

Lifting weak lensing degeneracies with a field-based likelihood

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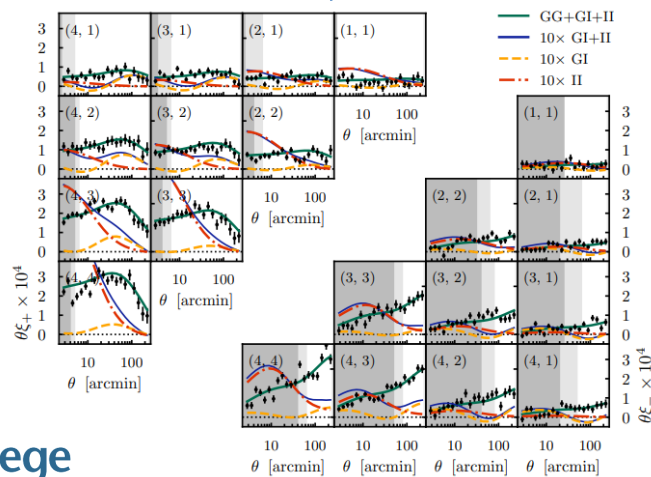
with Alan Heavens, Daniel Mortlock & Guilhem Lavaux

The standard approach to WL analysis



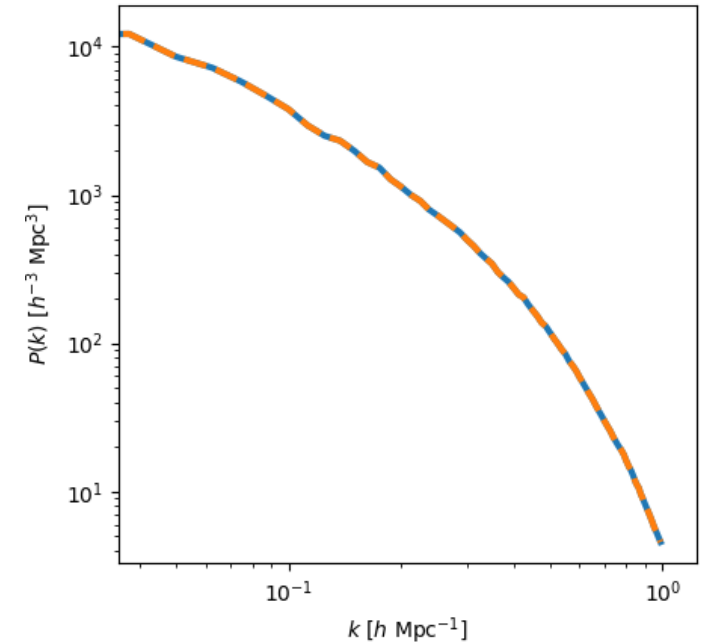
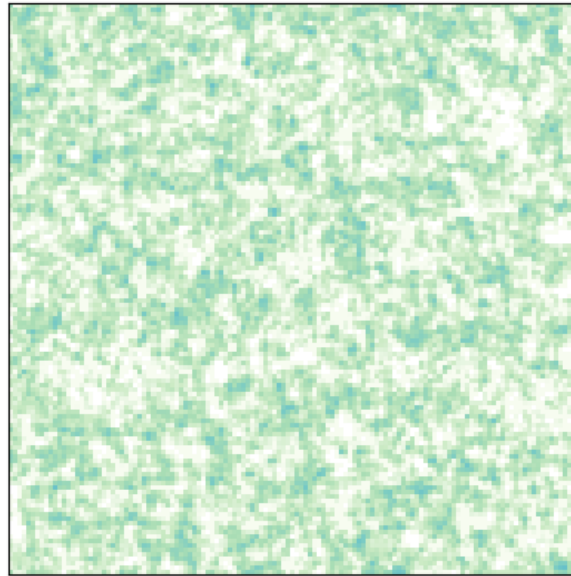
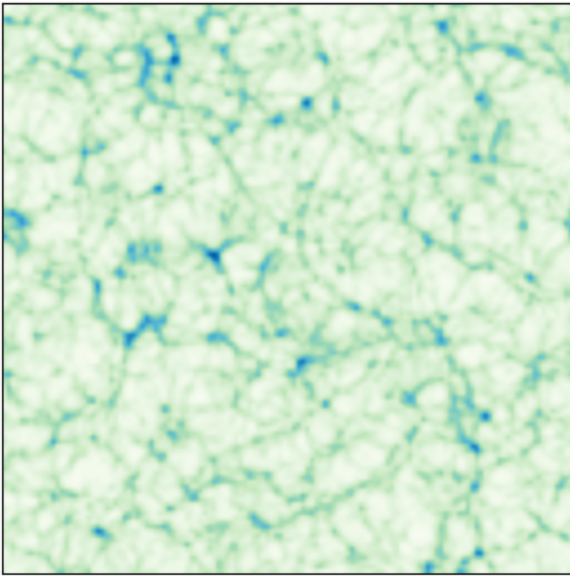
Summarise

Does this capture all the information?



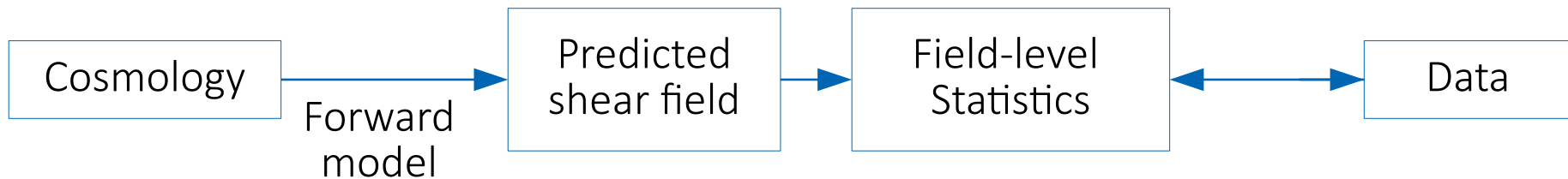
Why a field-based approach?

The standard approach misses information!



- Higher-order statistics: what are the sampling distributions? And covariance matrix?
- Field-based approach without data compression of the pixelised shear.

