

Inferring the dynamical growth of structures at high-redshift

Natalia Porqueres

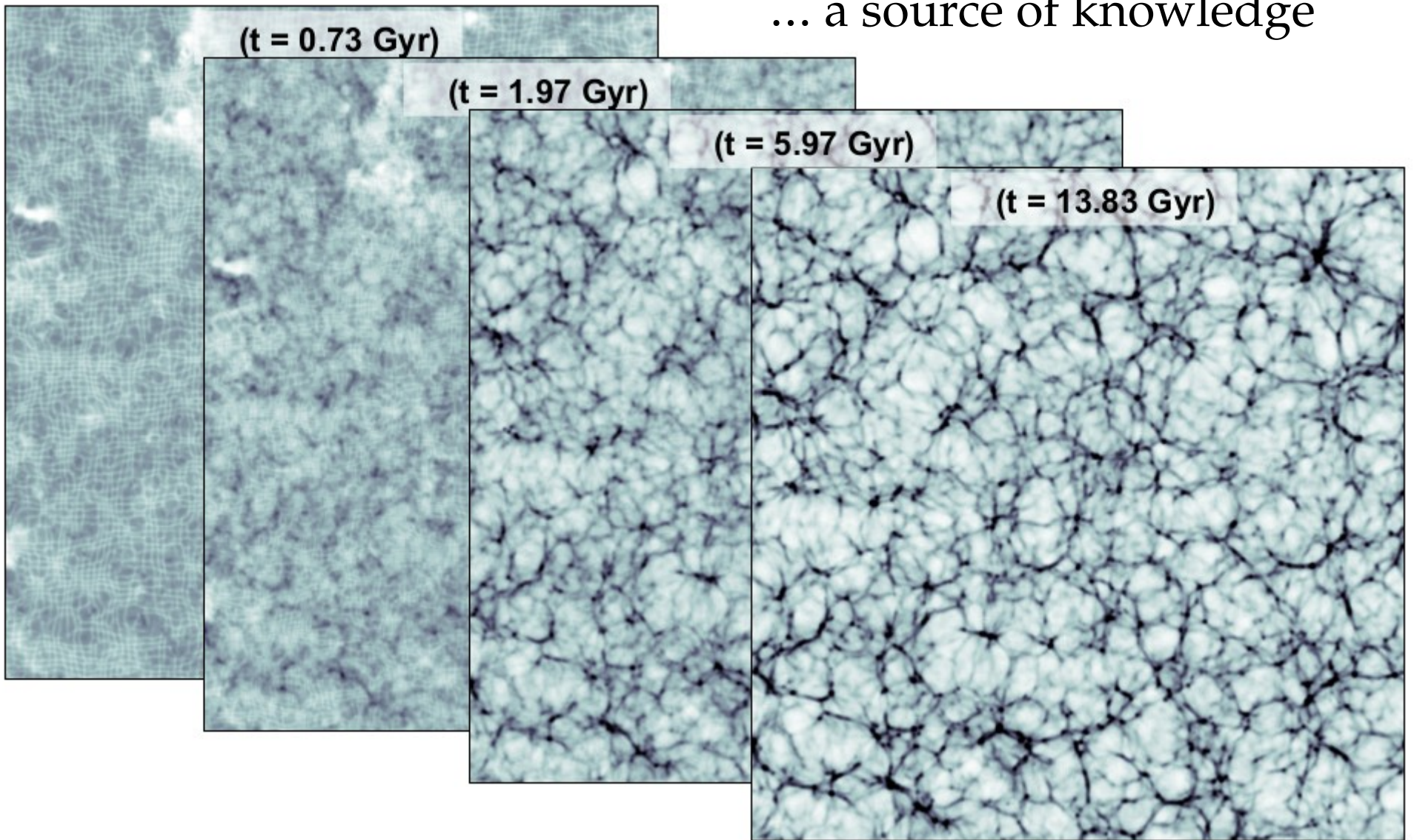
Jens Jasche, Guilhem Lavaux, Torsten Enßlin

Observatoire de la Côte d'Azur

25.02.20

The large-scale structure

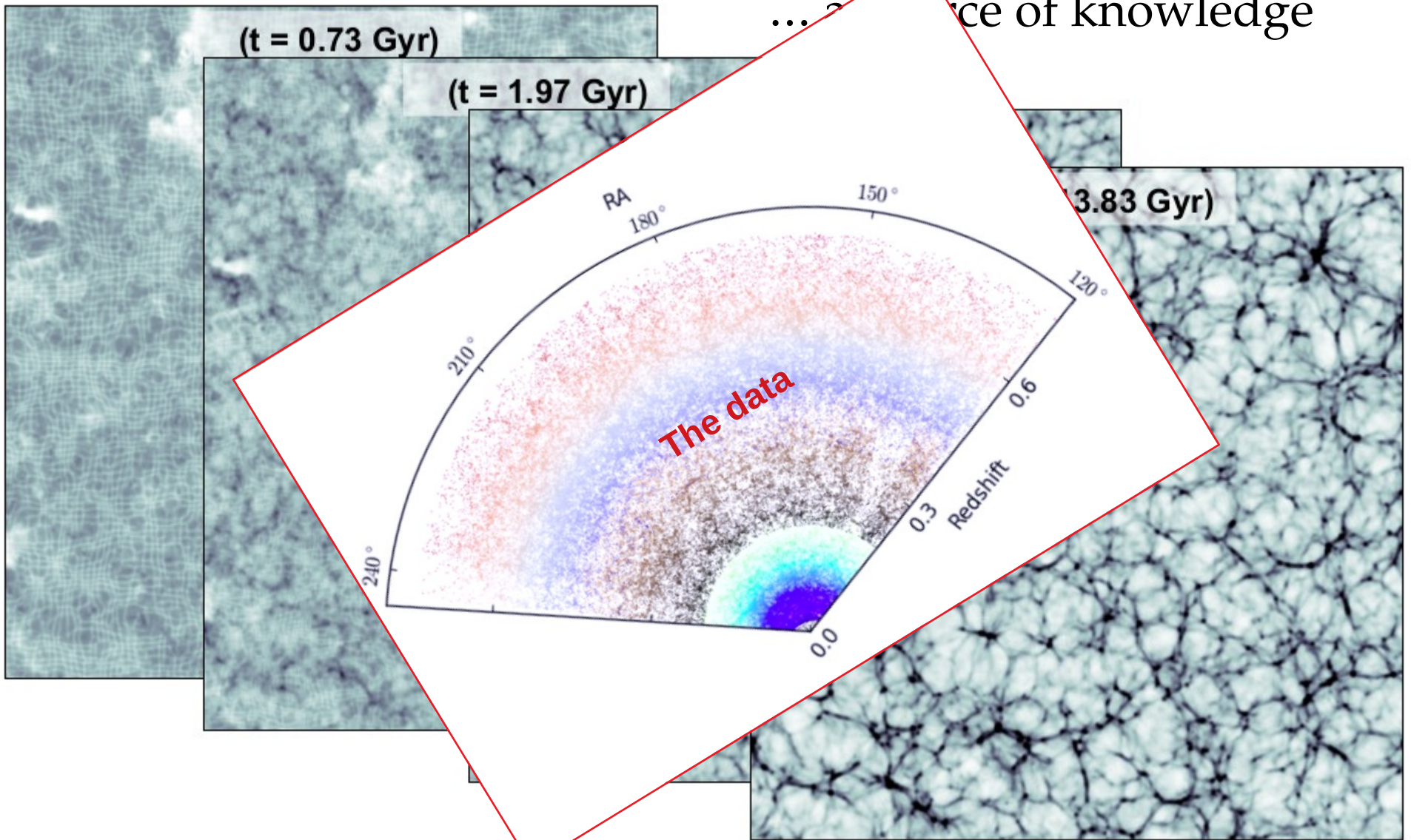
... a source of knowledge



[Image credit: Jens Jasche]

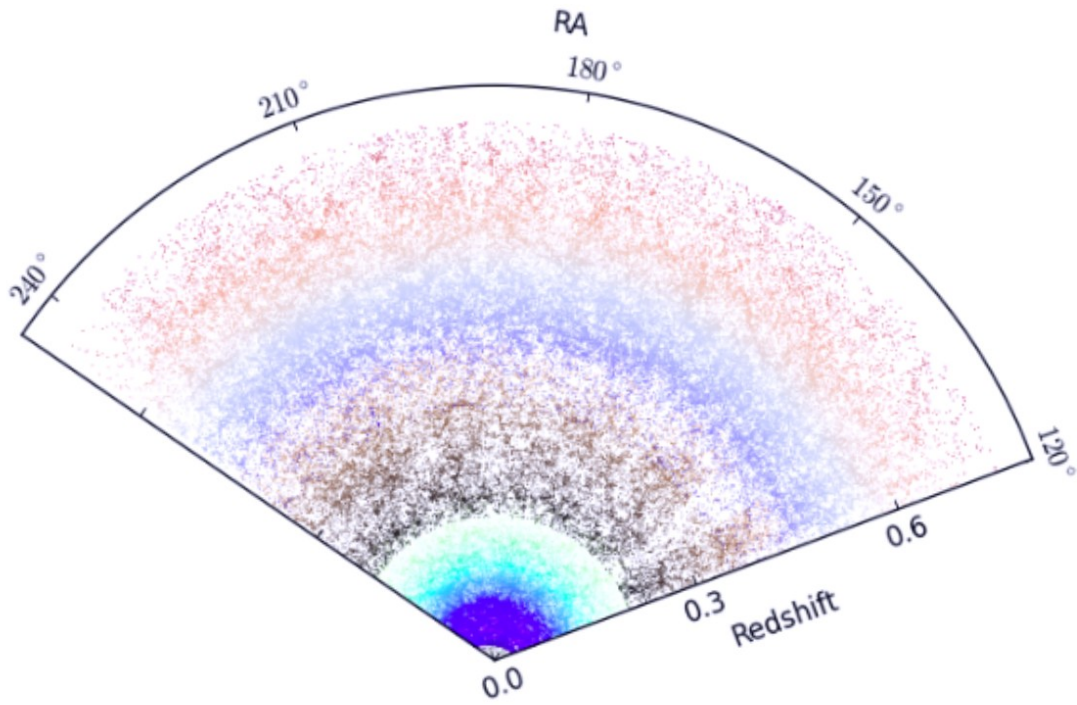
The large-scale structure

... a source of knowledge



[Image credit: Jens Jasche]

From data to cosmology



incomplete

noise

no unique recovery of LSS

Statistics!

$$P(\delta|d) = \frac{P(d|\delta)P(\delta)}{P(d)}$$

Ω
cosmology

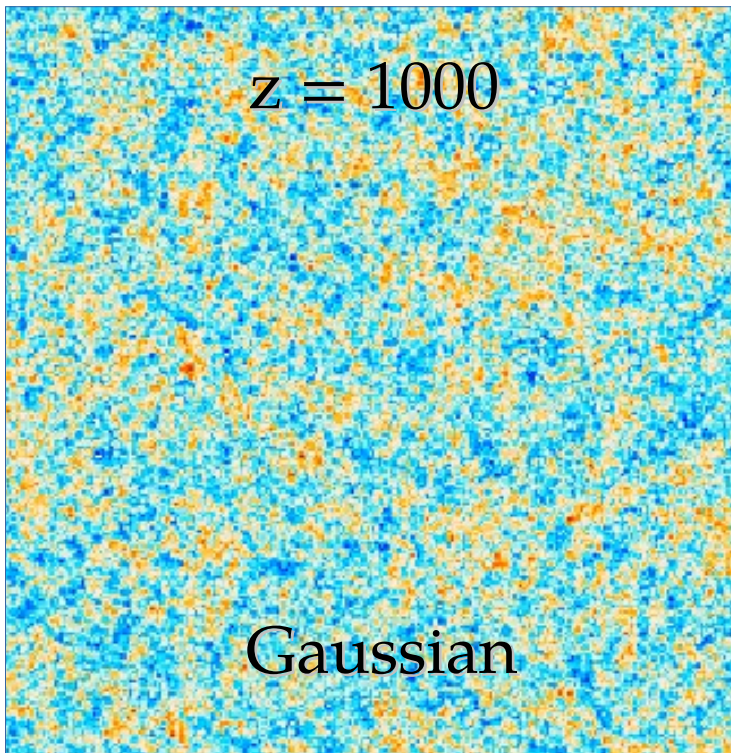
The BORG framework

[Jasche & Wandelt 2012]

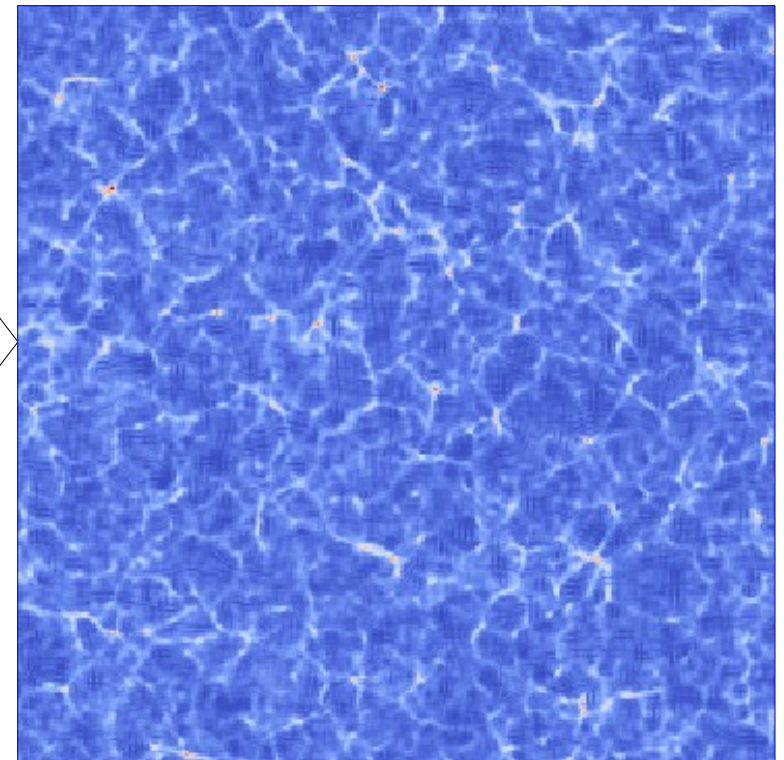
$$P(\delta|d) \propto P(d|\delta)P(\delta)$$

likelihood prior

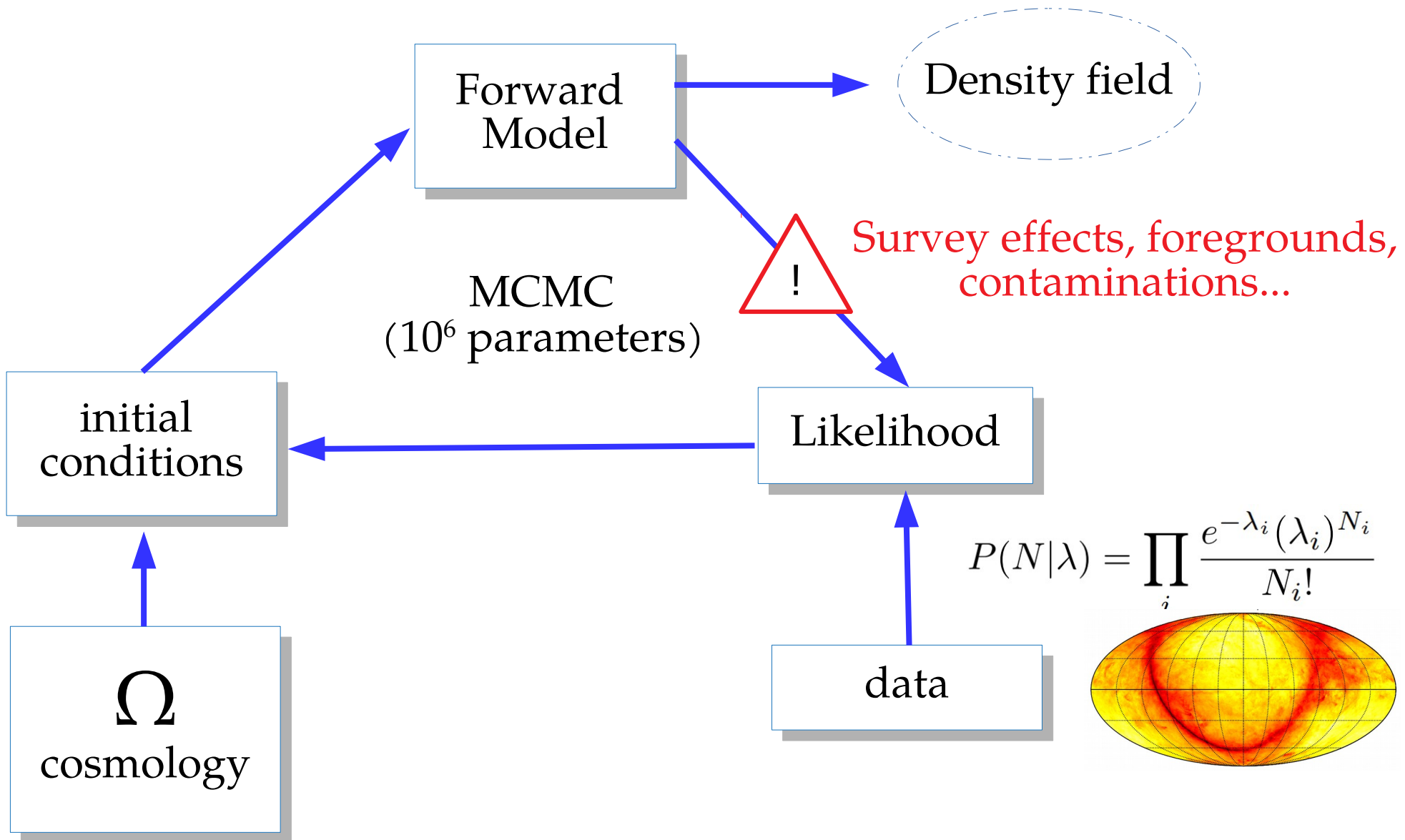
$$\propto P(d|\delta) \underbrace{P(\delta|\delta_0)}_{\delta^D(\delta - \mathcal{F}(\delta_0))} \underbrace{P(\delta_0)}_{G(0,S)}$$



