

SKA Cosmo Meetings x Snow in Manchester

19th Jan 2023



16th Jan 2024



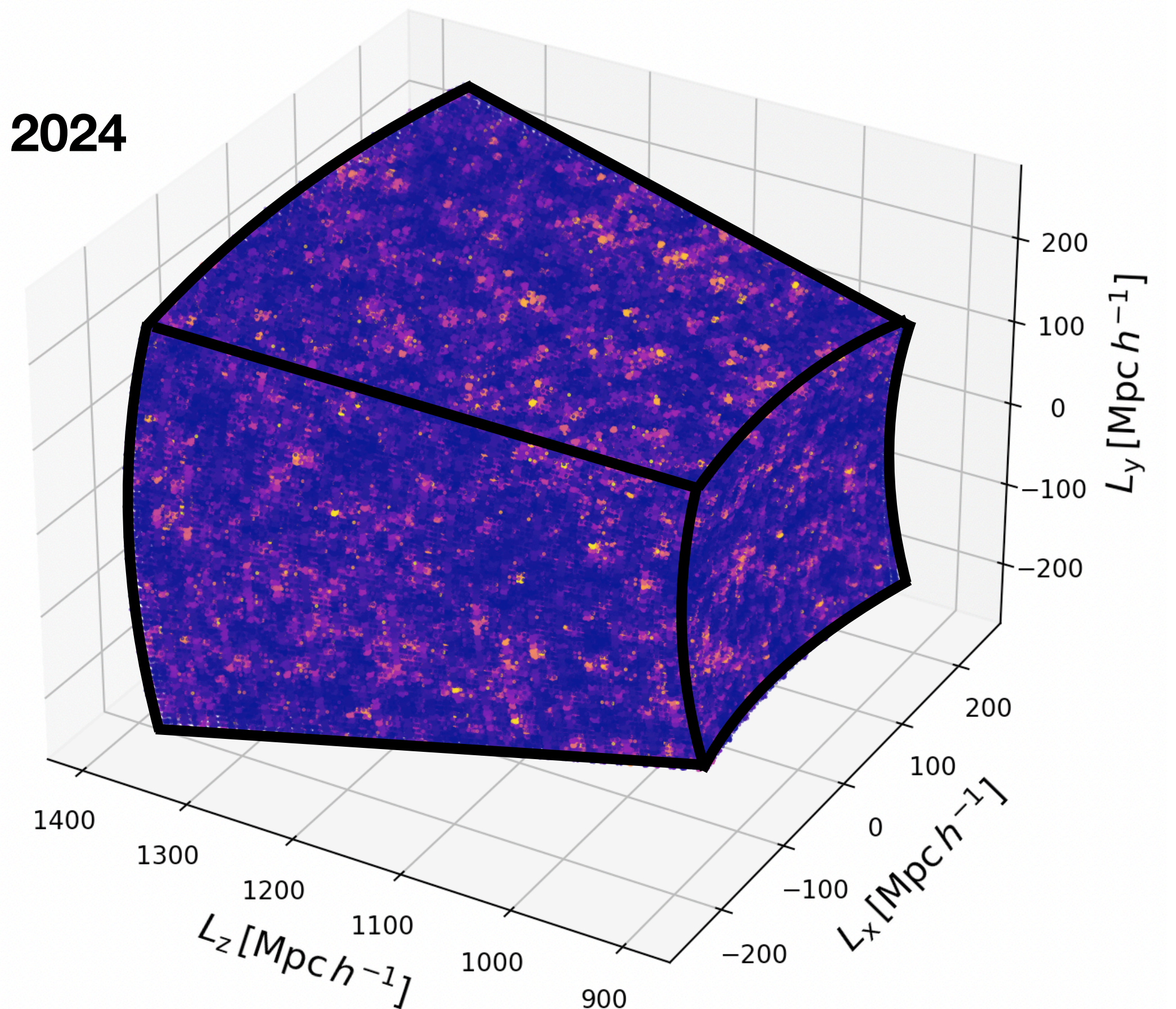
Accurate Fourier-space statistics for line intensity mapping