Field-level approach

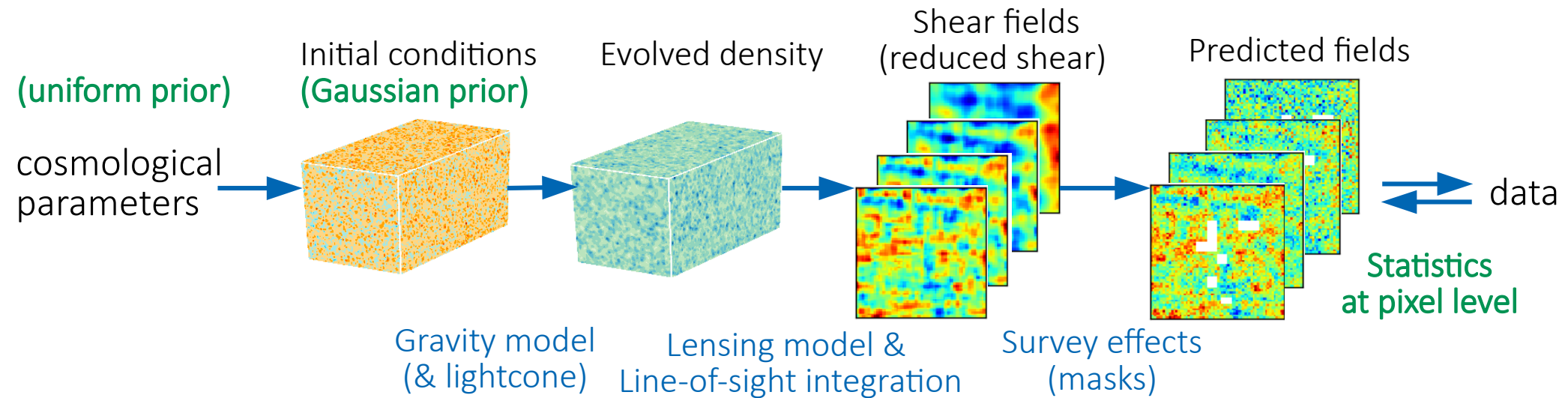


Statistics pixel by pixel can access all the information

Probabilistic forward model → Simulations are constrained by the data

BORG framework: gravity model

Forward model of BORG-WL



Sensitivity to cosmology:

- Initial matter power spectrum
- Growth of structures
- Geometry

Likelihood:

$$-\log P(\hat{\gamma}_1, \hat{\gamma}_2 | \gamma_1, \gamma_2) = \sum_b \sum_{mn} \frac{(\gamma_{1,mn}^b - \hat{\gamma}_{1,mn}^b)^2 + (\gamma_{2,mn}^b - \hat{\gamma}_{2,mn}^b)^2}{2\sigma_{mn}^2}$$

Assumes Gaussian noise in pixelised shear field

Doesn't need covariance matrix! $\sigma_{mn}^2 = \sigma_\epsilon^2 / N_{\text{sources}}$

Getting the full posterior distribution

$$P(\theta, \delta^{\text{ic}} | \hat{\gamma}) \propto P(\hat{\gamma} | \gamma) P(\gamma | \delta^{\text{ic}}) P(\delta^{\text{ic}} | \theta) P(\theta)$$

posterior

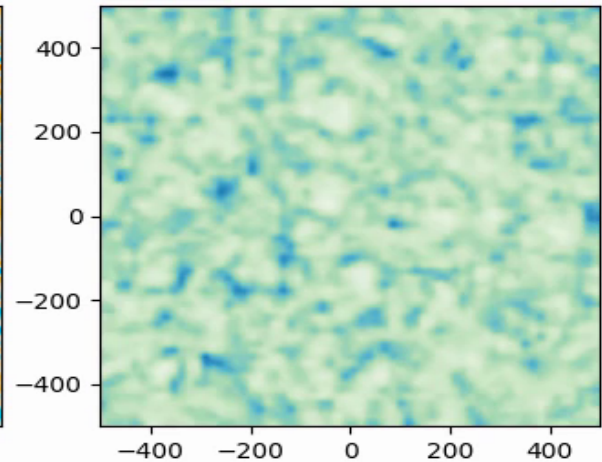
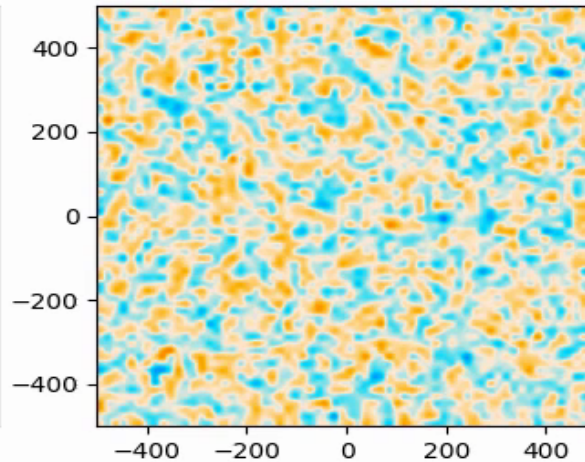
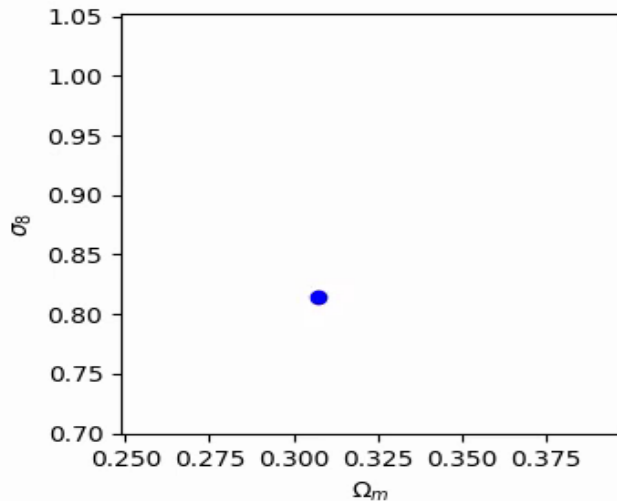
likelihood