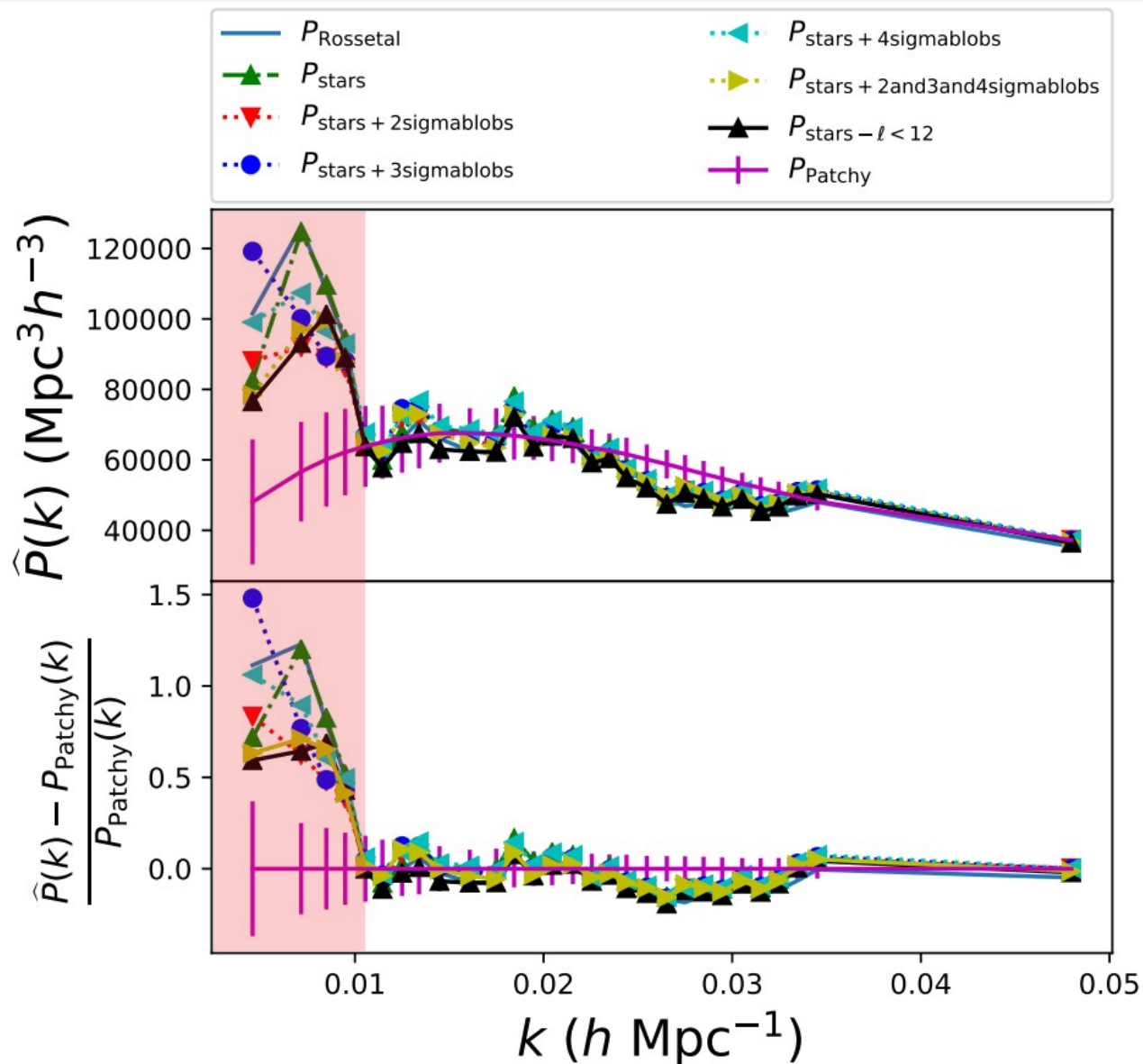
gravity



The inference chain

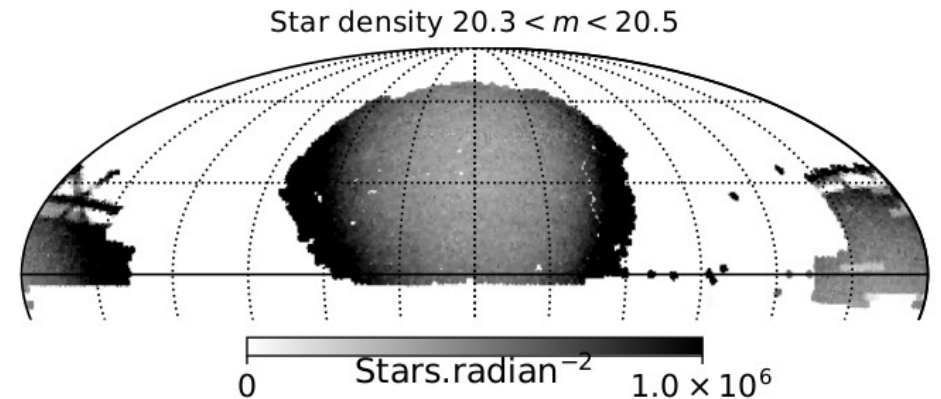
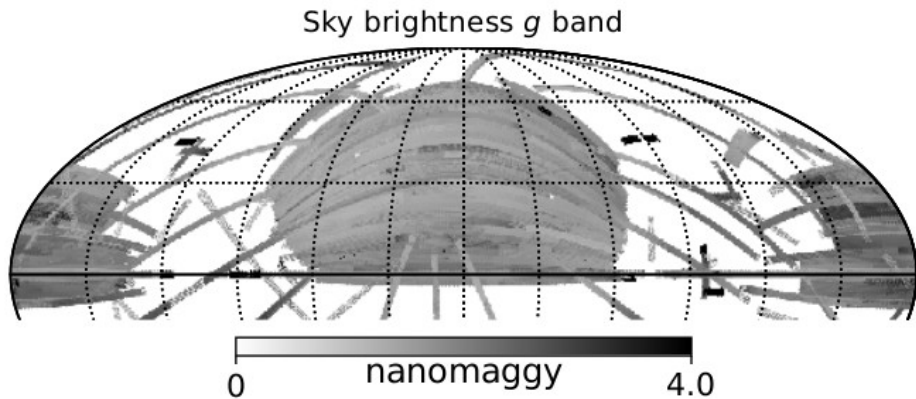
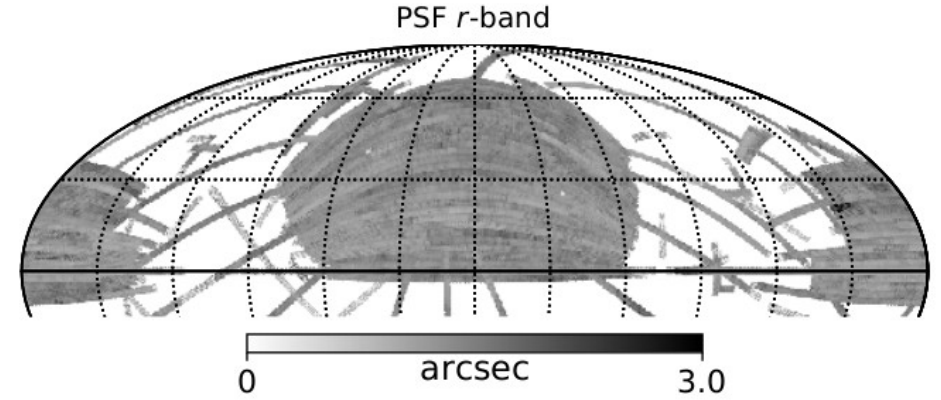
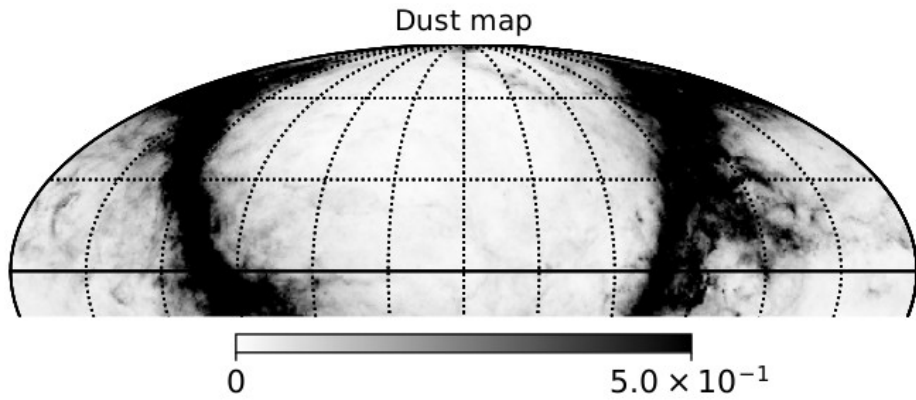


Effect of contaminations



[Kalus, Percival et al 2018]

Foreground templates



[Lavaux, Jasche, Leclercq 2019]

Unknown systematics?

Robust Likelihood

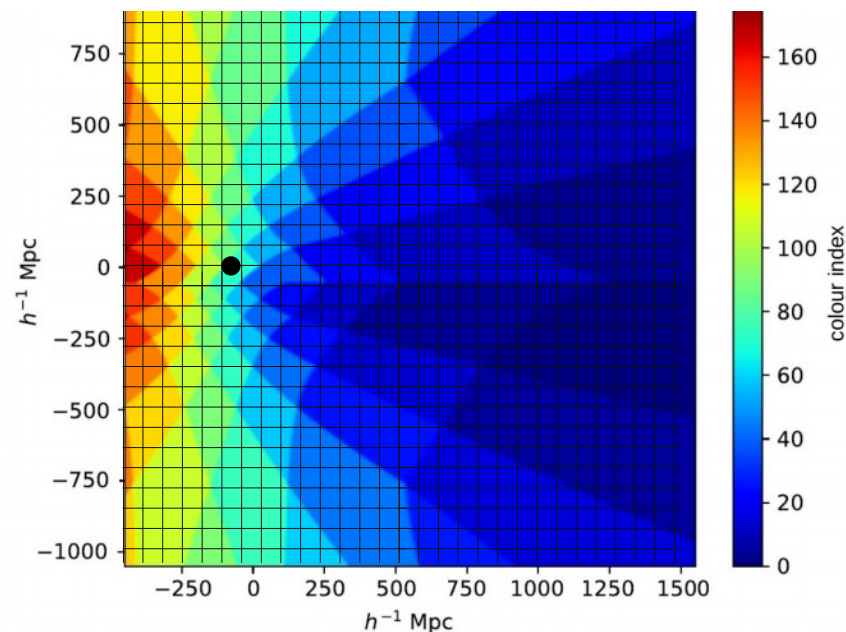
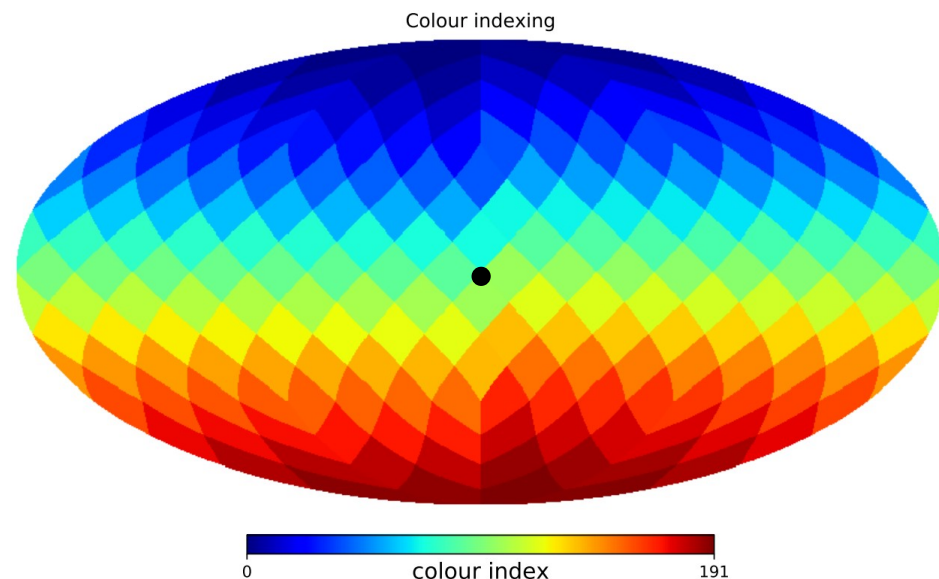
[Porqueres et al. 2019]

$$\lambda_c = A_c \bar{\lambda}_c$$

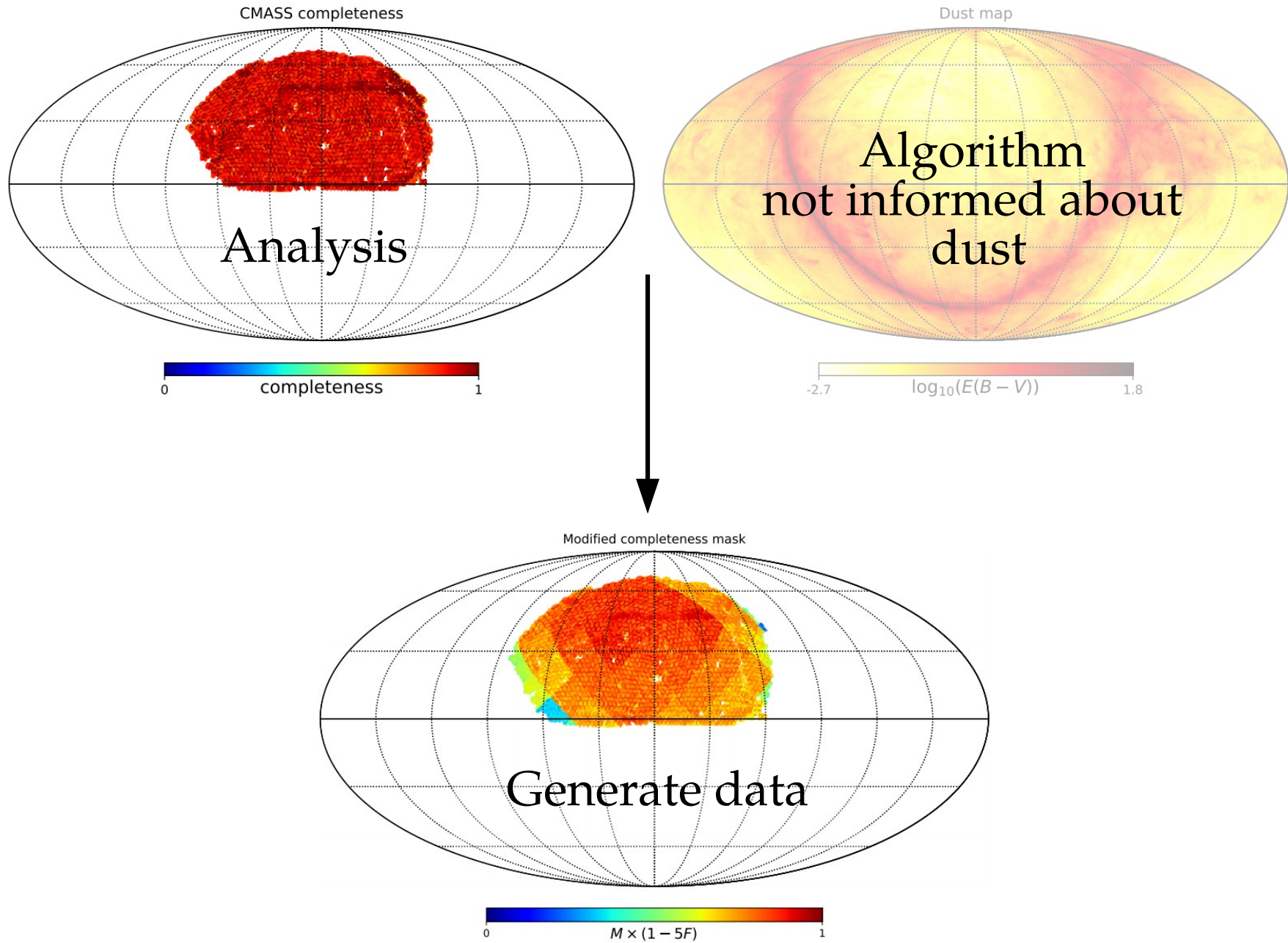
$$P(N|\lambda) = \prod_i \frac{e^{-\lambda_i} (\lambda_i)^{N_i}}{N_i!}$$

Marginalizing over A:

$$P(N|\lambda) \propto \prod_c \prod_{i \in \mathcal{A}_c} \left(\frac{\lambda_i}{\sum_{j \in \mathcal{A}_c} \lambda_j} \right)^{N_i}$$