Steve Cunnington

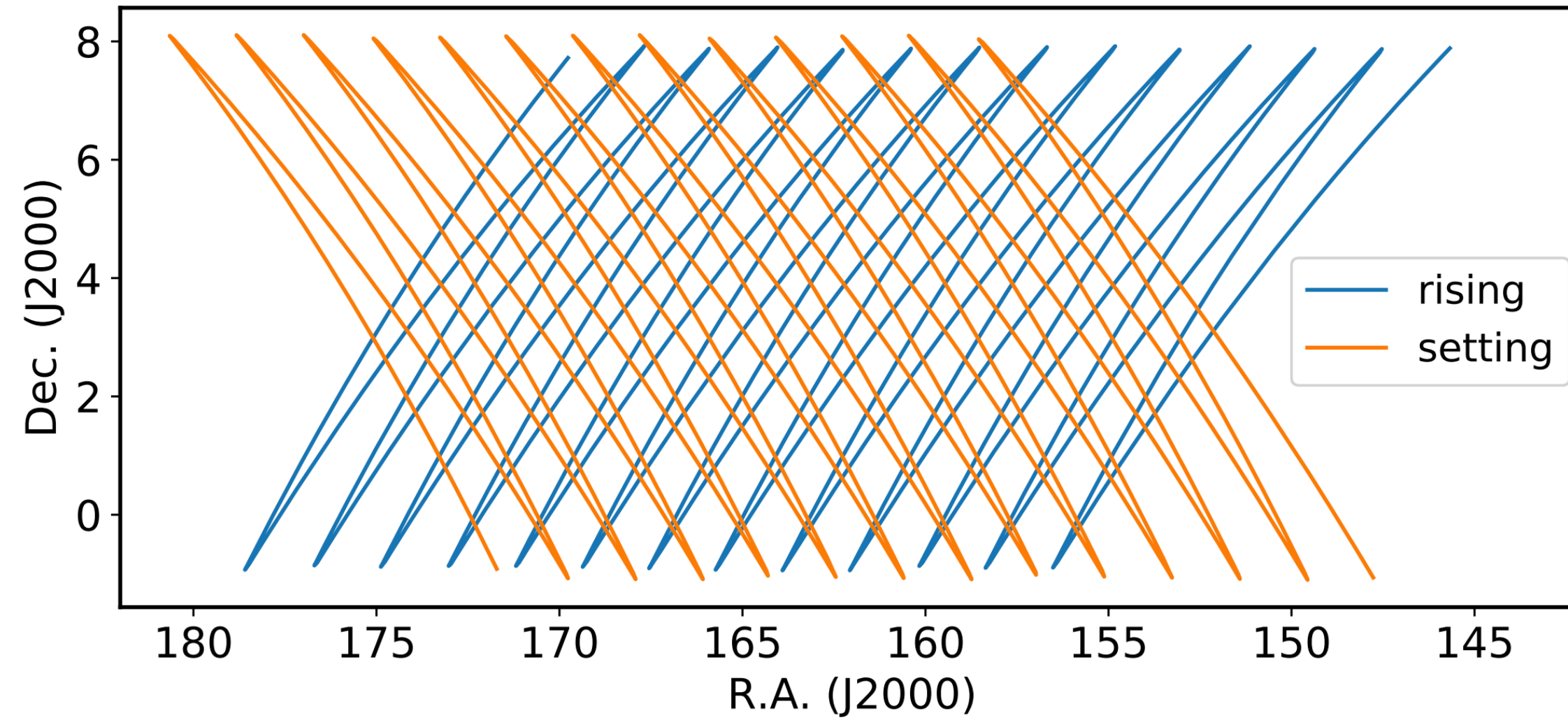
Jodrell Bank Centre for Astrophysics - The University of Manchester