prior

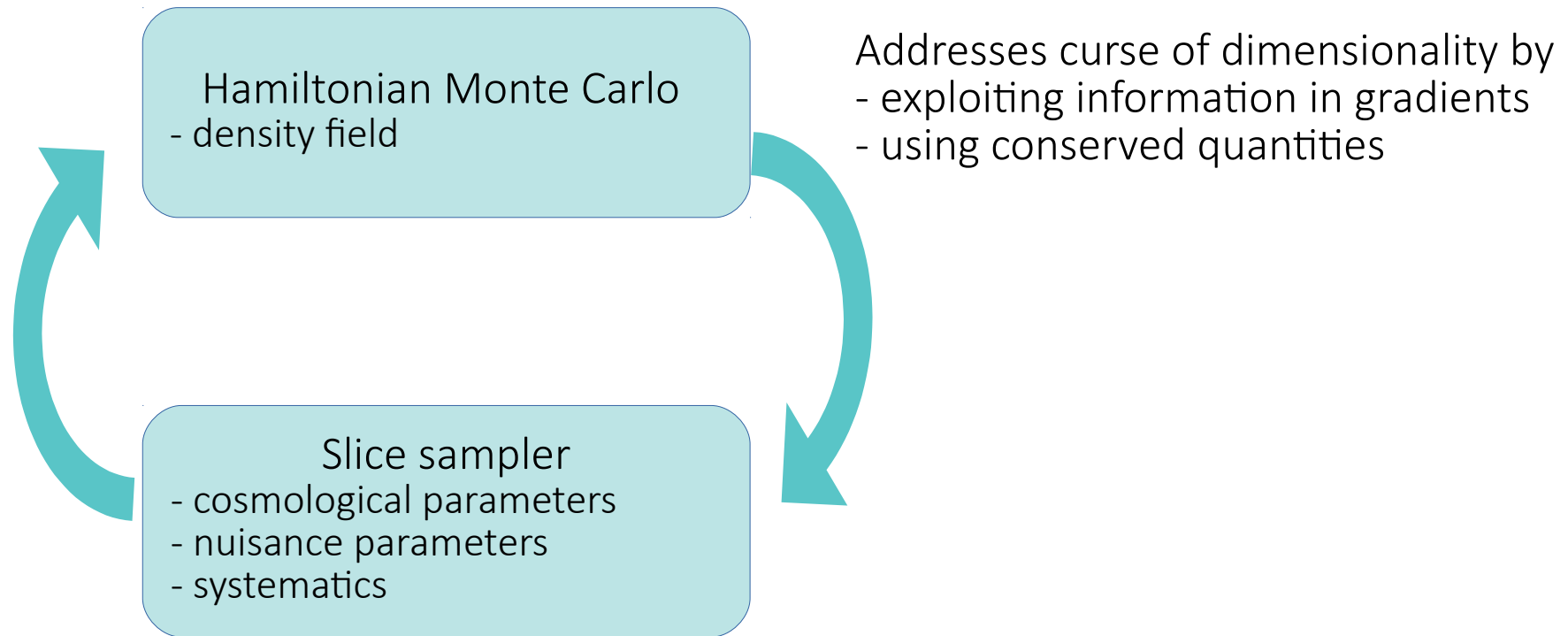
10⁷ parameters!

$$\delta^D(\gamma - \mathcal{F}(\delta^{\text{ic}})) \quad G(0, S(\theta))$$

forward model prior

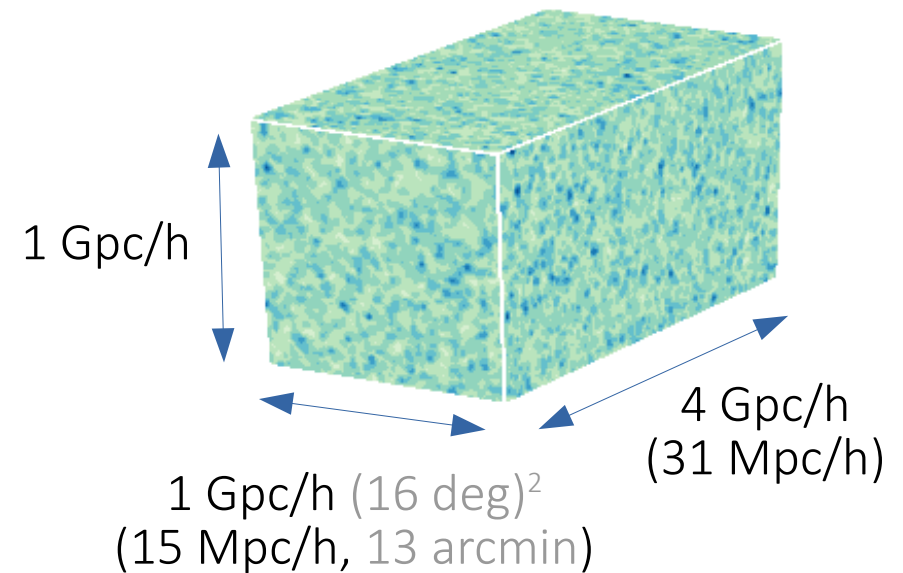
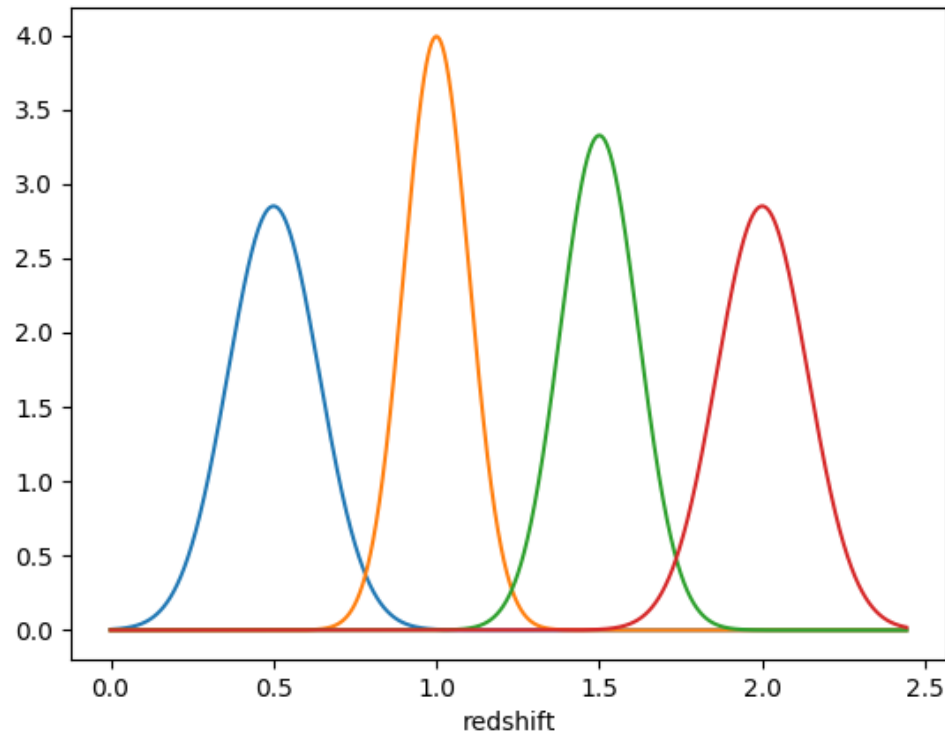