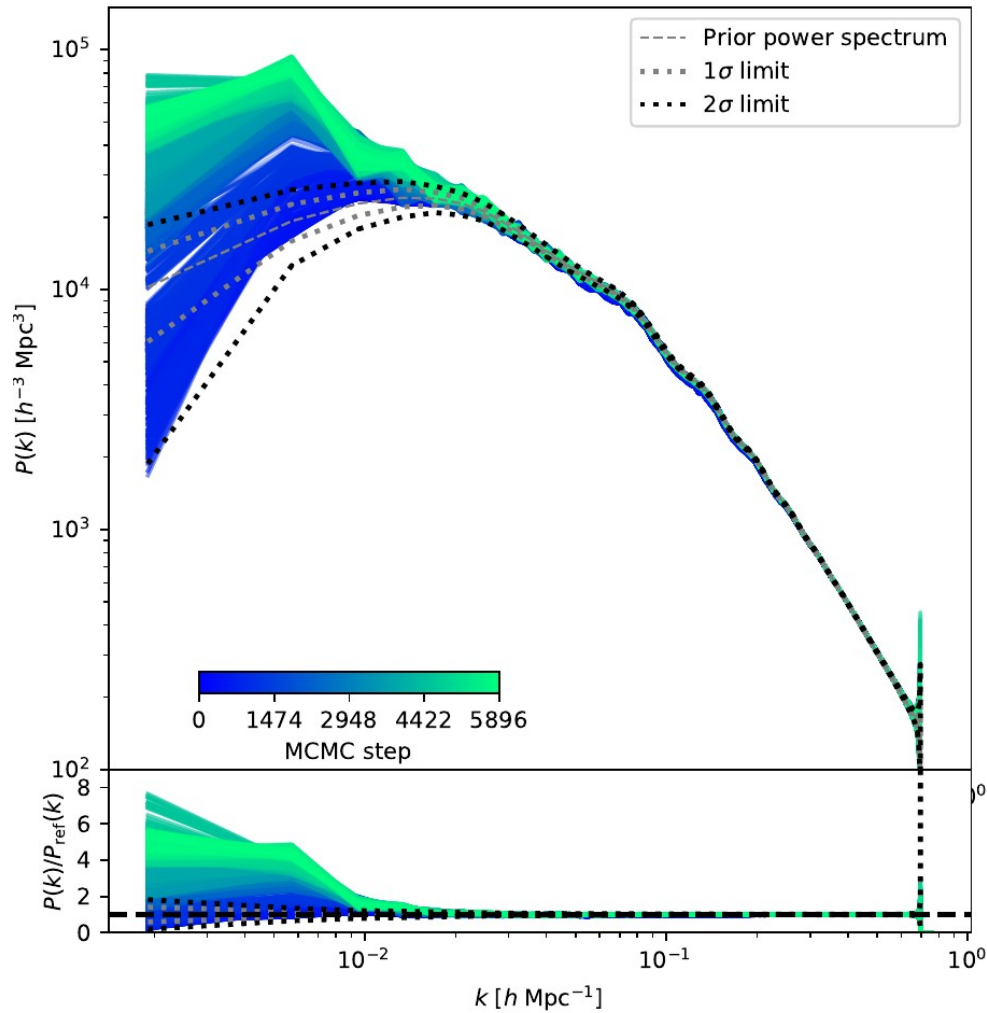


Contaminated mock data

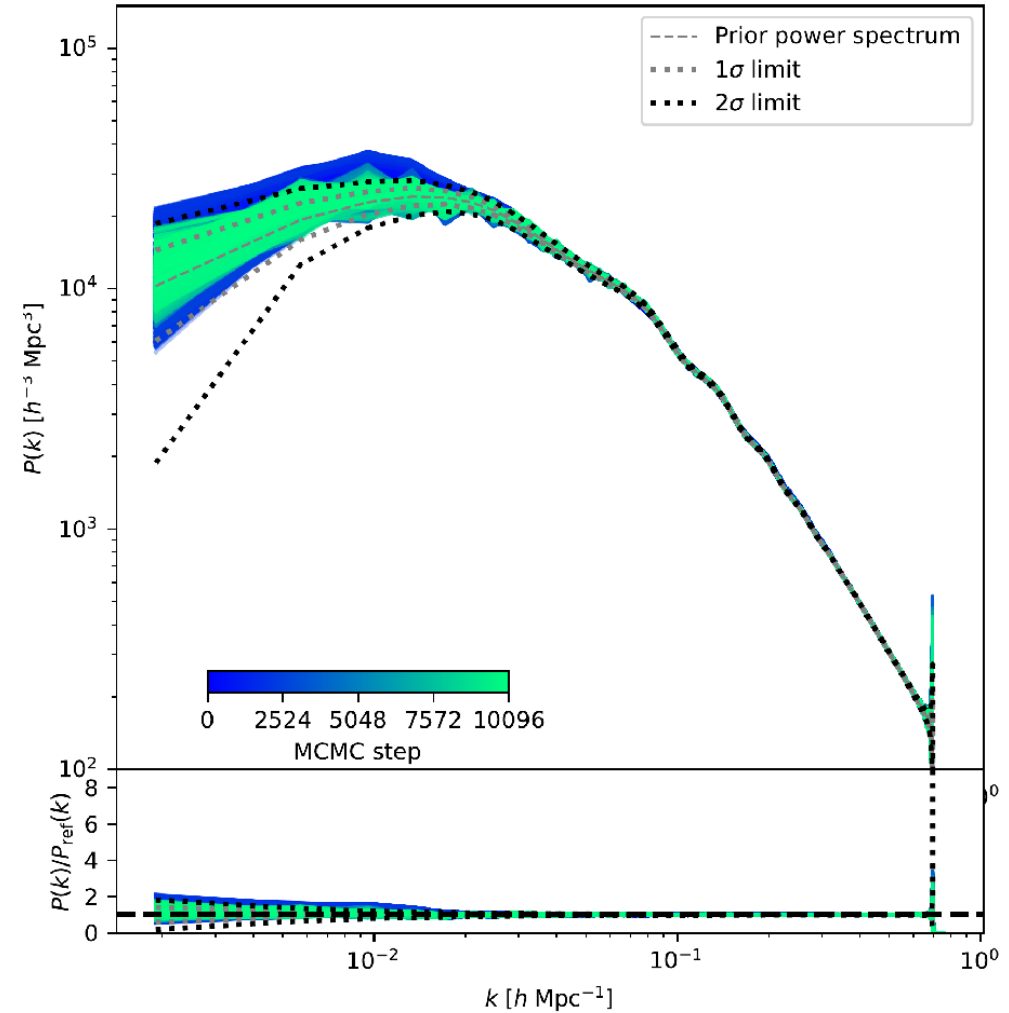


Poisson vs Robust likelihood

[Porqueres et al. 2019]



Poisson



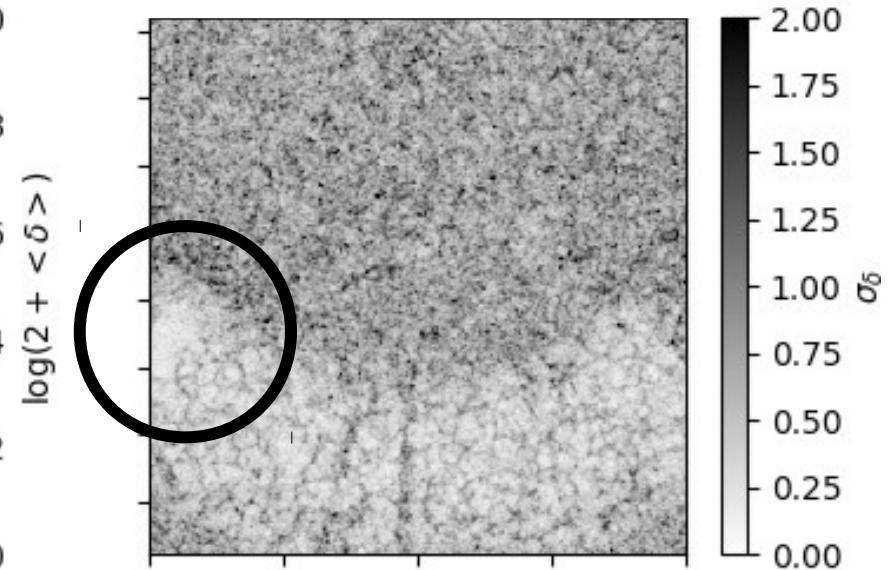
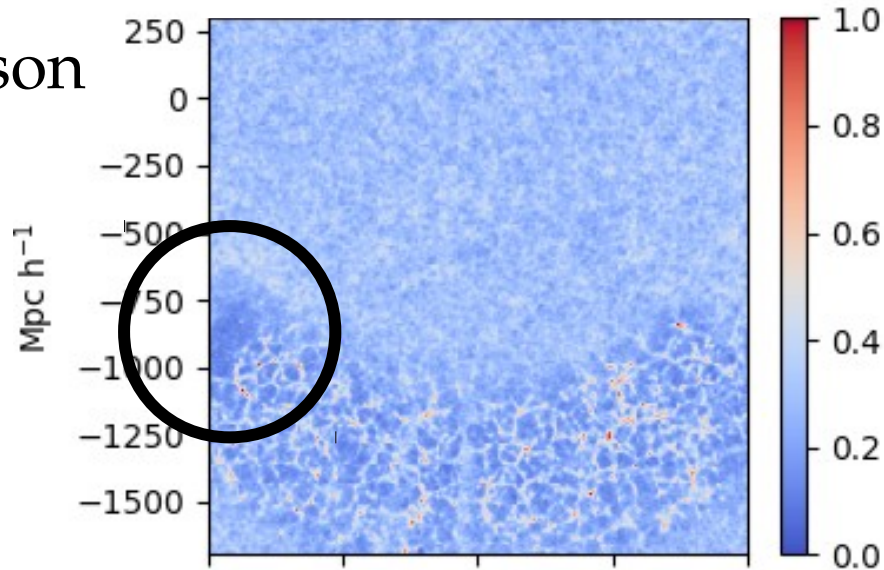
Robust

Poisson vs Robust likelihood

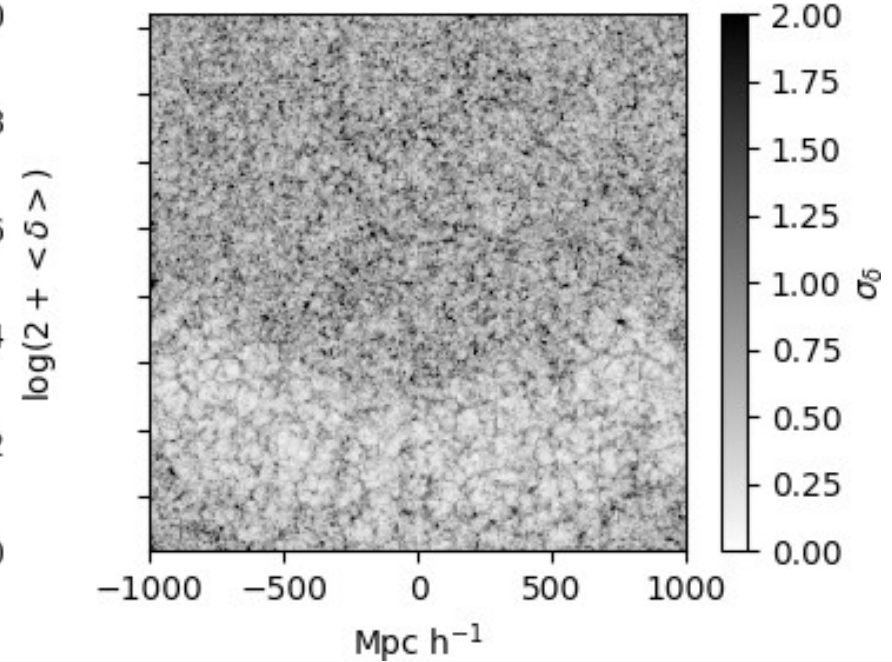
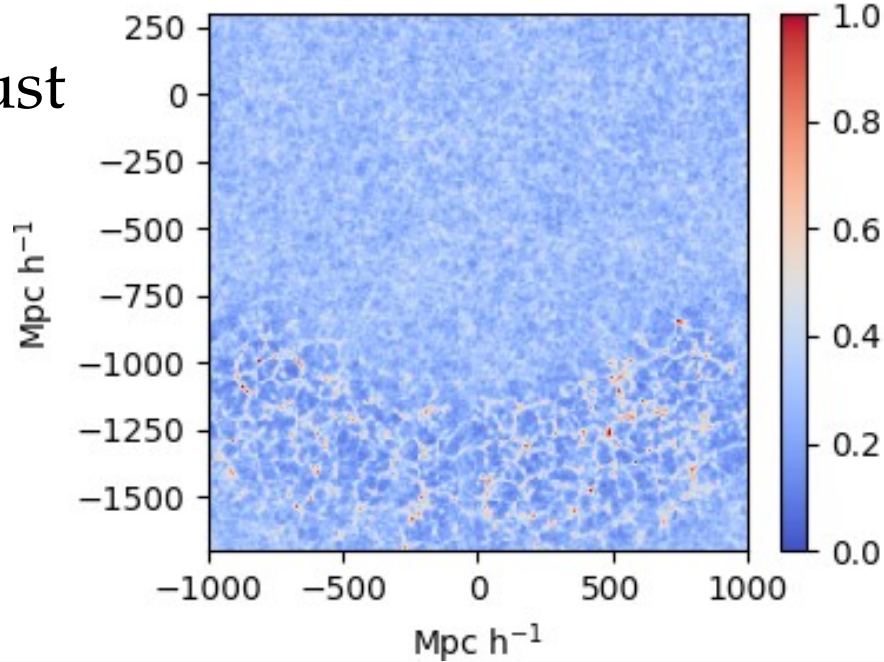
Mean

