymbol{m}_{i}/N_{\mathrm{side}}}.$$

Can then express poly-spectra in terms of sampled FT of 21 cm field

$$\mathscr{P}(\boldsymbol{k}_{1},\boldsymbol{k}_{2},...\boldsymbol{k}_{p}) pprox H^{p} rac{1}{V} rac{\sum\limits_{\boldsymbol{n}}^{N_{\mathrm{pix}}} \prod\limits_{i=1}^{p} \delta(\boldsymbol{n},\,\boldsymbol{k}_{i})}{\sum\limits_{\boldsymbol{n}}^{N_{\mathrm{pix}}} \prod\limits_{i=1}^{p} I(\boldsymbol{n},\,\boldsymbol{k}_{i})}, \qquad \delta(\boldsymbol{n},\,\boldsymbol{k}_{i}) = \sum\limits_{\boldsymbol{k}_{i}/k_{\mathrm{F}}\simeq\boldsymbol{m}_{i}} \Delta_{\mathrm{FFT}}(\boldsymbol{m}_{i}) \mathrm{e}^{i2\pi\boldsymbol{n}\cdot\boldsymbol{m}_{i}/N_{\mathrm{side}}}, \qquad I(\boldsymbol{n},\,\boldsymbol{k}_{i}) = \sum\limits_{\boldsymbol{k}_{i}/k_{\mathrm{F}}\simeq\boldsymbol{m}_{i}} \mathrm{e}^{i2\pi\boldsymbol{n}\cdot\boldsymbol{m}_{i}/N_{\mathrm{side}}},$$

For example, estimator of the bispectrum

$$B(k_{\rm F} m_1, k_{\rm F} m_2, k_{\rm F} m_3) pprox rac{V^2}{N_{
m pix}^3} rac{\sum\limits_{m{n}}^{N_{
m pix}} \delta(m{n}, m{k}_1) \delta(m{n}, m{k}_2) \delta(m{n}, m{k}_3)}{\sum\limits_{m{n}}^{N_{
m pix}} I(m{n}, m{k}_1) I(m{n}, m{k}_2) I(m{n}, m{k}_3)} ,$$

Watkinson+ 2017



Can also use ML for inversion (rather than Bayesian inference)

e.g. Neural Networks to identify astrophysics parameters from shape of power spectrum Shimabukuro & Semelin 2017





Beyond LOFAR... SKA

• MID - dishes (SA) - 250MHz - 20 GHz; LOW - antennae (AU) - 50-350MHz



- International headquarters at Jodrell Bank
- SKA will be the premier radio telescope facility with science case ranging from cradle of life, through cosmology, to epoch of reionization
- Cost control exercise currently underway
- Form IGO 2019 Commissioning/Early science 2021-2024 Key science projects 2024-29



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Conclusions

- Reionization is still interesting! CMB, HST, and LAE surveys seem to indicate interesting behaviour at z~7
- LOFAR, MWA, PAPER are all confronting data with increasingly mature analysis pipelines. Still much to be done.
- 21 cm upper limits are beginning to make contact with plausible (if extreme) models
- Imaging will become interesting with SKA e.g. granulometry an for measuring bubble size distributions
- Power spectrum easiest statistic to work with currently
- Application of emulation may prove a useful conduit between full numerical simulations and Bayesian analysis
- Other non-Gaussian statistics also useful e.g. bispectrum. Need to be fast to evaluate to include in analysis pipelines
- Fruitful to apply machine learning techniques
- SKA finalising design with construction expected to begin ~2019

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Analysis more complicated than for, say, CMB because

I) No complete and agreed upon model of reionization

- each simulation uses different parametrization & methods
- different prescriptions for sources and sinks of photons
- no guarantee that any are "correct"
- unclear any one simulation covers all possibilities

2) Not clear simple statistics e.g. power spectrum capture all information

- 3) Evaluation of models is computationally expensive
 - full hydro + RT requires hundreds of hours of super-computer time

Strategies:

- I) Analytic models
- 2) Semi-numerical codes e.g. 21cmFAST, Simfast21, ...
- 3) Emulation

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IF find z>6 radio bright QSOs or other radio bright objects e.g. GRB

10kHz resolved spectra of 21cm forest in bright radio sources at z>6



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Overview



S.G. Djorgovski et al. & Digital Media Center, Caltech

Reionization

• CMB constraints of reionization

- 21 cm observations
- CMB optical depth from 21cm

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SKA-LOW observing strategy



Shallow: LOFAR-like power spectrum sensitivity over 10000 deg². Middle: Shallow imaging + power spectrum over 1000 deg² Deep: Power spectrum to z<27 & deep imaging over 100 deg² IM: OmegaHI & cosmology at 3<z<6 over 1000 deg²

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Intensity mapping with LOW for OmegaHI & cosmology



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Existing observations leaves much unanswered:

- I) Lyman-alpha forest: end point z>6.5
- 2) CMB optical depth: mid point $z \sim II$ (Revised down by Planck to $z \sim 8.8 \pm 1.5$)
- 3) kSZ amplitude: duration z<4.4 ?

HST probes skewer much smaller than scale of ionized regions + only brightest sources

Large galaxy samples with LAE surveys or Euclid possible to z~8

Fundamental need for new types of observation to understand details of reionization

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Euclid synergies?

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- z=8 QSO to find those that are radio loud
- Bright galaxies for stacking
- LAE 21 cm cross-correlation as probe of HII region sizes
- ???



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- SKAI should begin commissioning/early science in 2021 with full operations in ~2024. (c.f. Euclid 2021-27)
- Taking shape now cost control exercise, IGO, ...
- EoR plans HI surveys at z = 3 27
 - SHALLOW 10,000 deg² at LOFAR like sensitivity, z~7-15
 - MEDIUM + IM 5000 deg² z ~ 3-20
 - DEEP 100 deg², z~6-27
- Synergies with Euclid from both MID and LOW
 - MID better matched to low redshifts z<3 (Pourtsidou talk)
 - LOW for 3<z<27
- For reionisation:
 - LAE-21 cm cross correlations; Stack on LBGs?
 - High-z quasars for radio follow up
 - OmegaHI
 - ???
- SKA-EoR Science Team preparing for KSP bid in ~2019 Synergy focus group leads: Pratika Dayal, Erik Zackrisson

SKA EoR Science Team

Leon Koopmans (NL, Chair ST), Gianni Bernardi (SA), Garrelt Mellema (SE), Andrei Mesinger (IT), Jonathan Pritchard (UK, Chair SWG), Cath Trott (AU) Abhi Datta (IN)

Various focus groups studying aspects of doing 21cm with SKA

https://sites.google.com/site/skacdeorscienceteam/home



Figure 1: Flow-diagram for End-to-End data simulations.



Cosmology (MID)

- galaxy surveys SKAI: Continuum, HI IM; SKA2: HI
- weak lensing with SKA2 different systematics (think 5000 deg² survey)

Reionization (LOW)

- 21 cm LAE cross correlation
- stack on bright galaxies at high-z
- Radio loud quasars at z>6 (especially z~8)
- HI intensity mapping at 3<z<6

•General

LOW (50-350 MHz),

MID - Band I (580-1015MHz), Band 2 (0.9 - 1.67 GHz), Band 5 (8-13.8GHz) **Imperial College**





21 cm Forest requires radio bright sources at z>7

- QSO
- GRB

Euclid will be key for providing z~8 QSOs (hopefully some radio loud)



Manti+ 2017

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