Sampling efficiently in a high-dimensional space



Simulated data

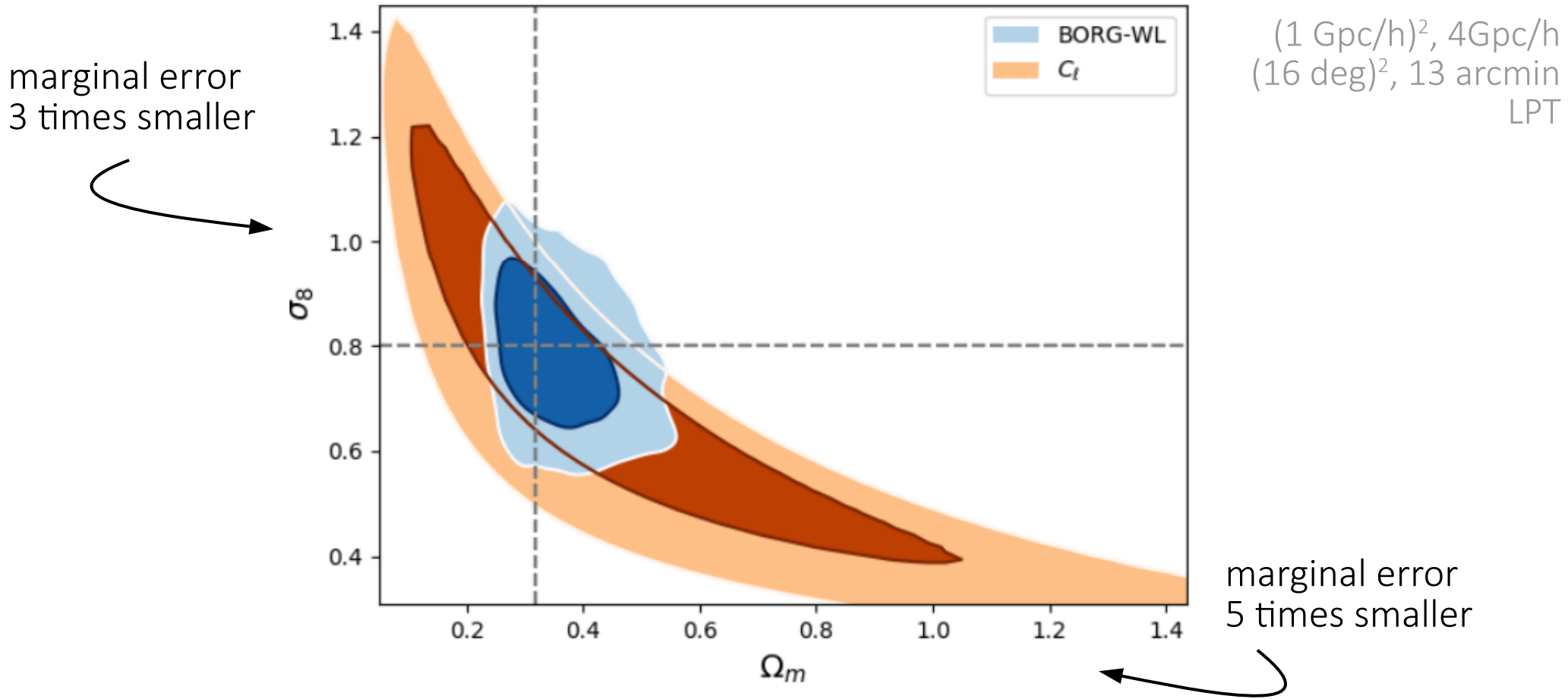
Tomographic bins



Gaussian pixel-noise corresponding to 30 sources per square arcmin

Gravity model for this experiment LPT (but PM is also available in BORG)

Comparison of cosmology constraints

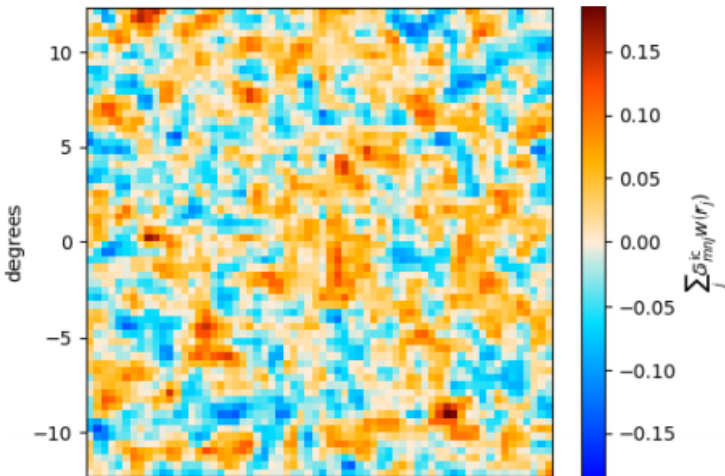


Field-level approach **lifts degeneracy** by extracting more information from the data