Variance

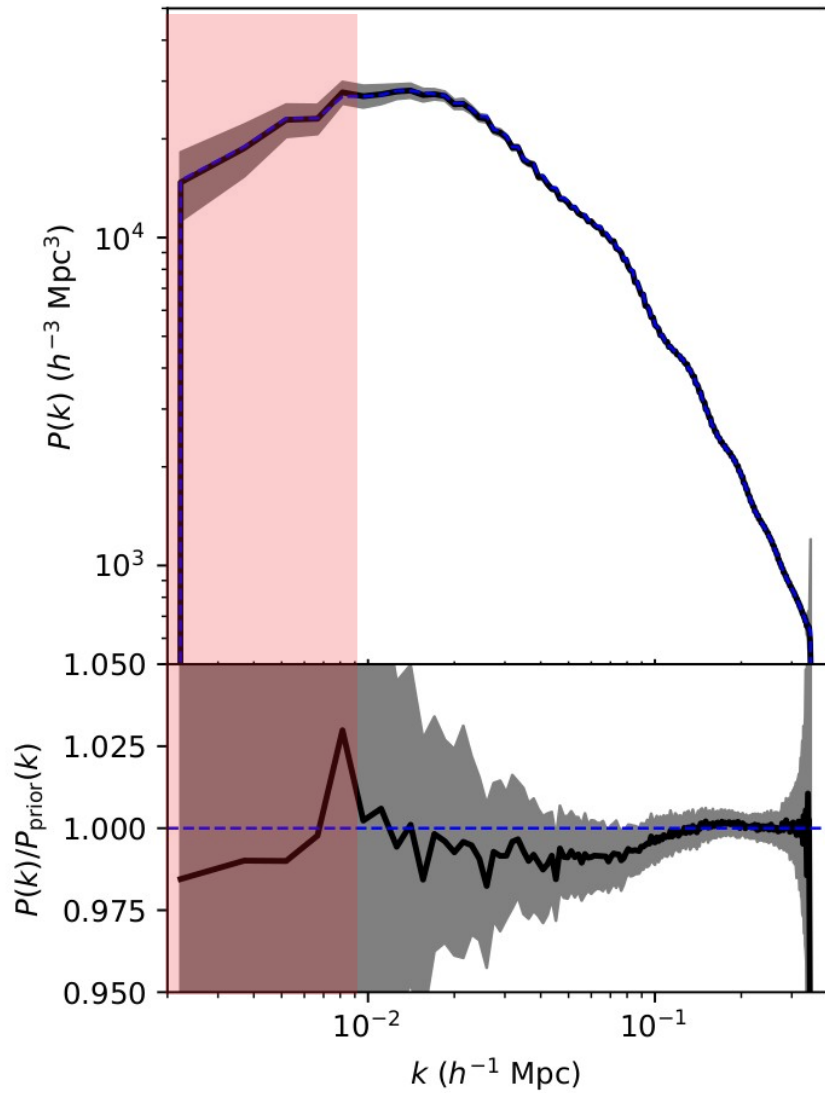
Poisson



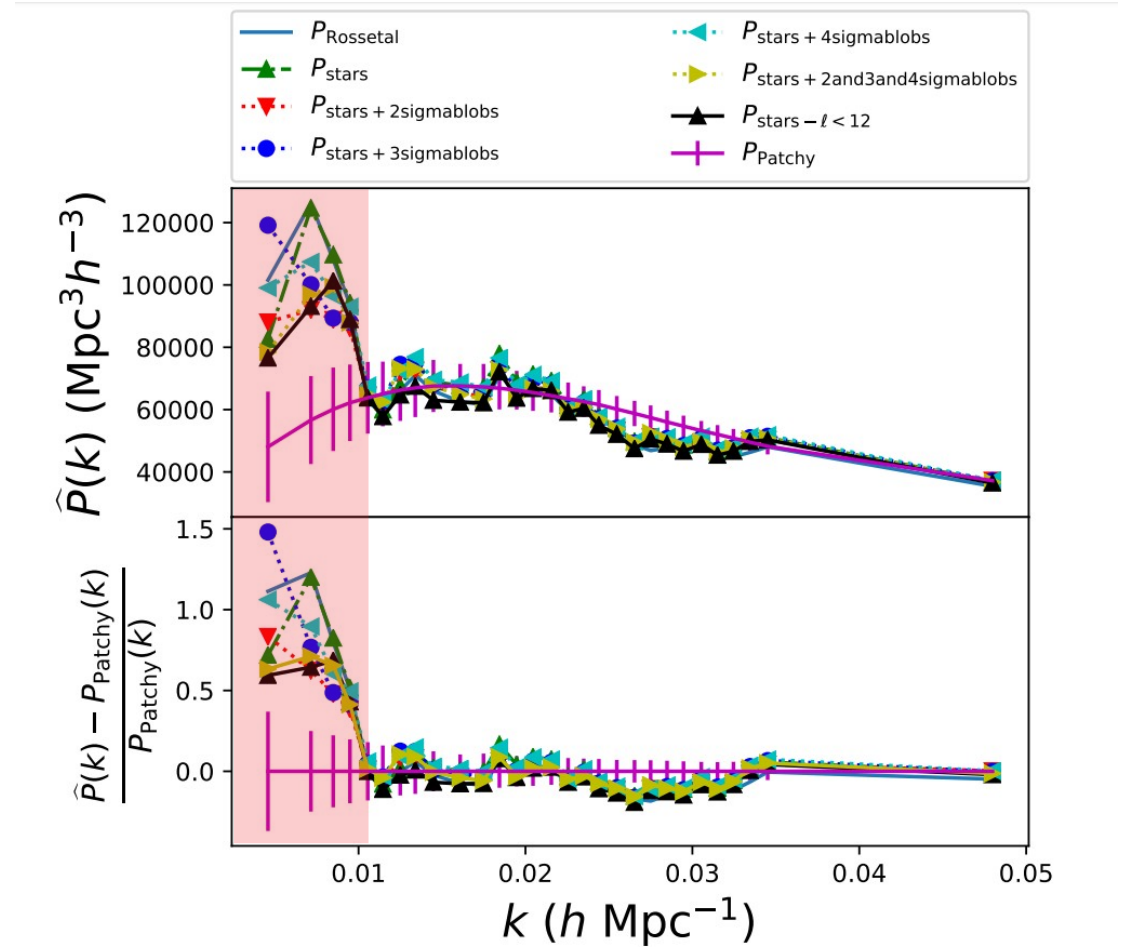
Robust



Robust likelihood and BOSS real data

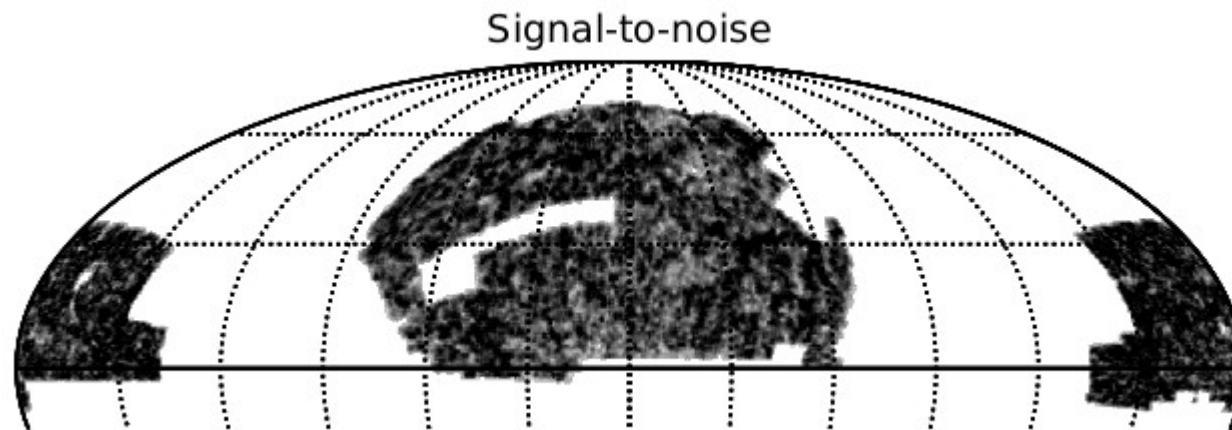
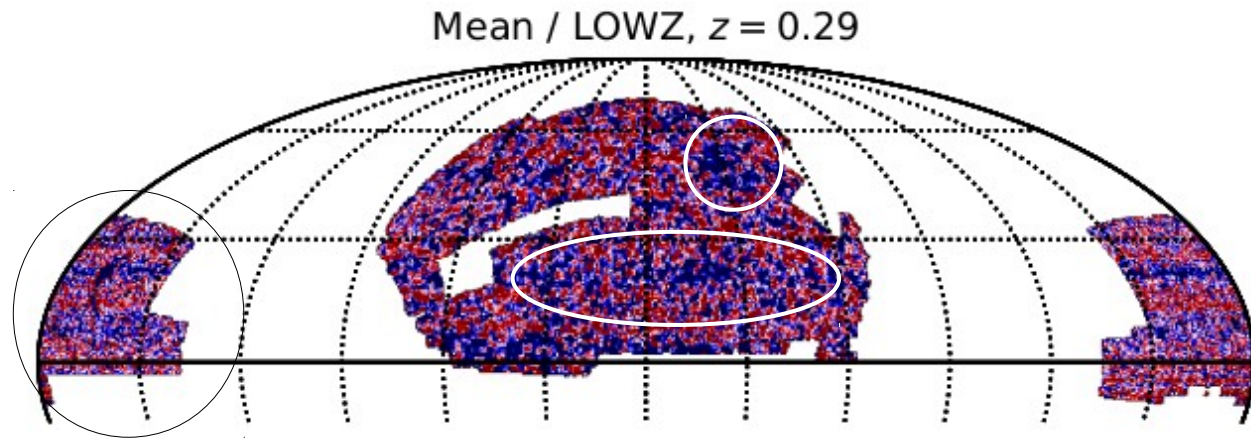


[Lavaux, Jasche, Leclercq 2019]



[Kalus, Percival et al 2018]

Templates of unknown foregrounds



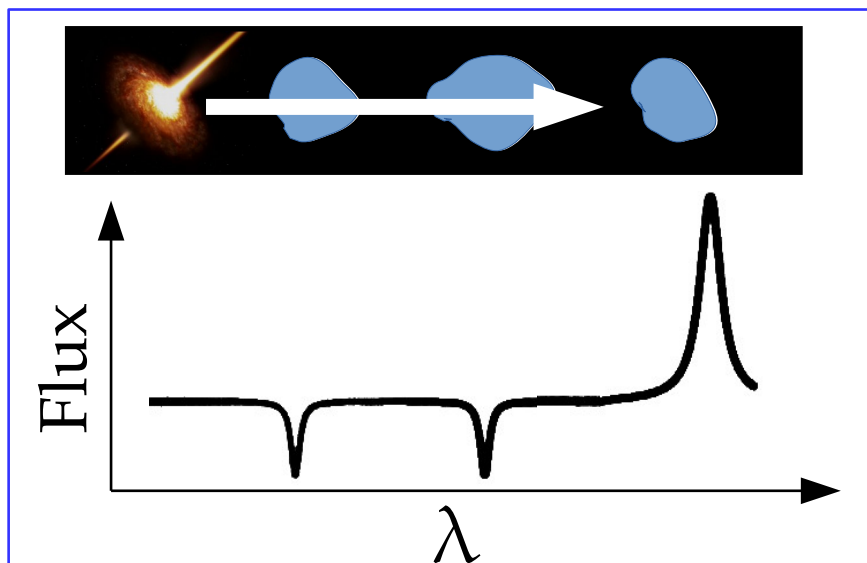
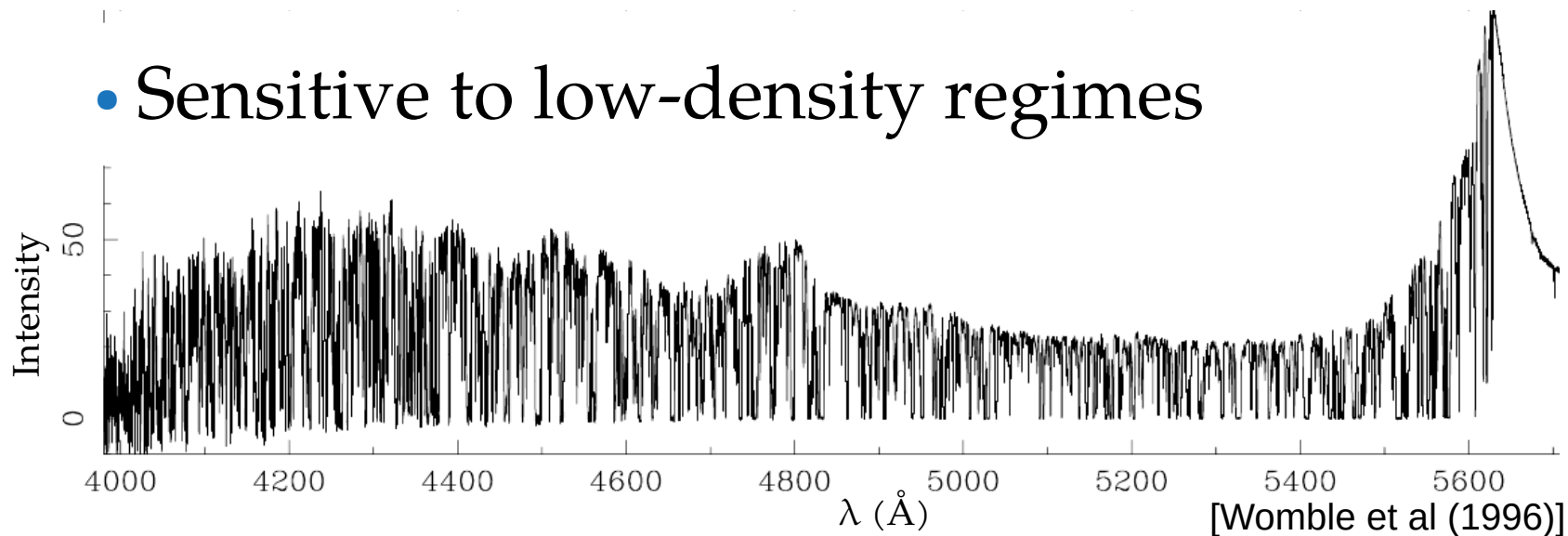
[Lavaux, Jasche, Leclercq 2019]

The Lyman- α forest:

A complementary source of information

The Lyman- α forest

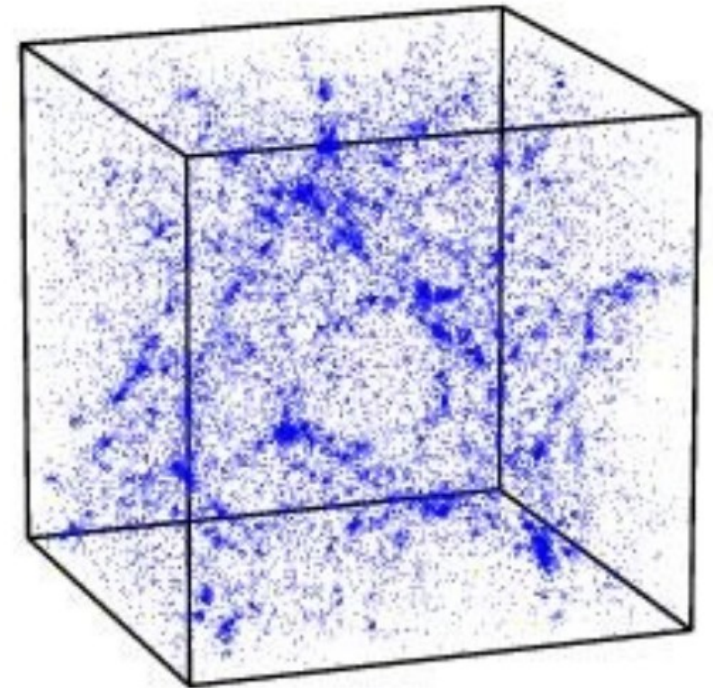
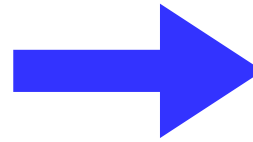
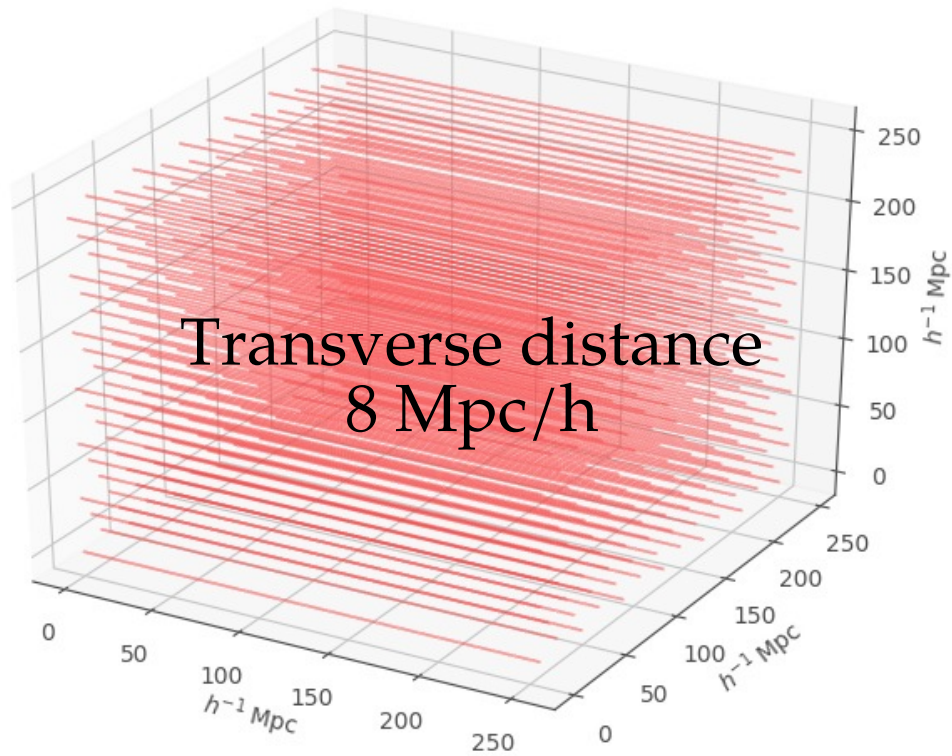
- High-resolution (Mpc) along the line of sight
- Sensitive to low-density regimes



Challenges:

- Non-gaussianity of data
- Interpolate between lines of sight

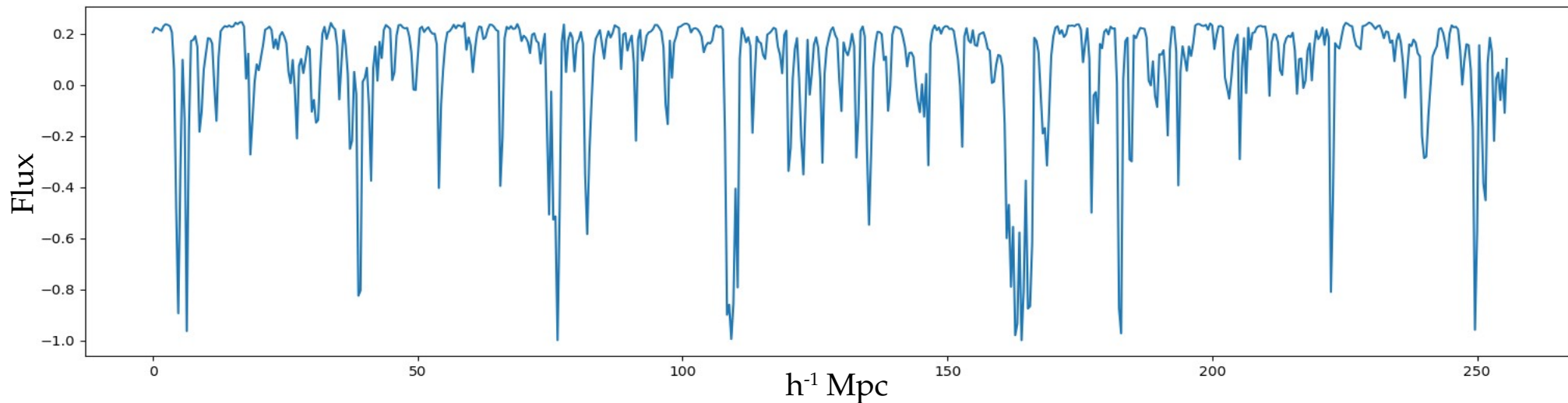
From 1D to 3D



Statistics!

$$P(\delta|F) = \frac{P(F|\delta)P(\delta)}{P(F)}$$

Data model

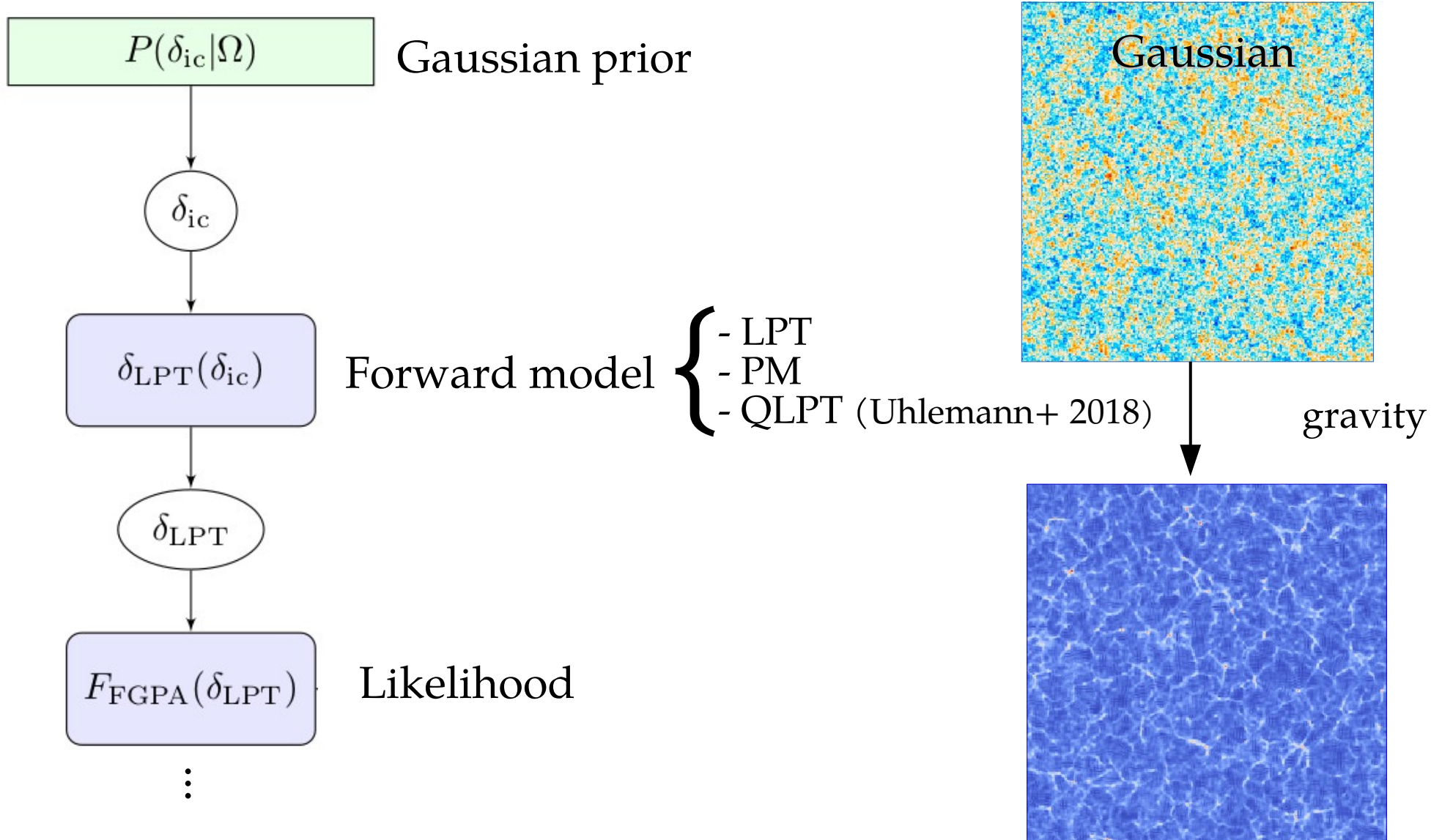


Likelihood

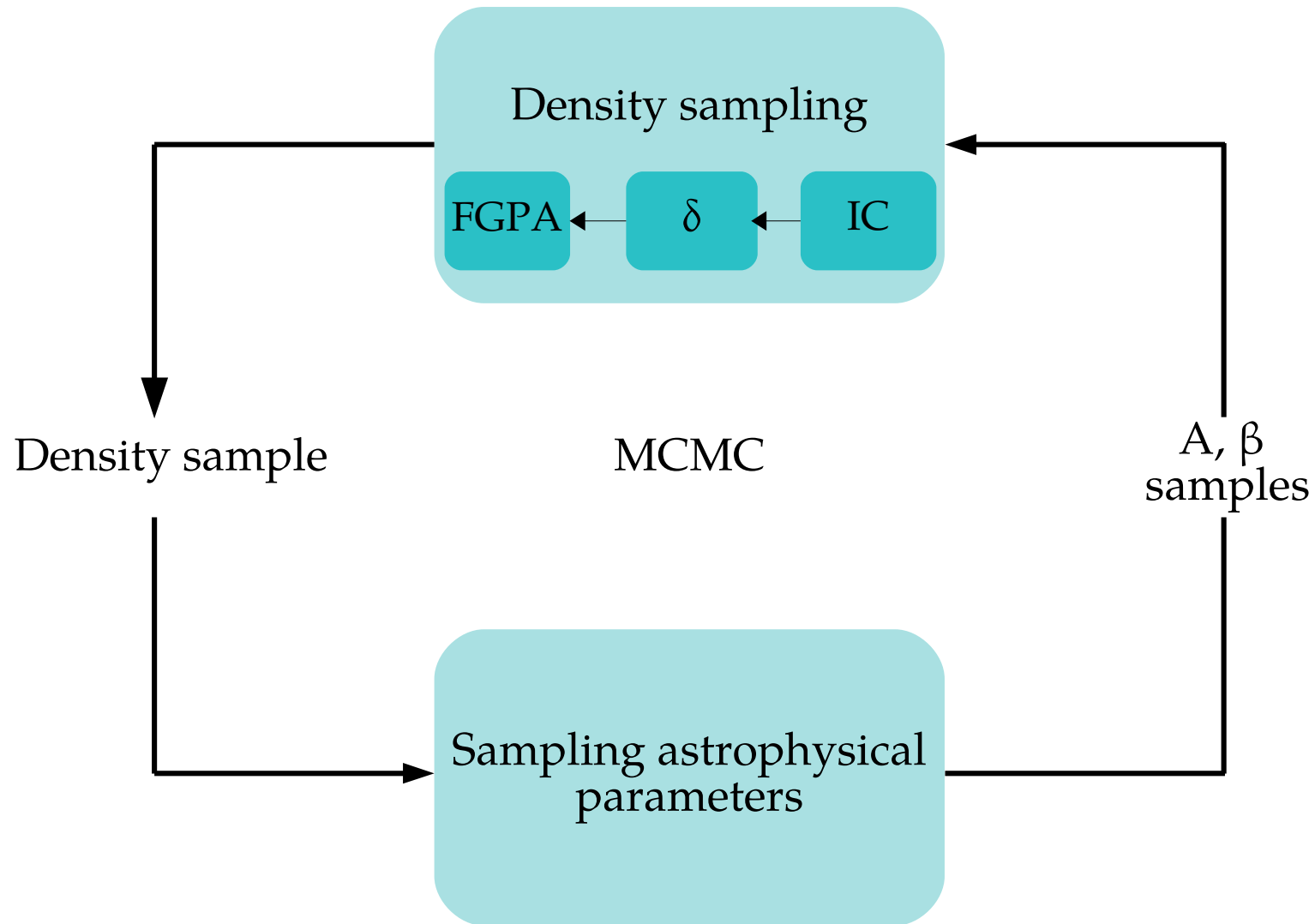
$$-\log P(F|\delta) = \frac{\left(F - \exp[-A(1 + \delta)^\beta]\right)^2}{2\sigma^2}$$

Sampling density and astrophysical parameters

The BORG framework



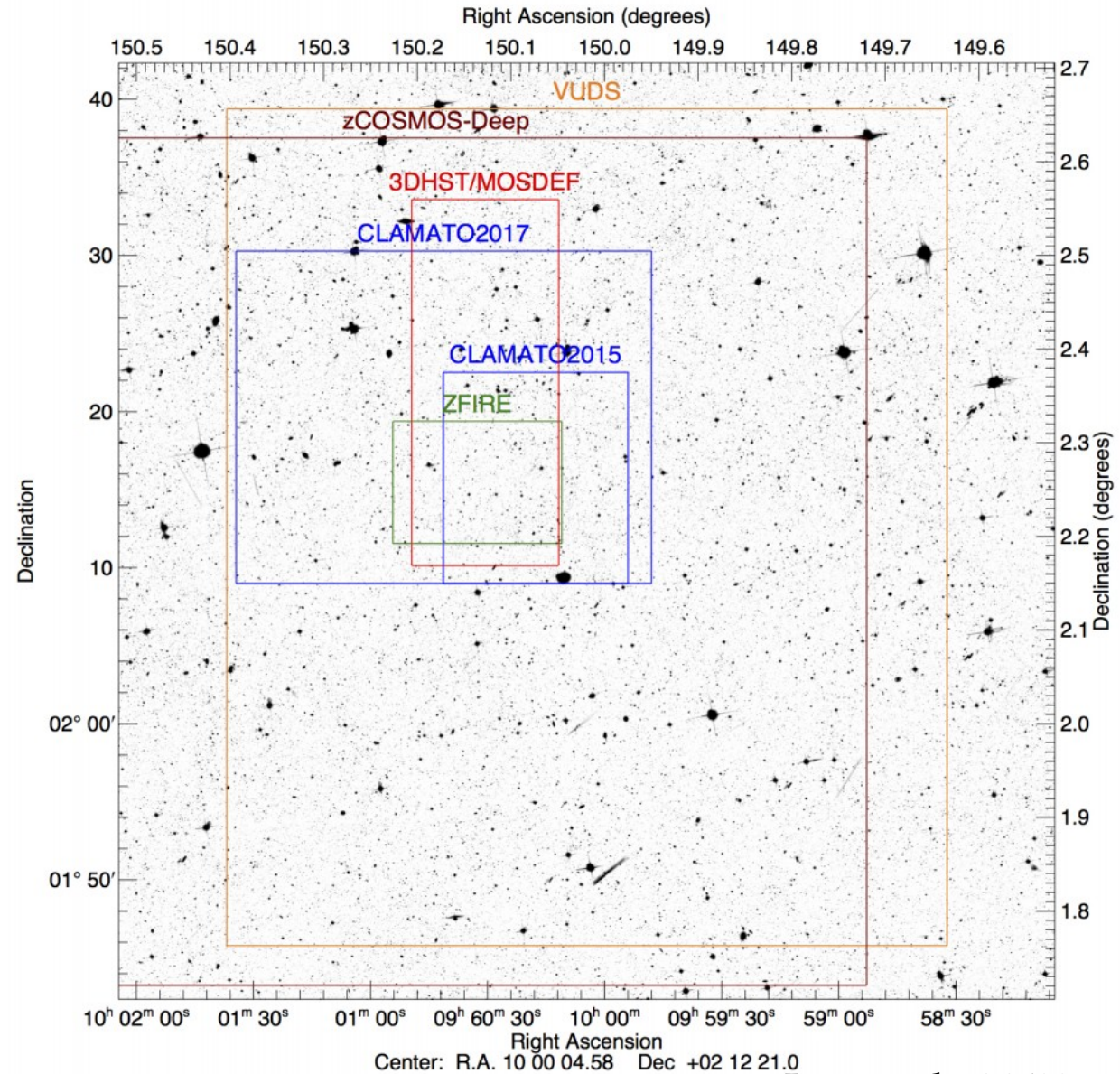
Statistical modular framework



Testing with mock CLAMATO data

Small sky area (0.8 deg^2)

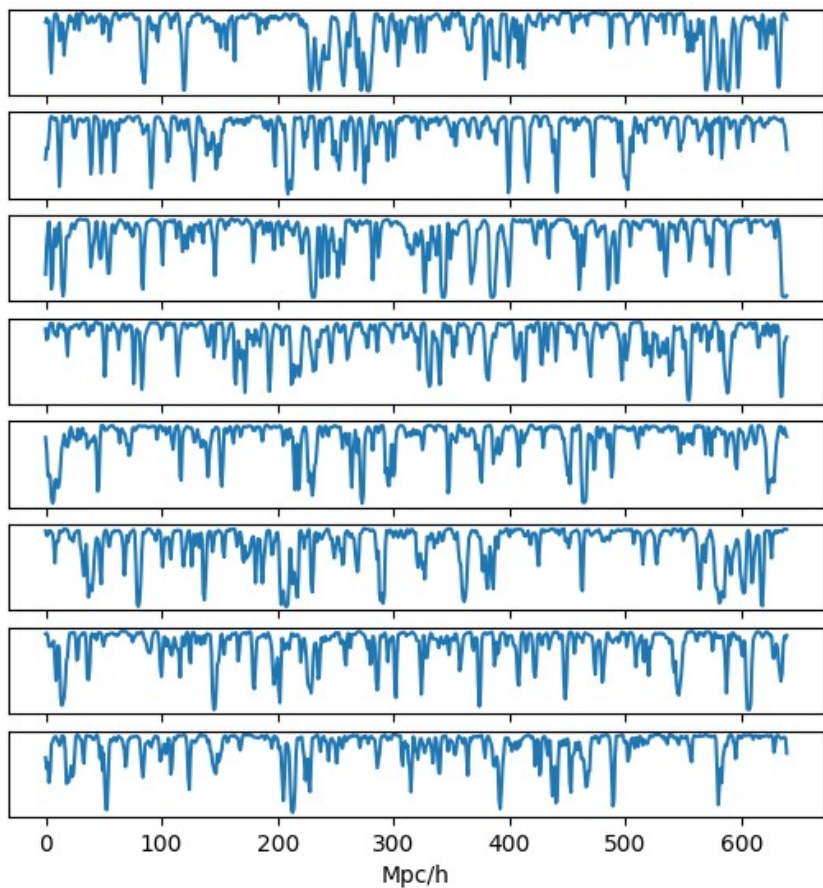
High density of los
(more than 400 spectra)



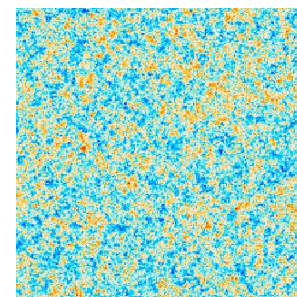
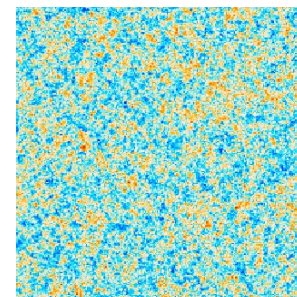
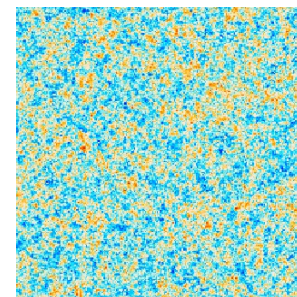
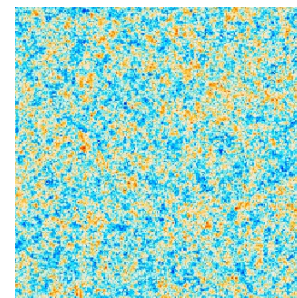
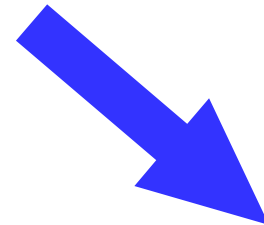
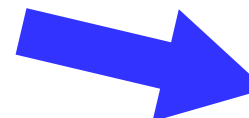
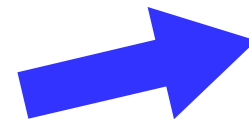
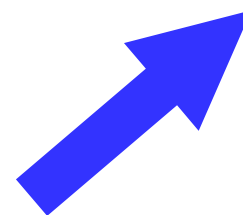
[Lee et al. 2017]