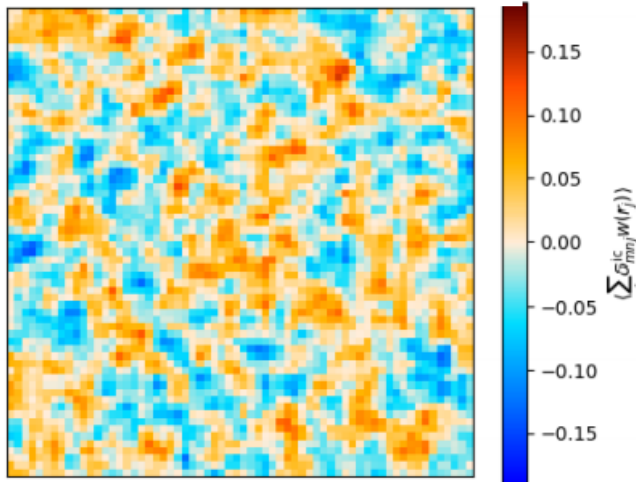
NP, Heavens+ 2022

Projection of inferred density fields

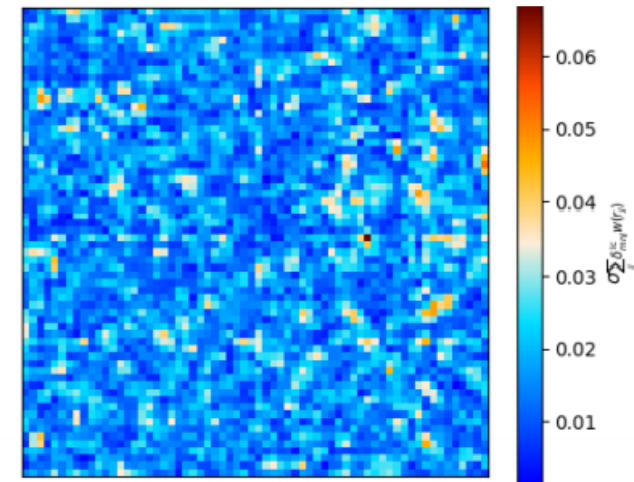
True initial conditions



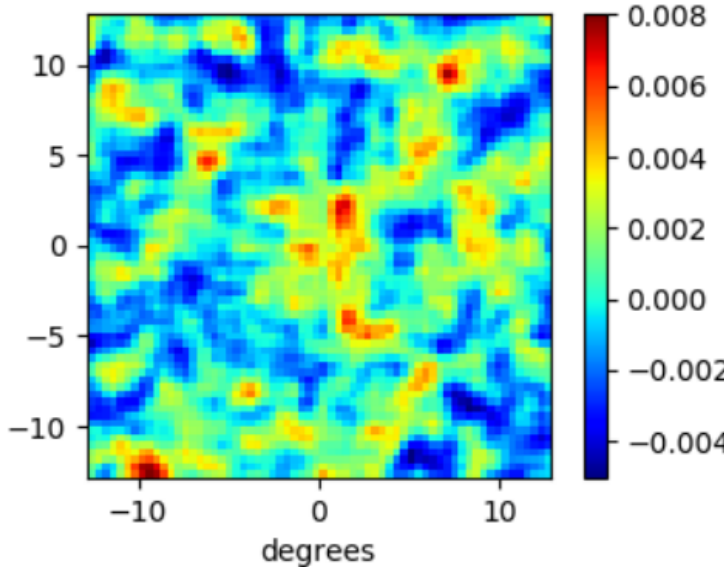
Mean initial conditions



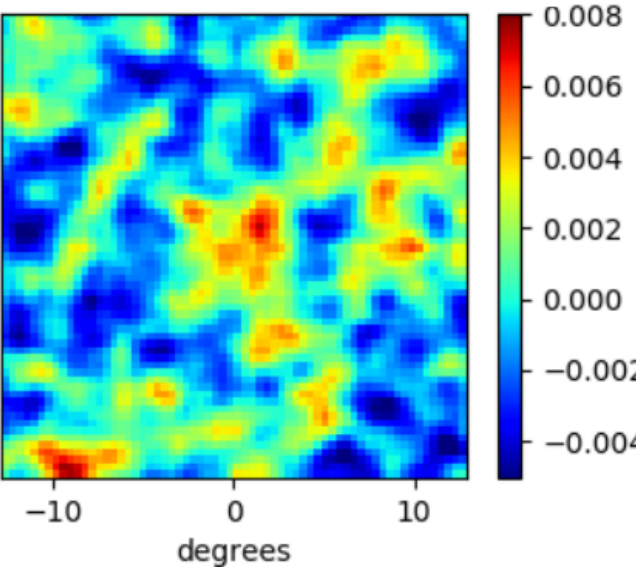
Standard deviation



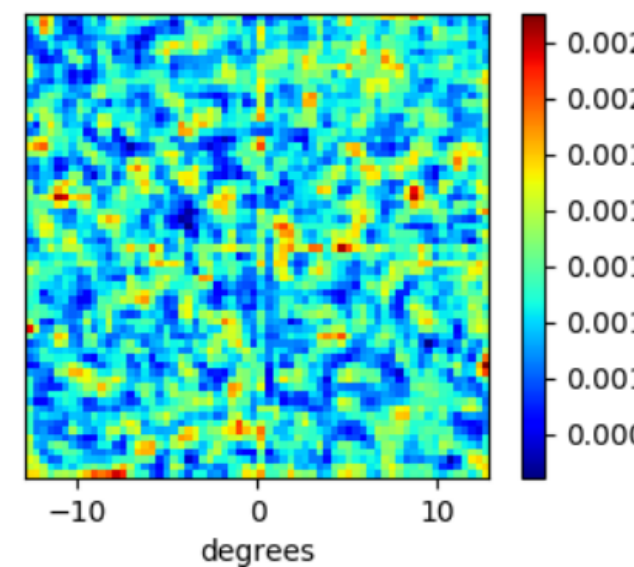
True convergence



Mean convergence

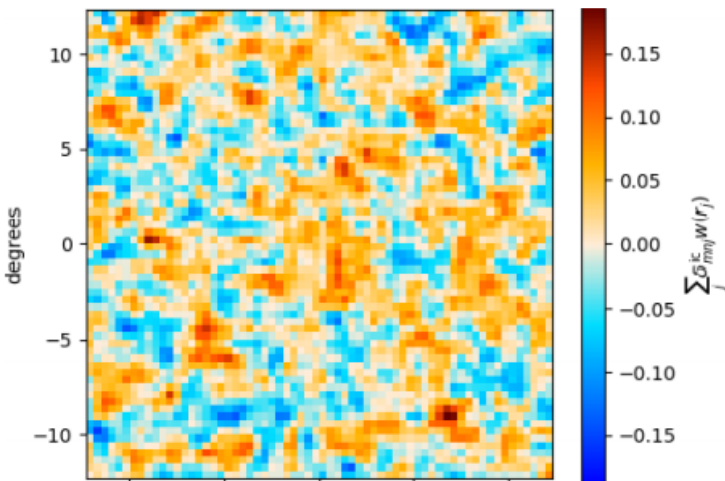


Standard deviation

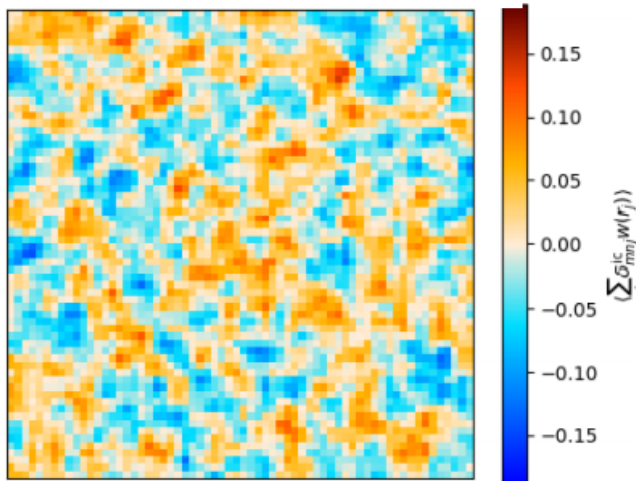


Projection of inferred density fields

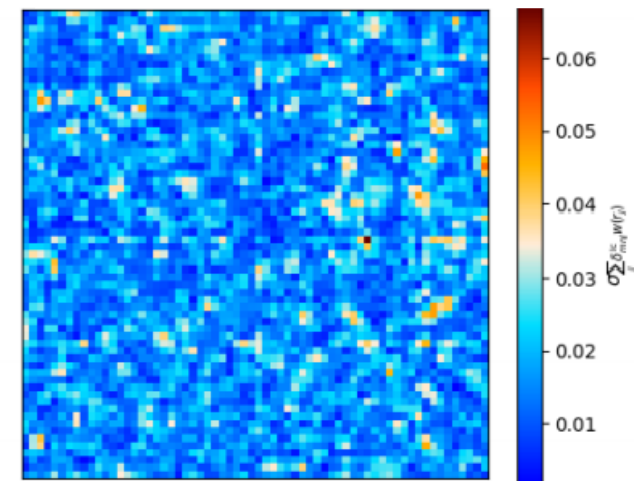
True initial conditions



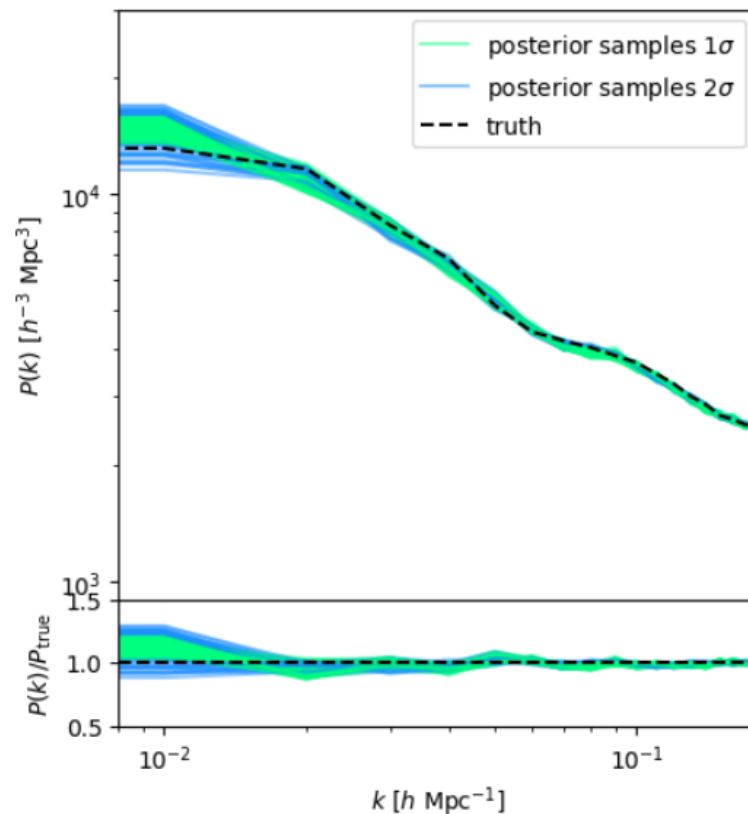
Mean initial conditions



Standard deviation



Initial matter power spectrum (3D)



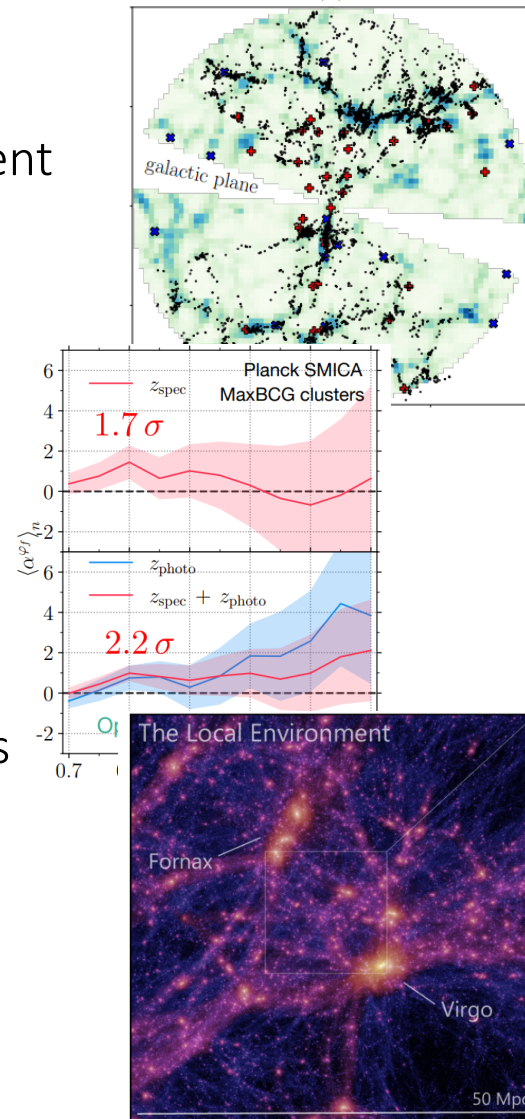
Why should we care about the density fields?