Inference of DM density

simulated QSO spectra

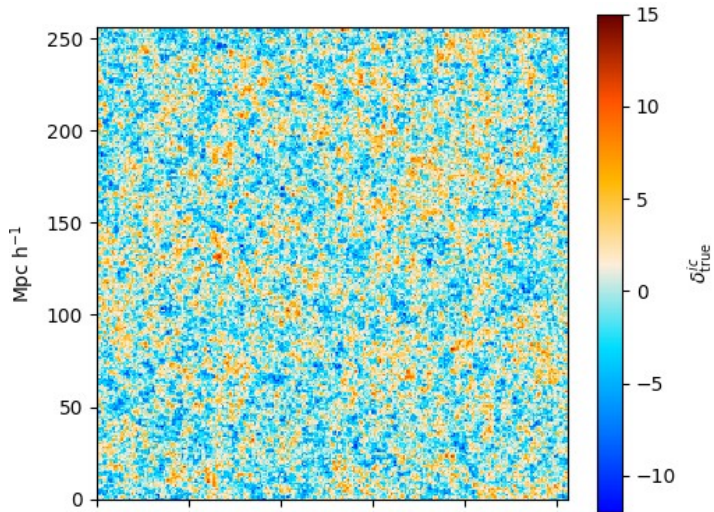


BORG

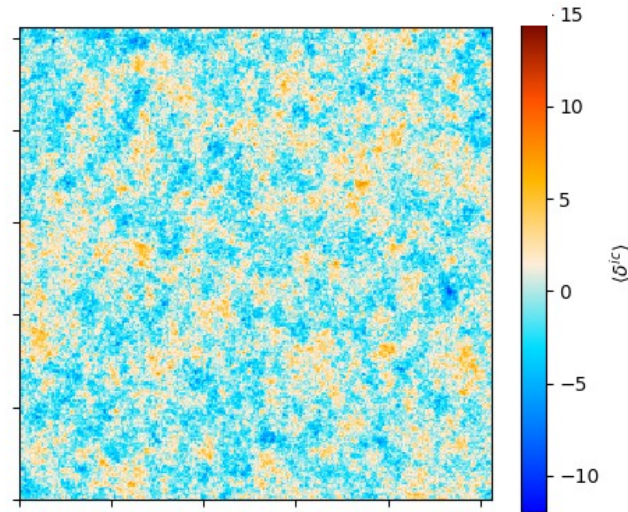


$L = 256 \text{ Mpc/h}$
resolution 1 Mpc/h

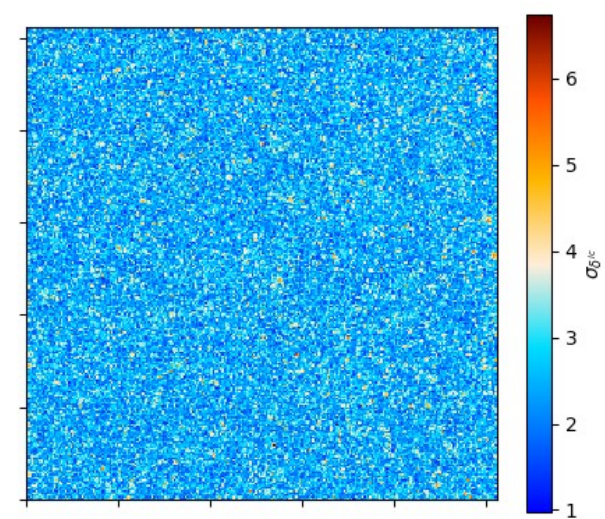
True initial conditions



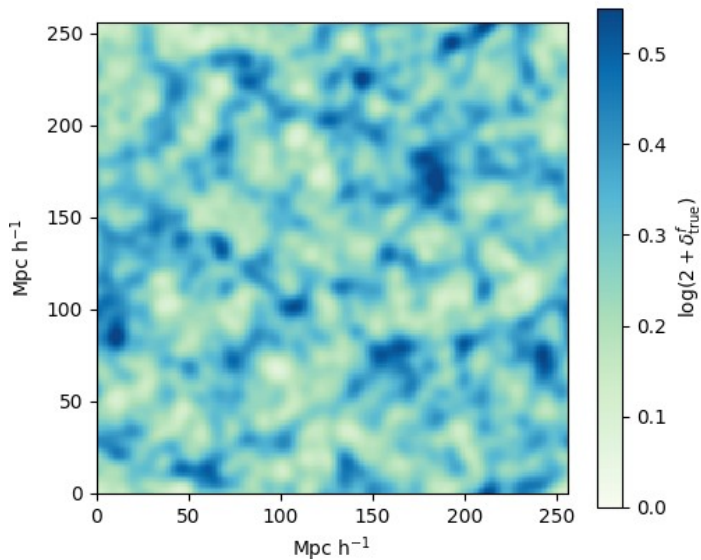
Mean initial conditions



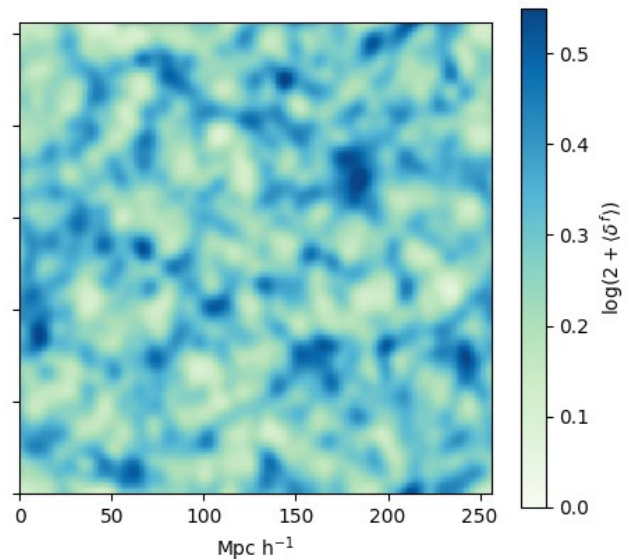
Variance initial conditions



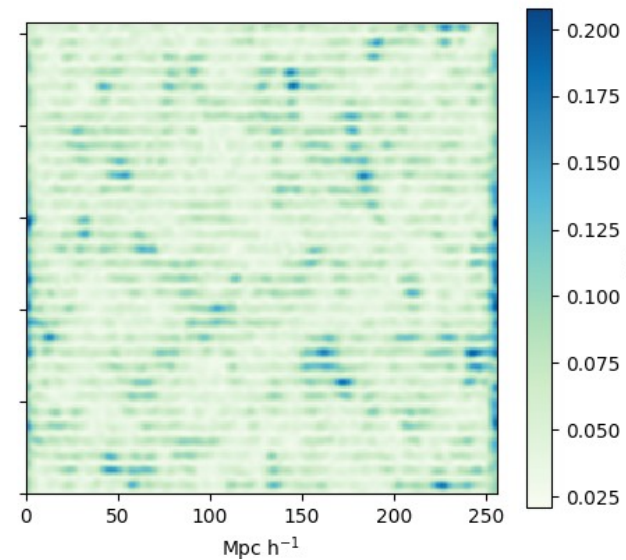
True density



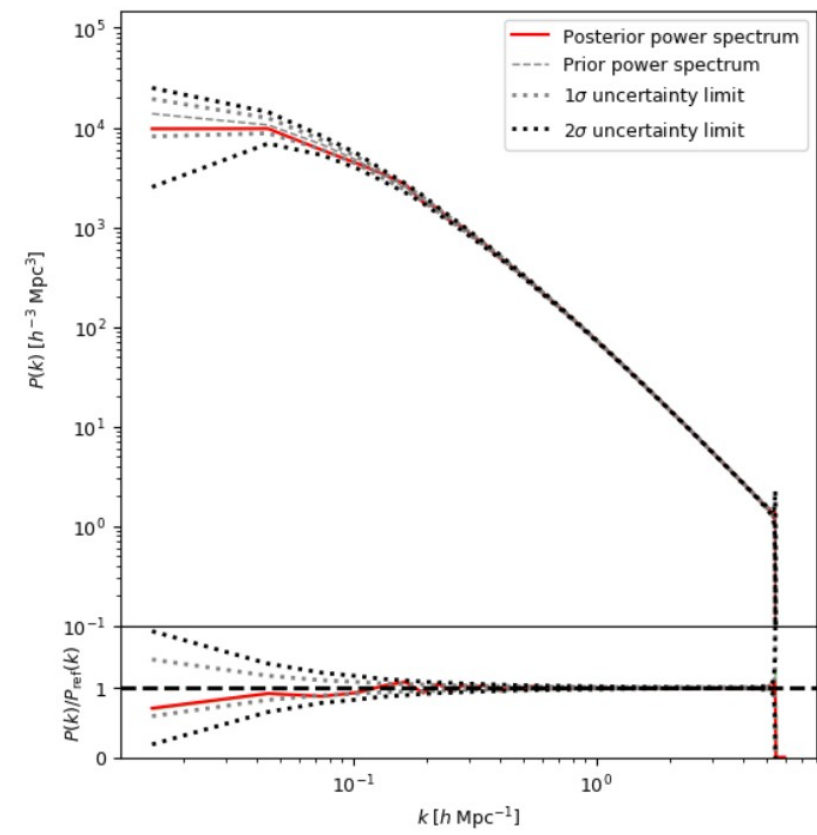
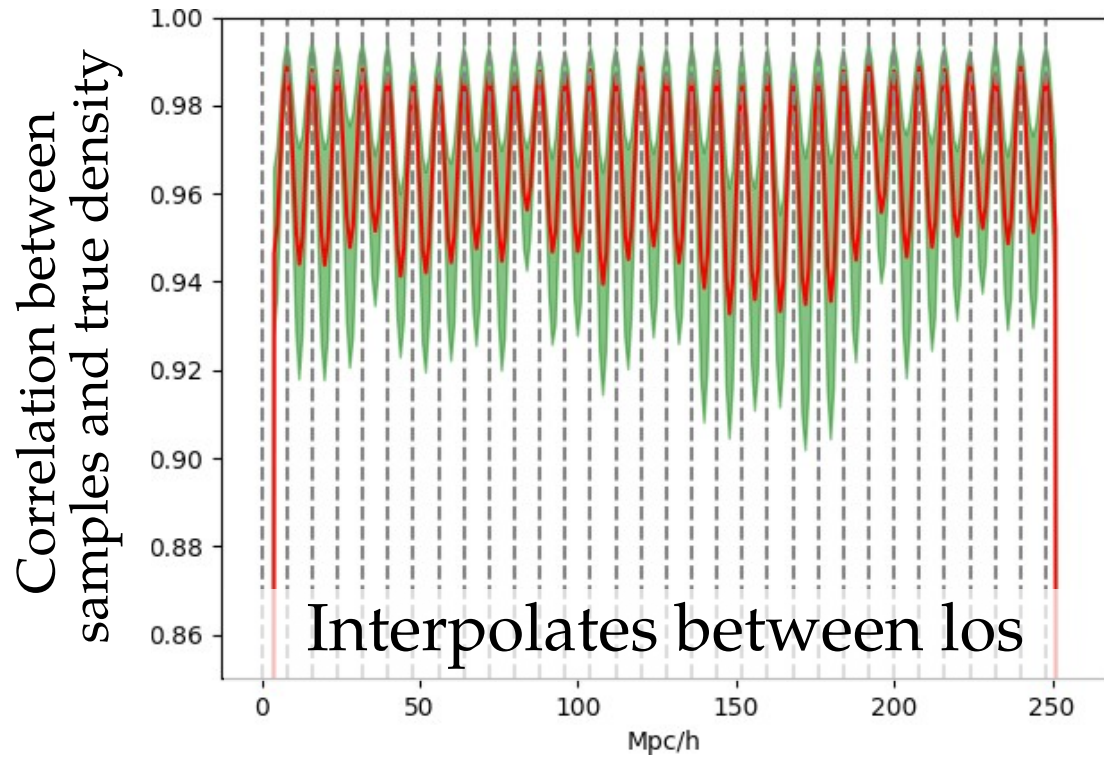
Mean density



Variance density

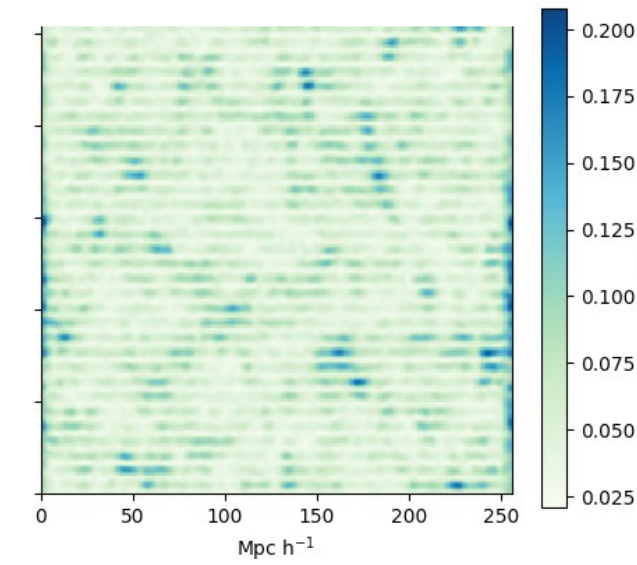
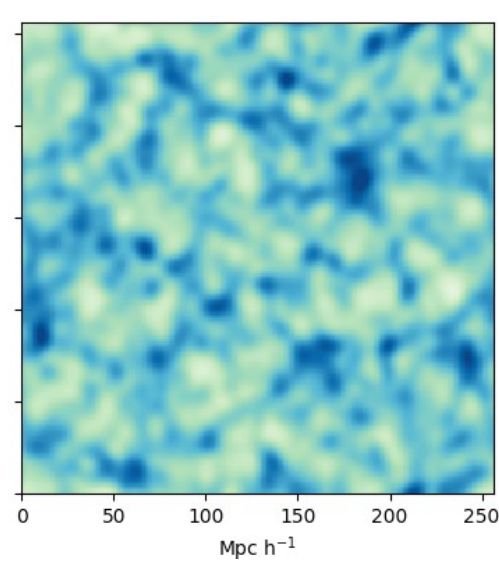
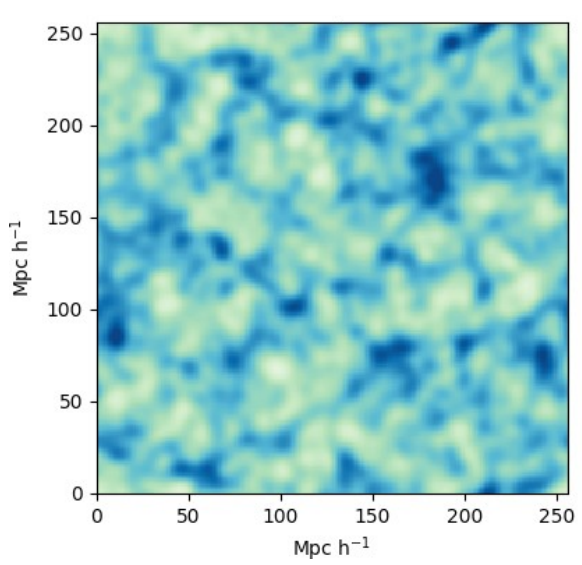


[Porqueres et al. 2019]



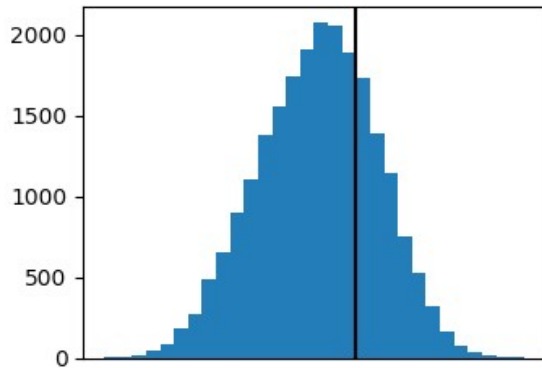
True density

Mean density

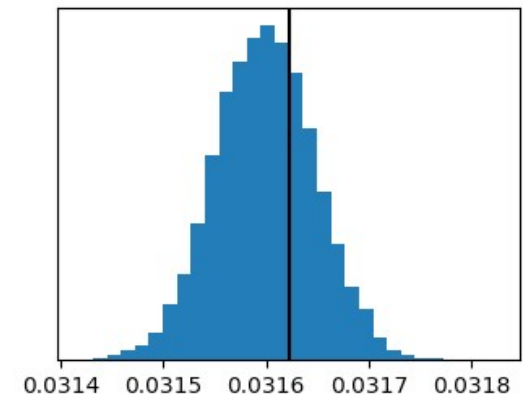
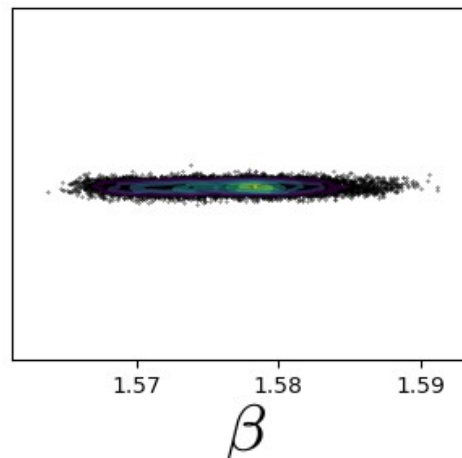
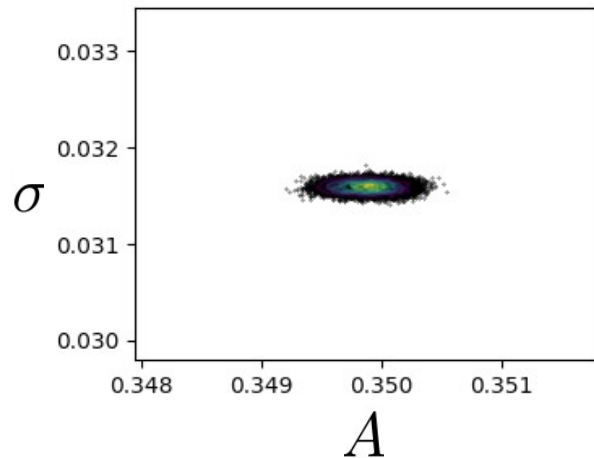
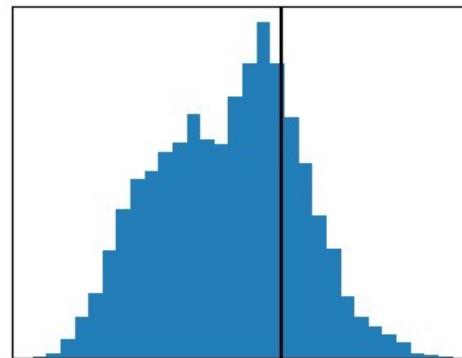
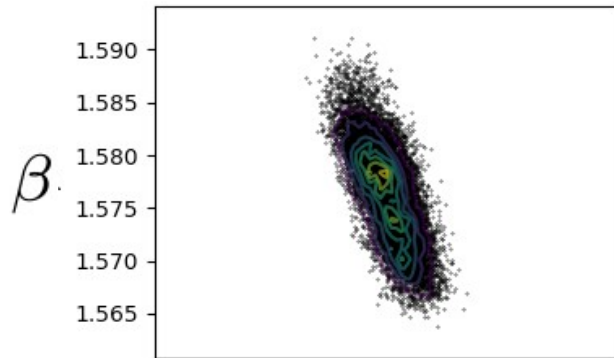


[Porqueres et al. 2019]

Astrophysical parameters

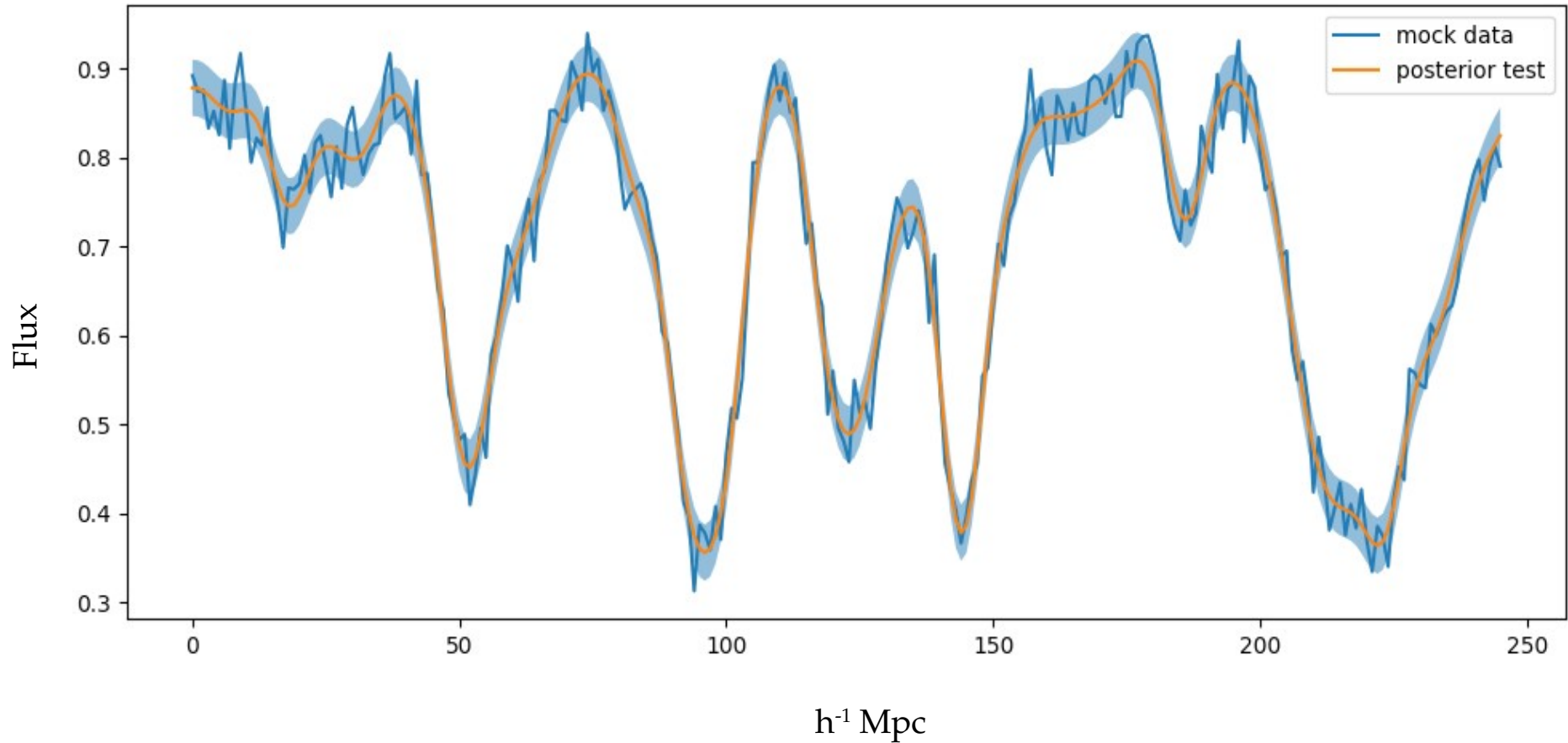


$$-\log P(F|\delta) = \frac{\left(F - \exp[-A(1 + \delta)^\beta]\right)^2}{2\sigma^2}$$



[Porqueres et al. 2019]

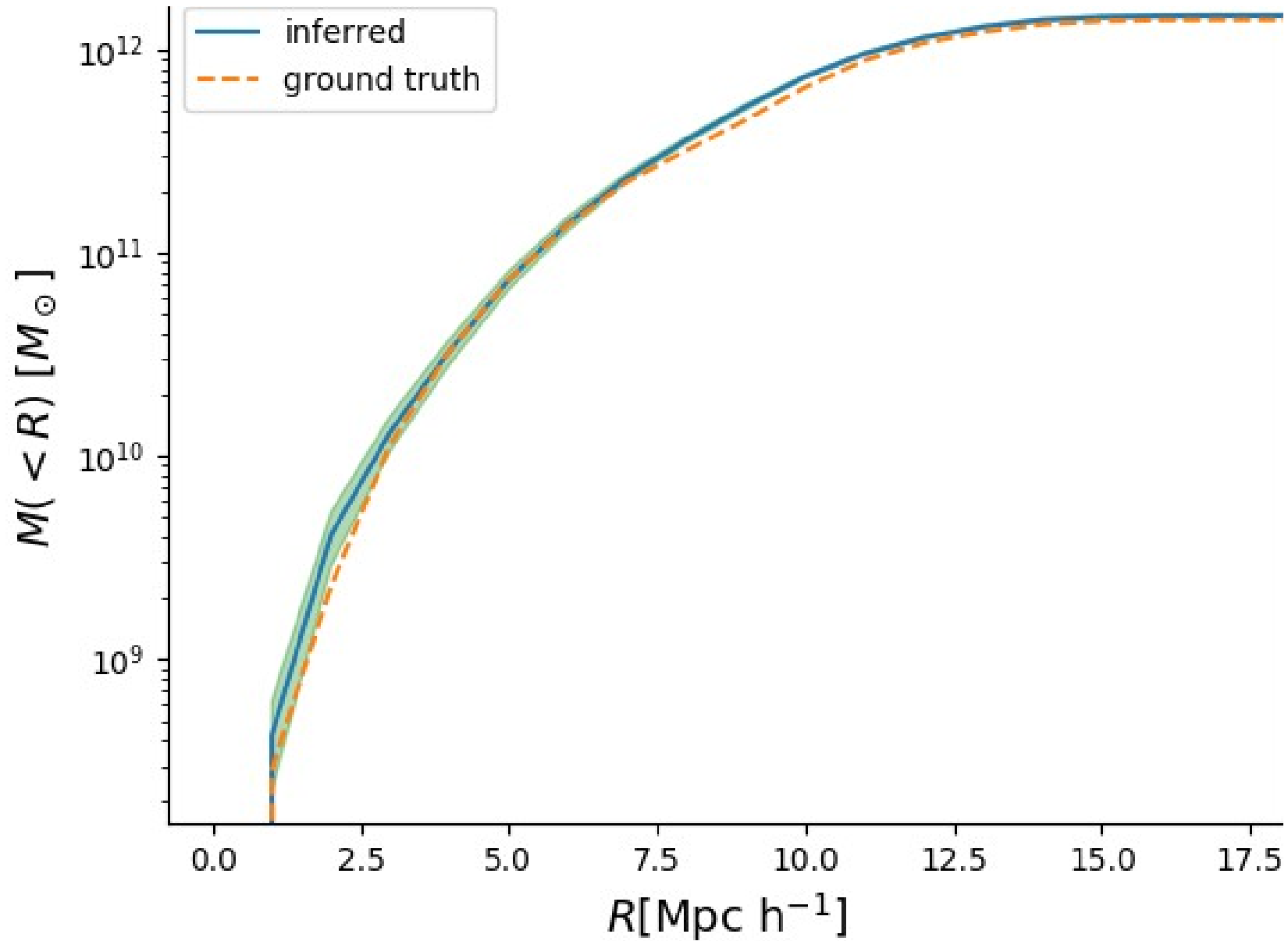
Posterior predictive test



[Porqueres et al. 2019]

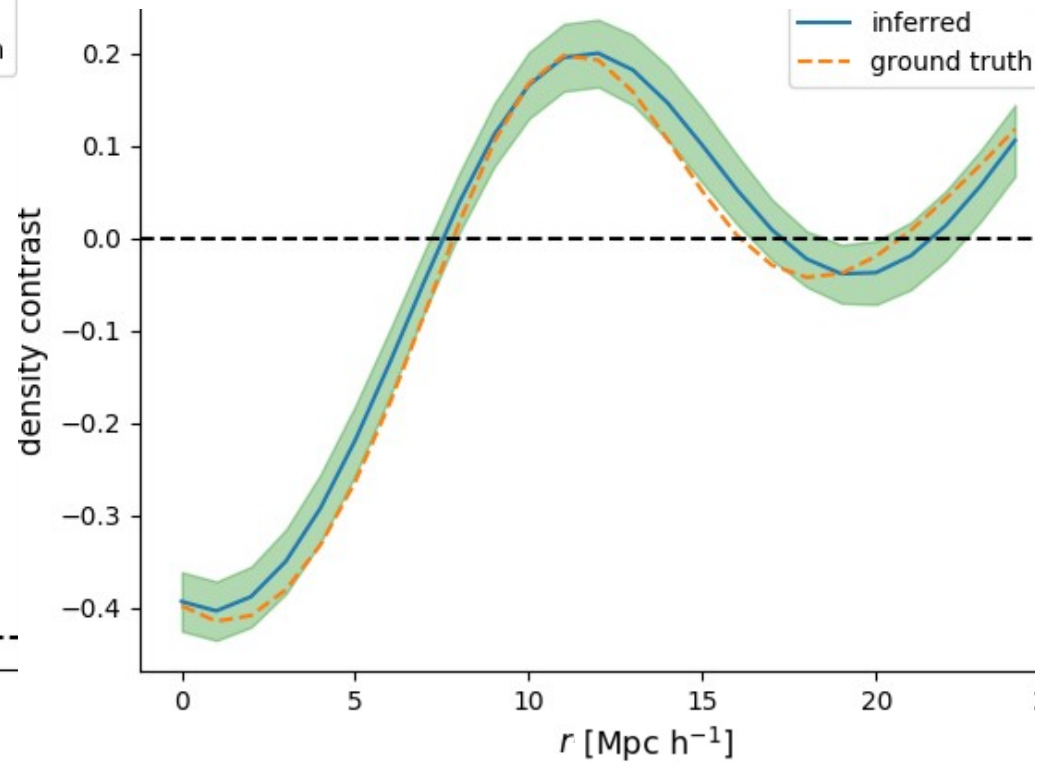
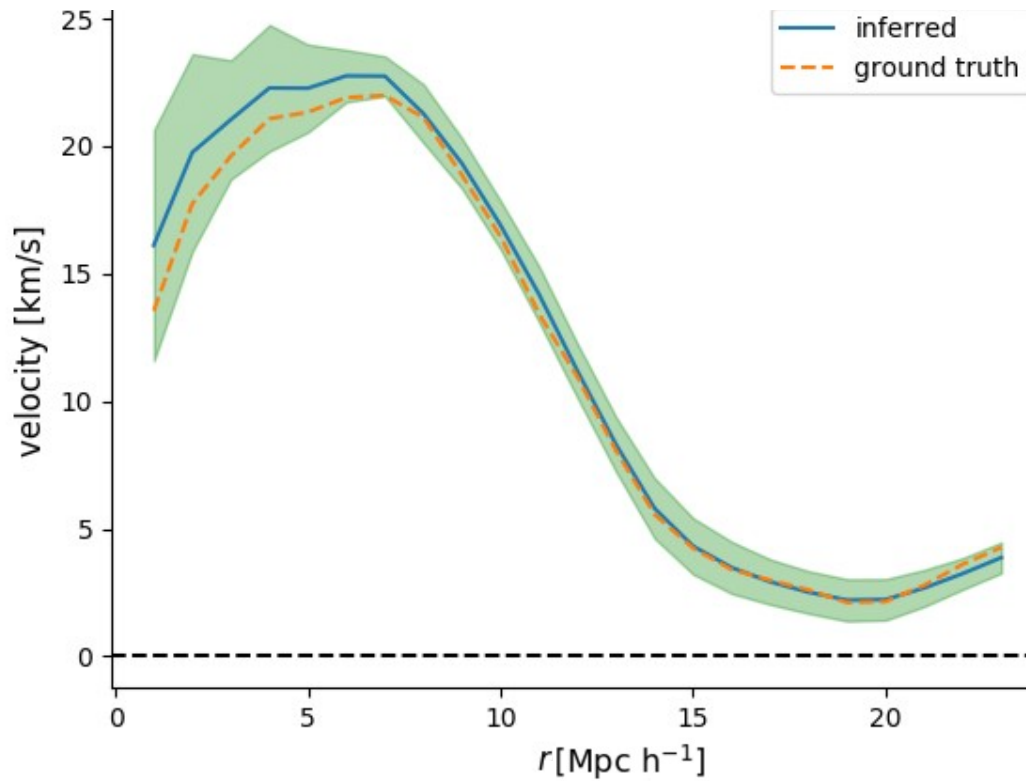
What physics can we do
with this method?

Cluster mass profile



[Porqueres et al. 2019]

Void profiles

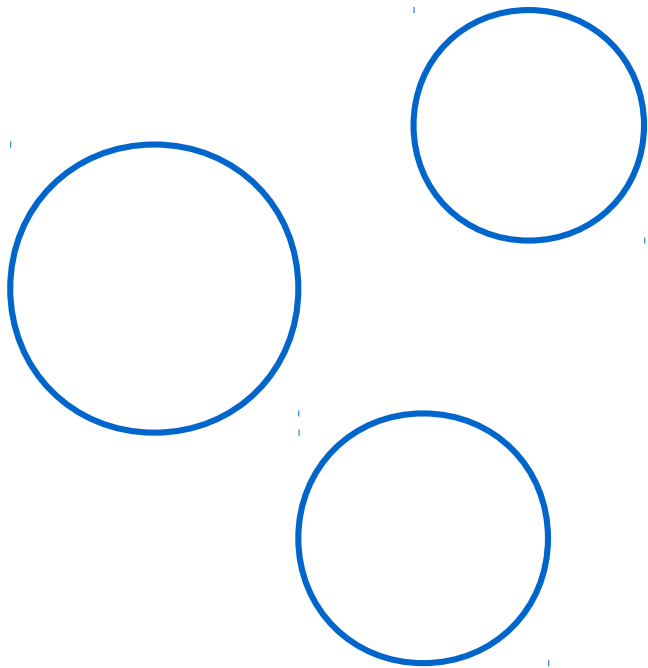


[Porqueres et al. 2019]

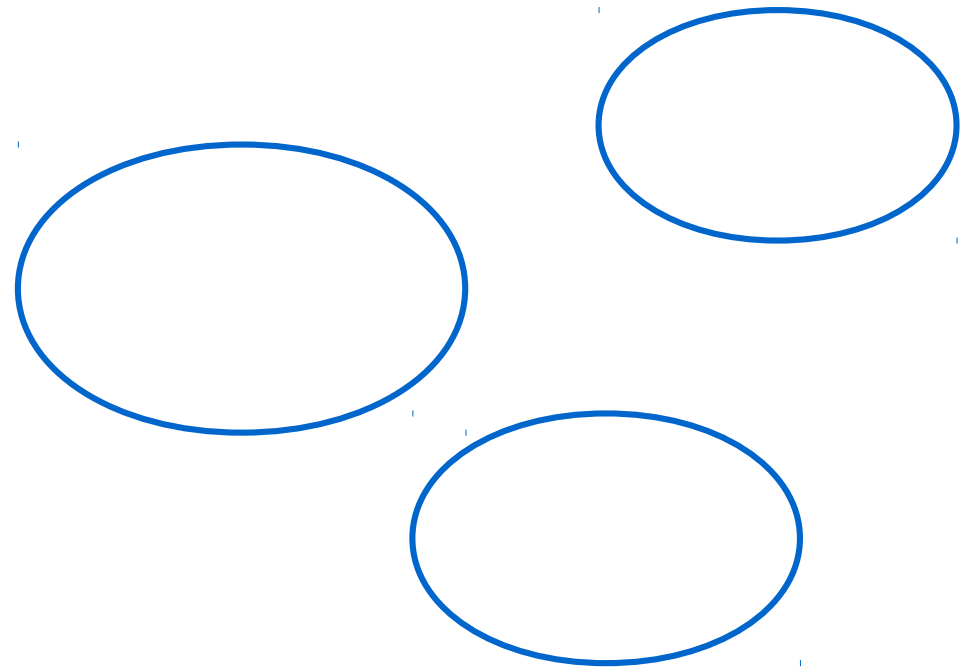
Alcock-Paczynski test

Isotropy of the Universe

Constrain $d(z, \Omega)$ relation

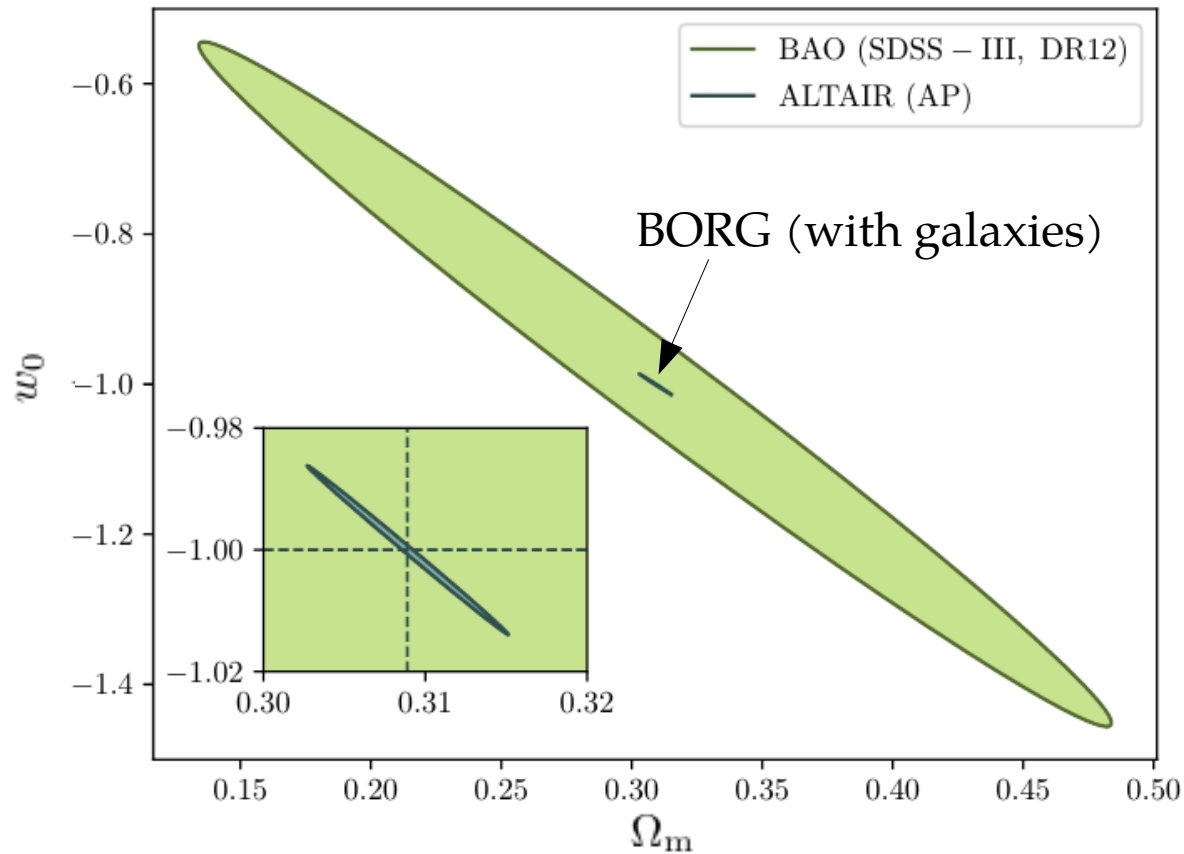


Correct cosmology

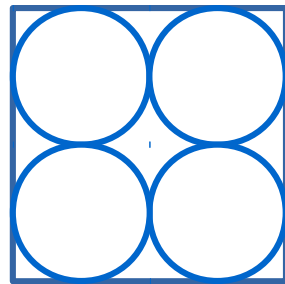


Wrong cosmology

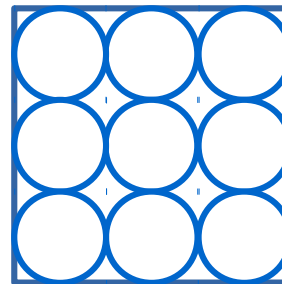
Cosmological parameters



[Ramanah et al. 2018]