We can do physics

- Galaxy or SN environment
(Porqueres et al. 2017)
(Tsaprazi et al. 2021)

- kSZ effect
(Nguyen et al 2020)

- Constrained simulations
(McAlpine et al 2022)

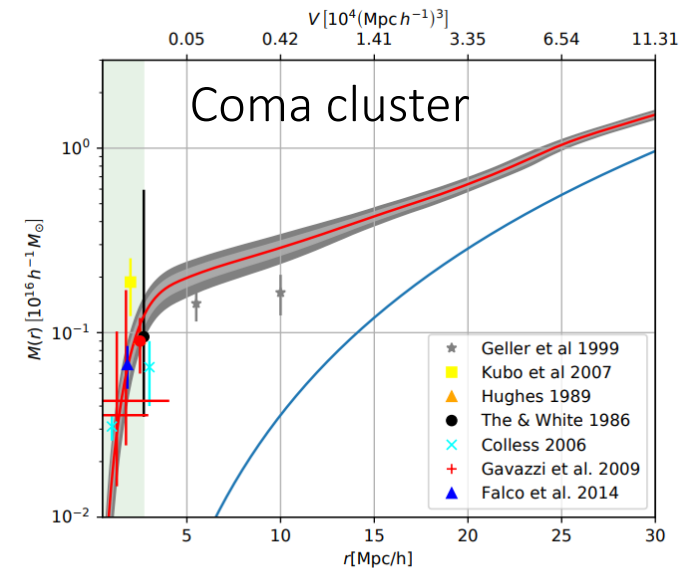


... and also powerful tests

- do we get the clusters we know they are there?

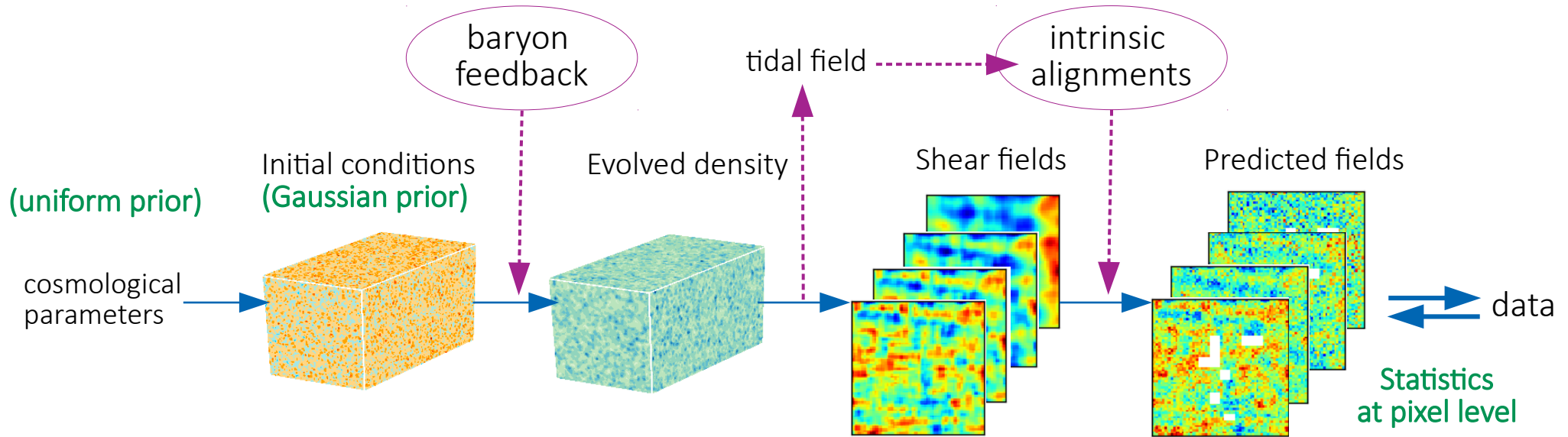
- do we get the expected mass profiles?

- are the inferred IC compatible with CMB observations?



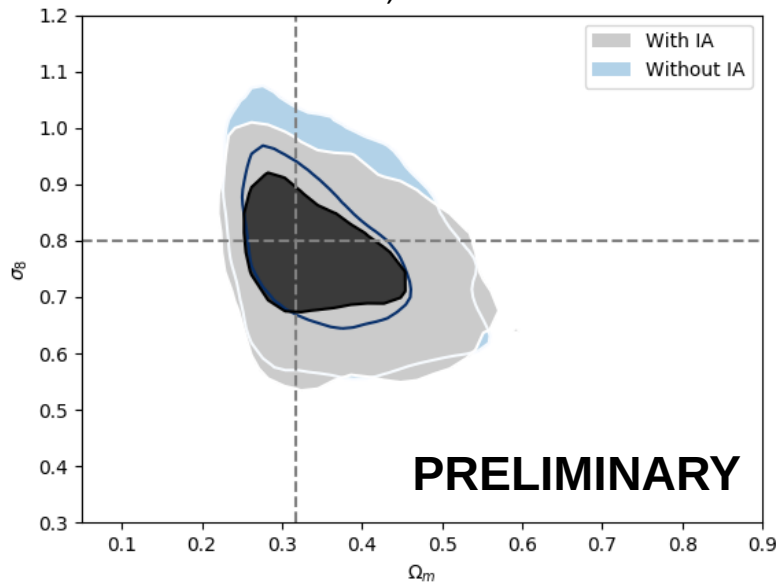
Jasche & Lavaux 2019

Control of systematics (work in progress)