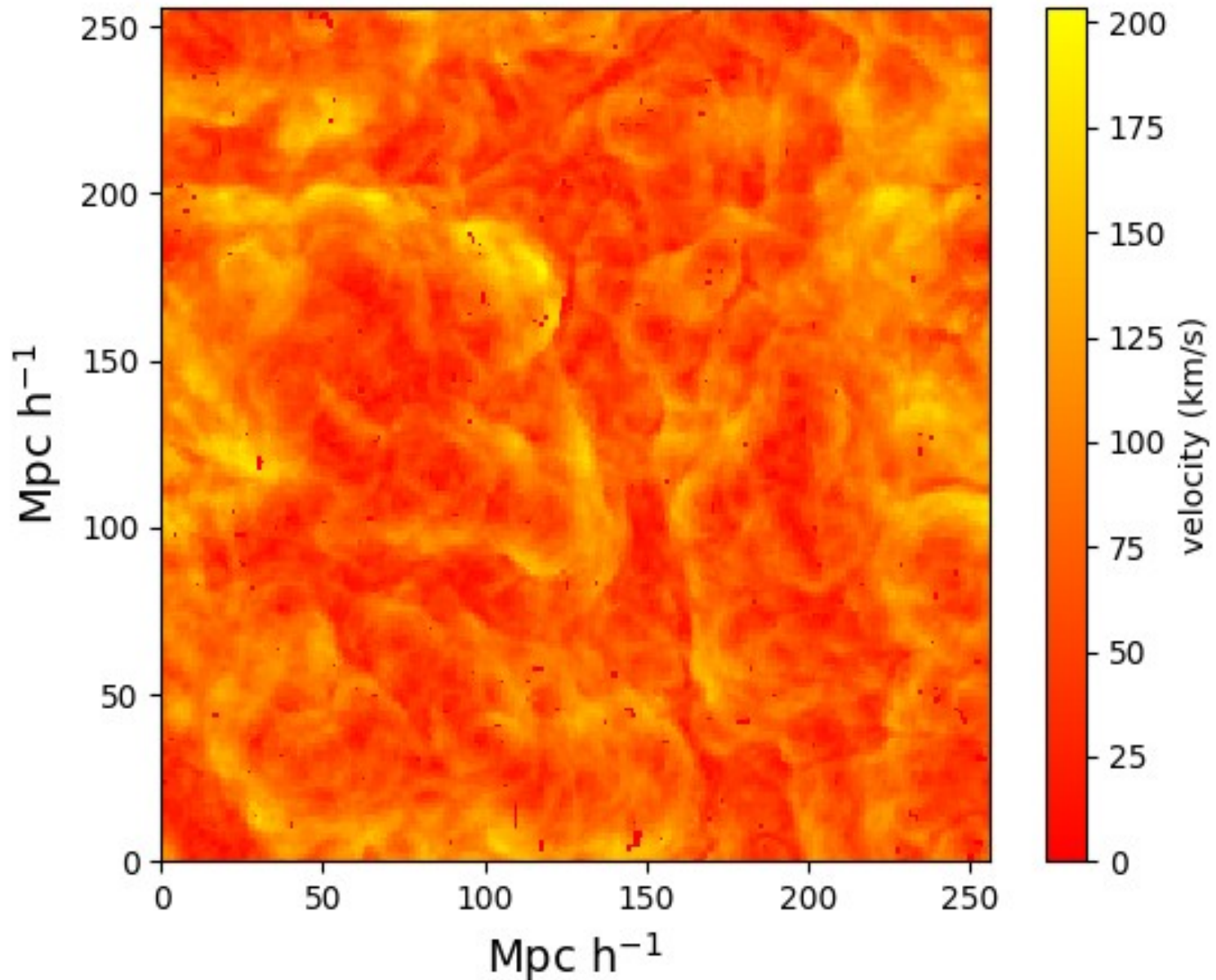


BAO
150 Mpc



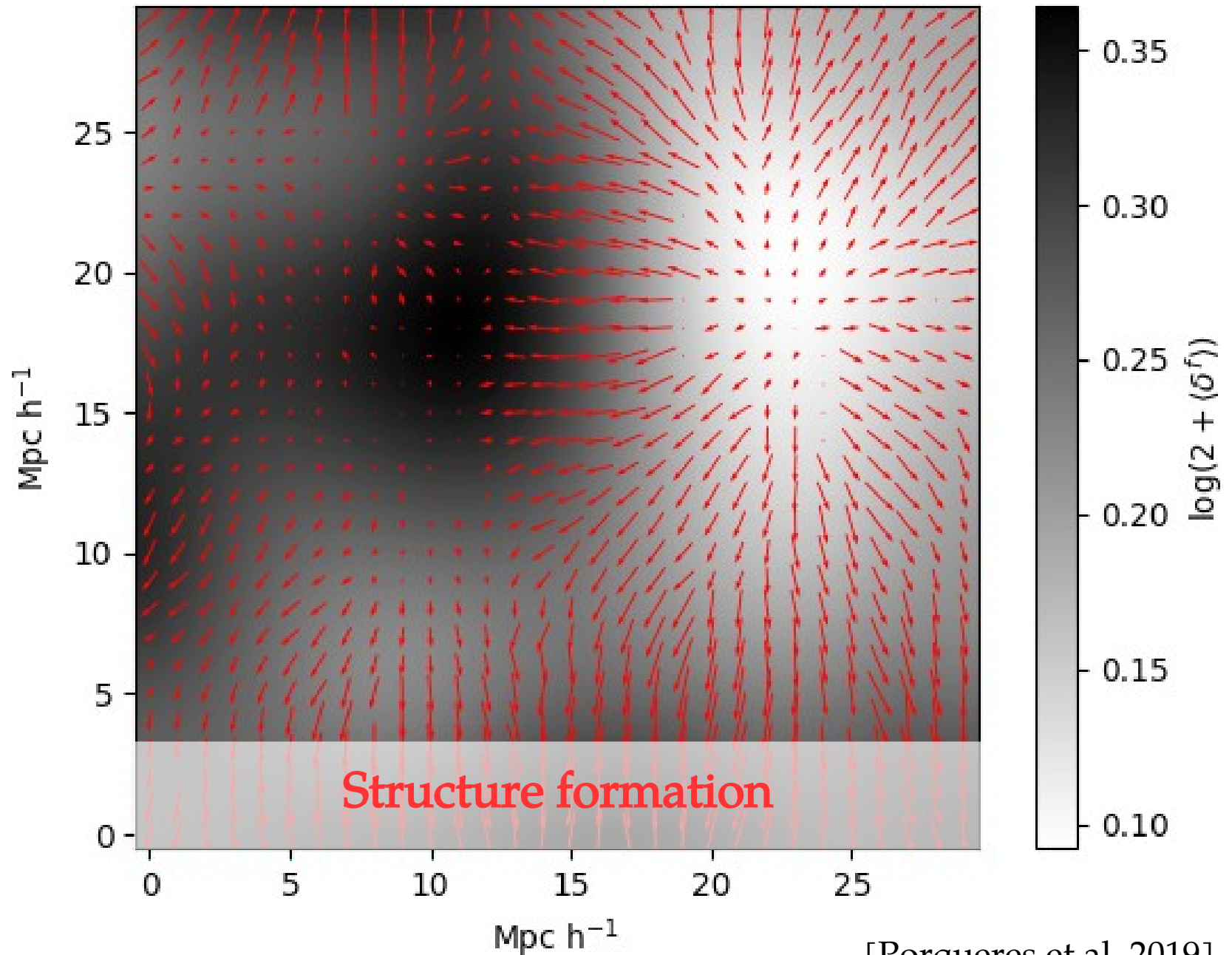
AP test
Smaller scales

Velocity field at $z > 2$



[Porqueres et al. 2019]

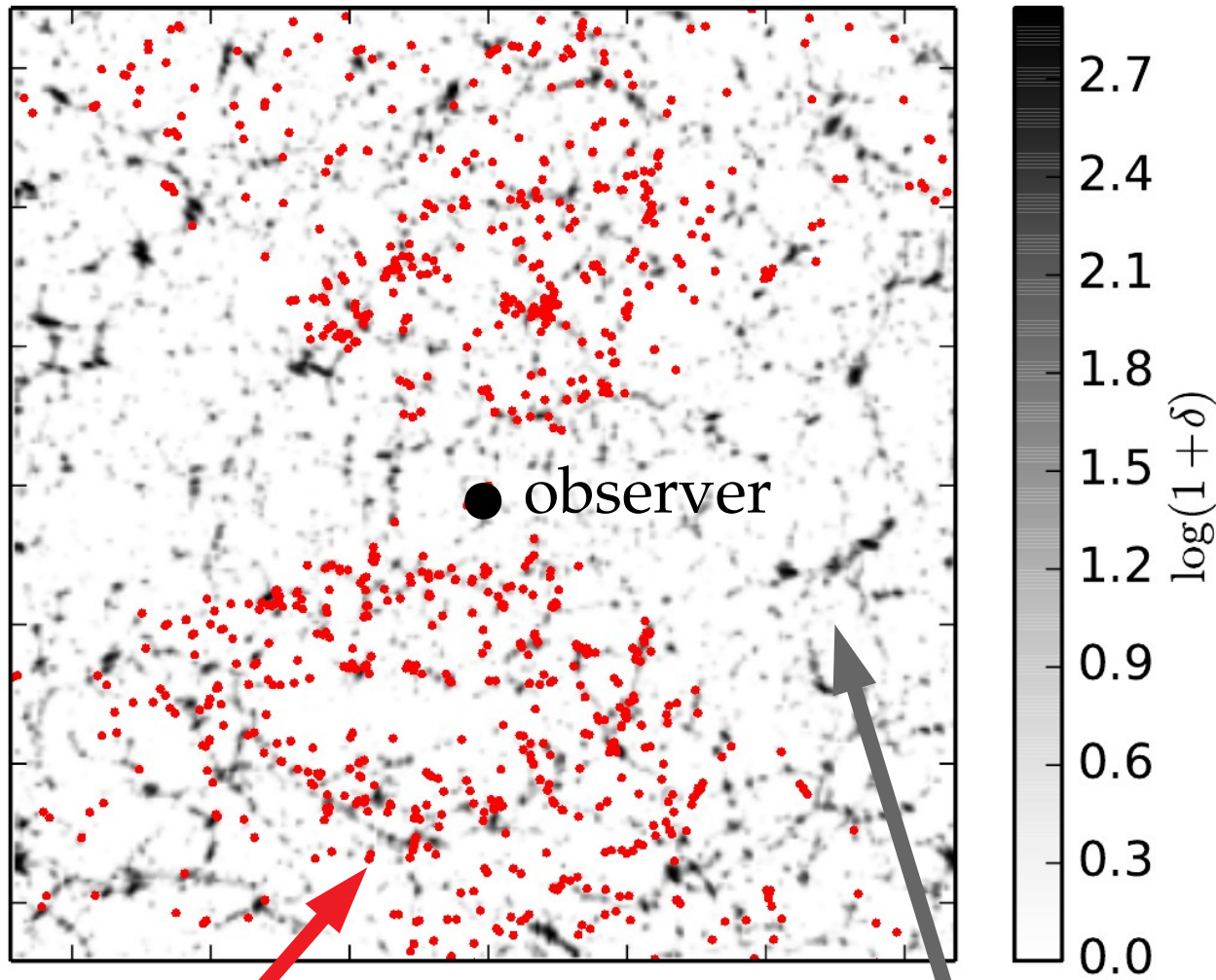
Matter flow



[Porqueres et al. 2019]

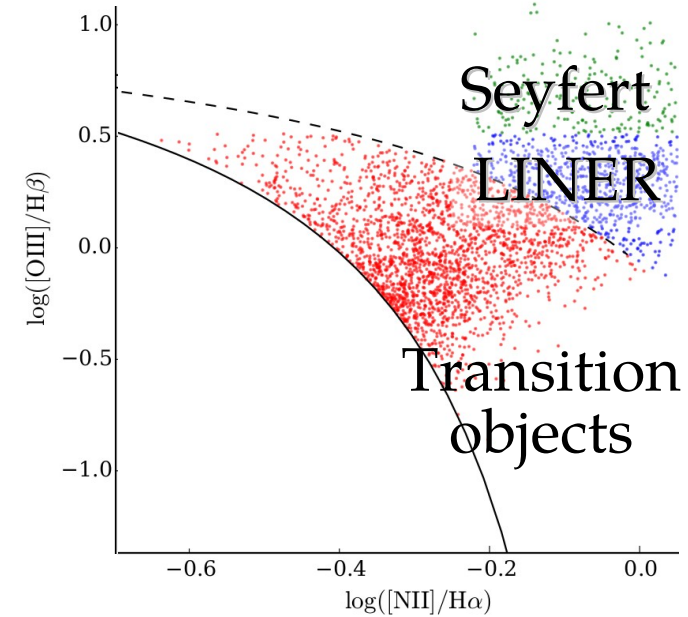
LSS effect on AGN evolution

[Porqueres et al. 2017]



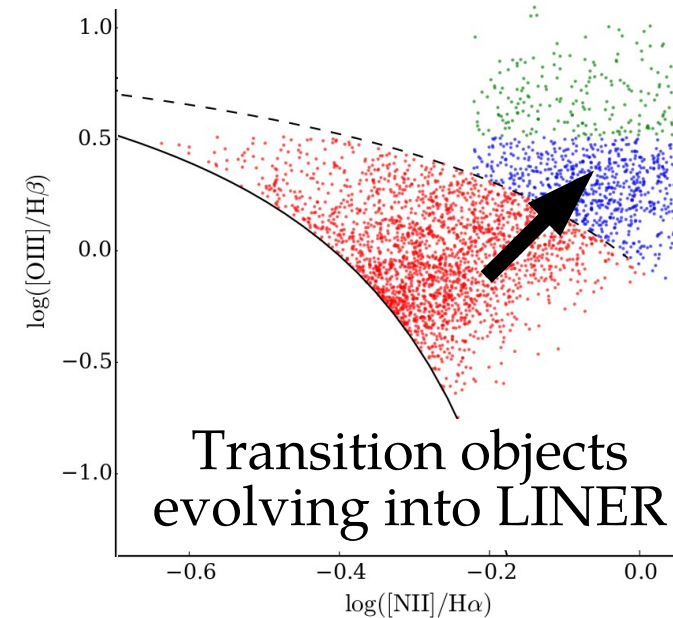
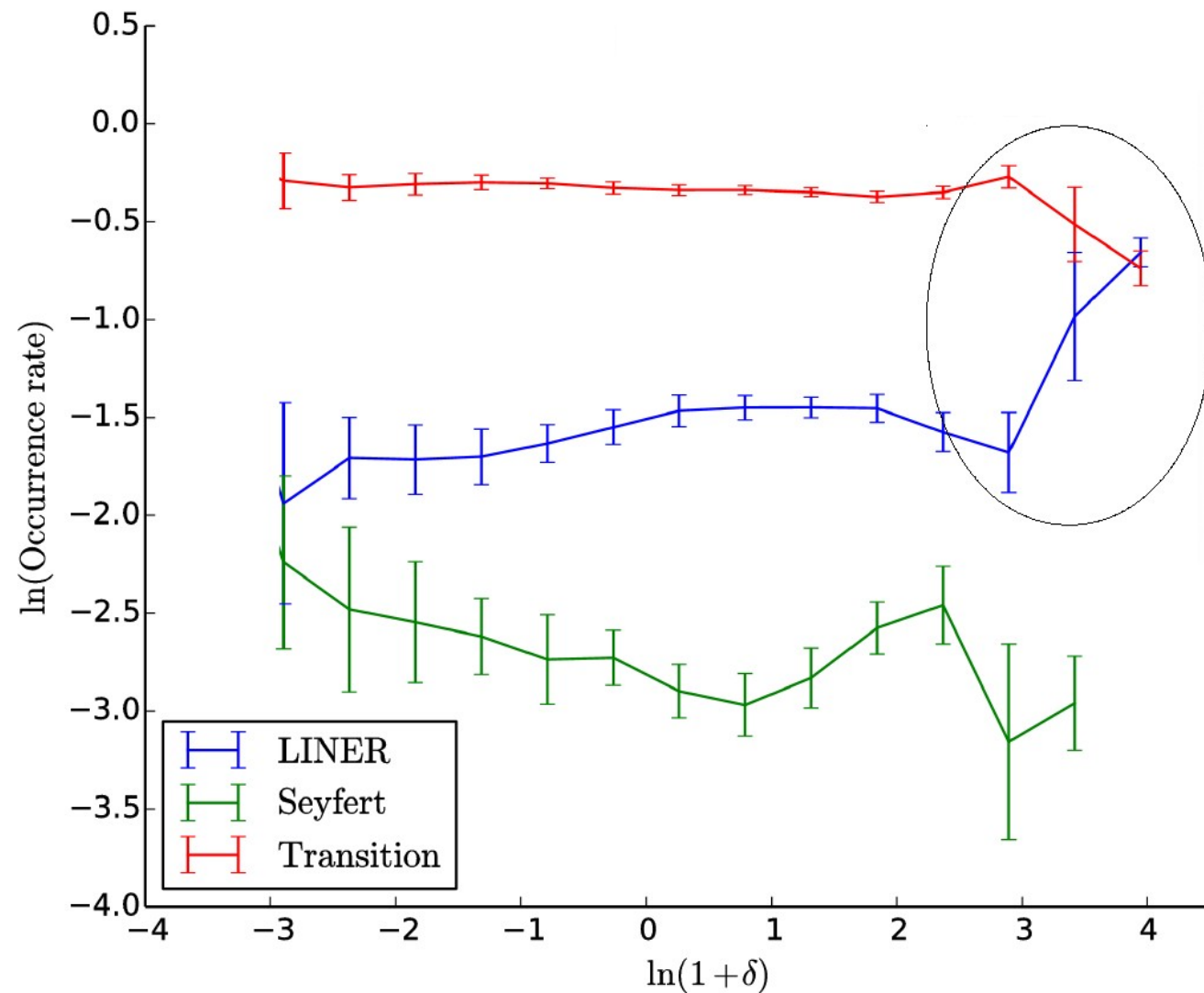
AGN

Density from
galaxy survey
(2M++)

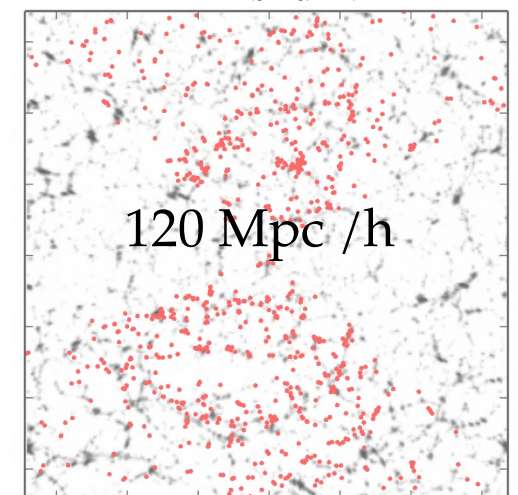


LSS effect on AGN evolution

[Porqueres et al. 2017]

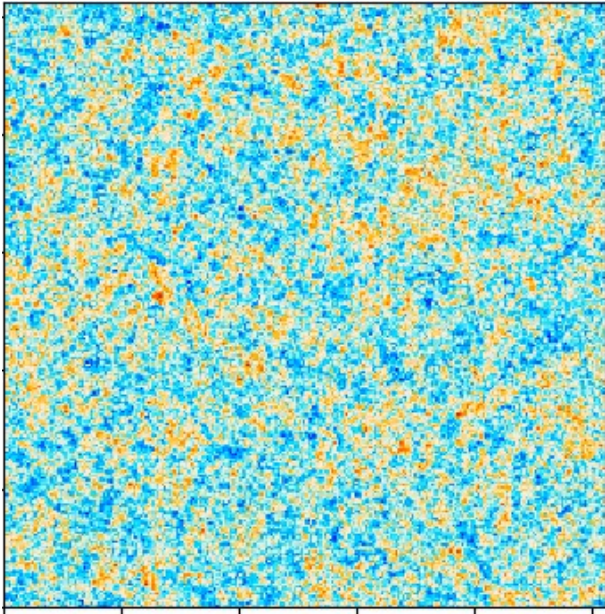


Transition objects
evolving into LINER

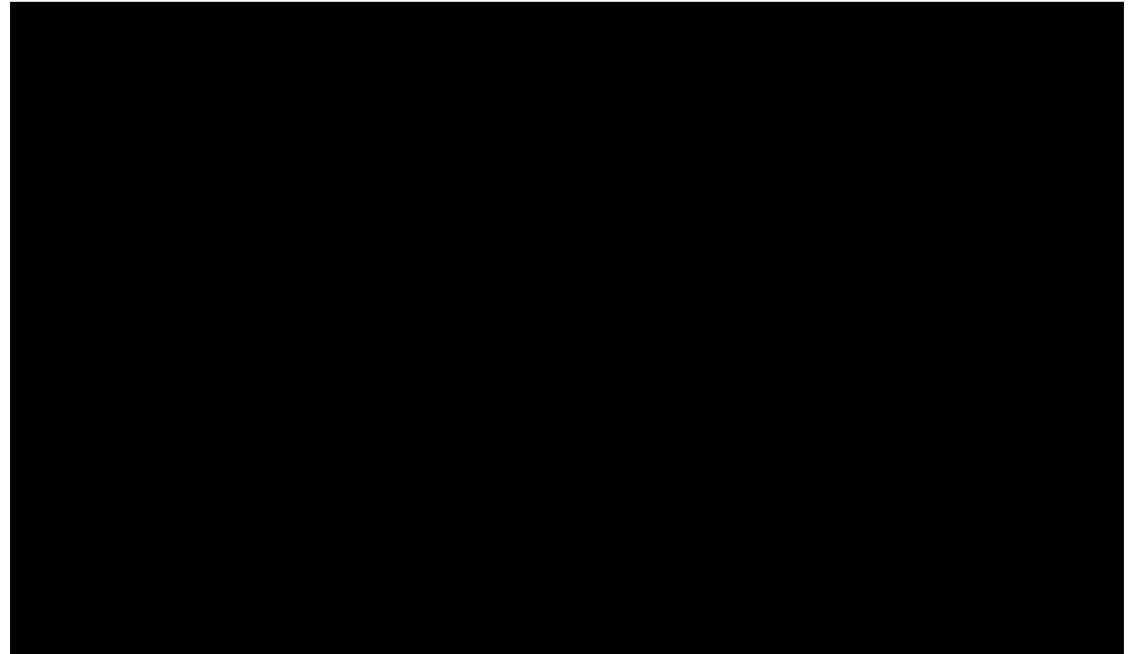


Constrained simulations

BORG initial conditions



Resimulation of Coma cluster (pure DM)



[G. Lavaux, S. Peirani, J. Jasche]

Hydrosimulations to study the physics of gas

Summary

- BORG provides a consistent picture of the cosmic dynamics. It can analyse different types of data: galaxies or Ly- α forest.
- 3D analysis of high- z density field is feasible from a set of 1D lines of sight.
- Physical parameters inference from the Ly- α forest: matter distribution inferred jointly with physical state of neutral hydrogen.
- Matter distribution and velocity fields can be used for
 - Cosmology: cluster and void profiles
 - Test structure and galaxy formation models