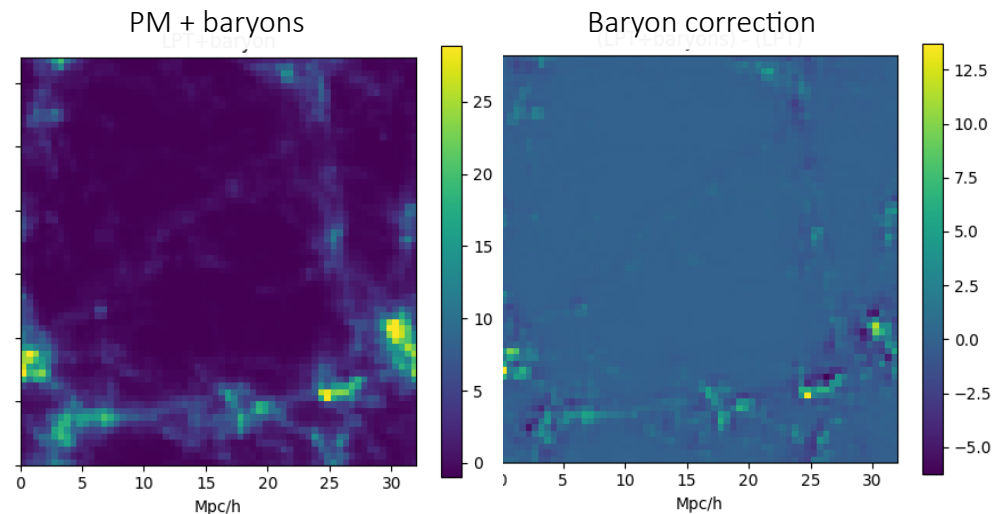


We can use **physical models of systematics** rather than effective models

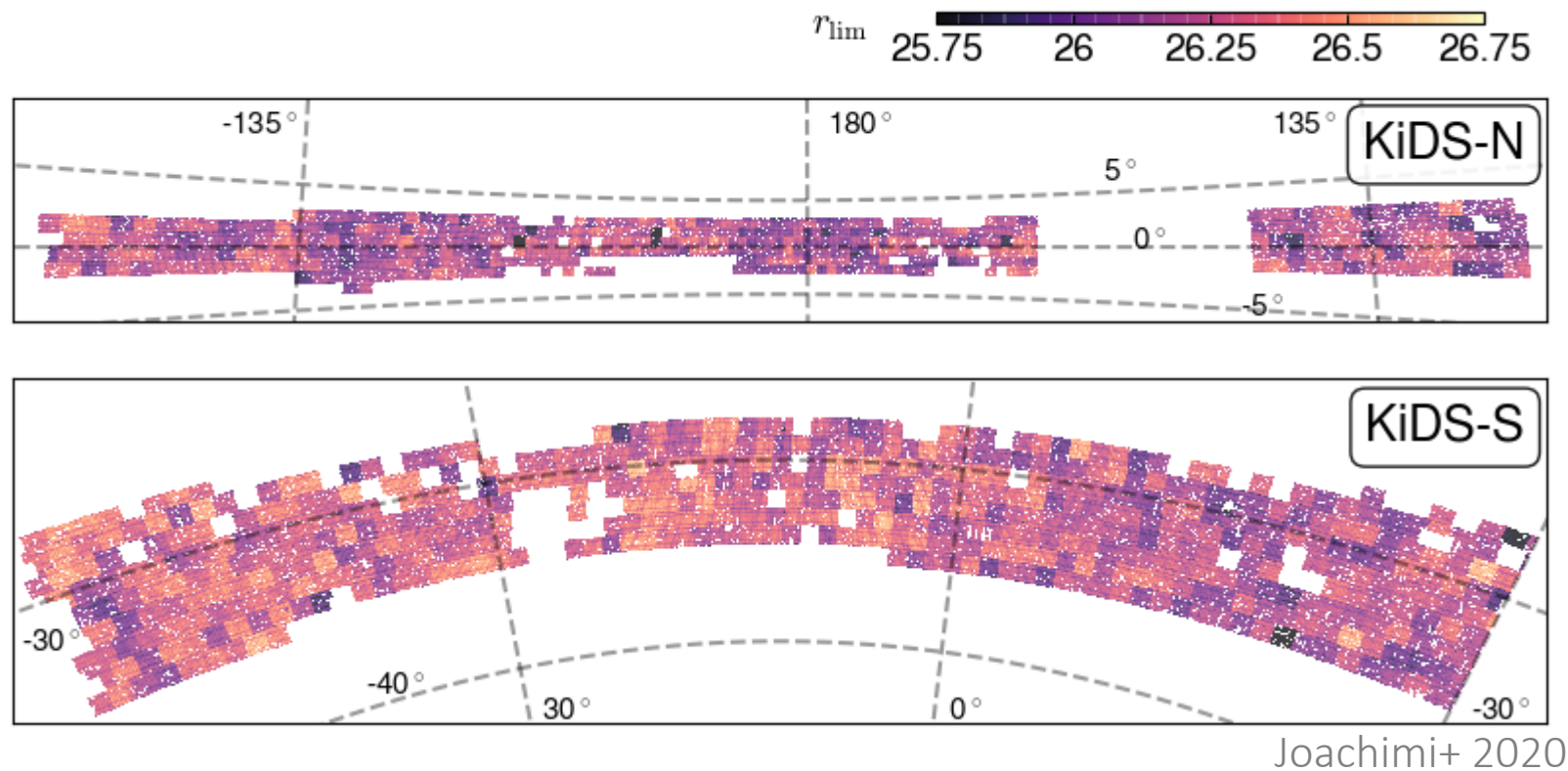
IA: NLA, TATT



Baryons: Dai et al 2018



Survey systematics



- Variable depth
- Variable seeing (PSF)
- Variation in source galaxy and redshift distribution

Summary and outlook

- There is more information in the data that the 2-point summary statistics do not capture.
- Field-based approach lifts degeneracy and reduces marginal error up to a factor of 5.
- What's next?
First real data analysis with BORG-WL