Cosmology SWG Annual Meeting

Centro de Astrofísica da Universidade do Porto - 16th January 2024



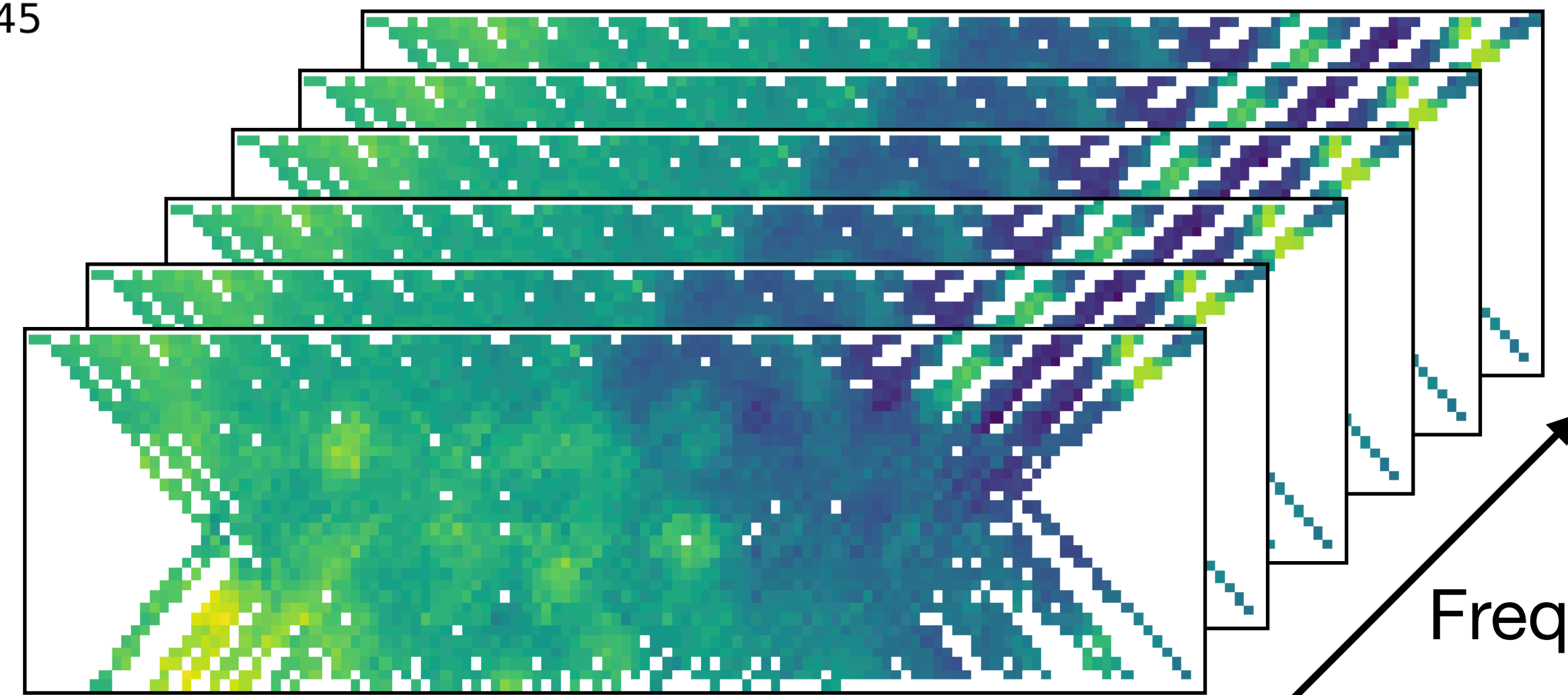
Observations on the sky



MeerKAT 2019 pilot survey scanning strategy
Wang+21 [arXiv:2011.13789]



Declination

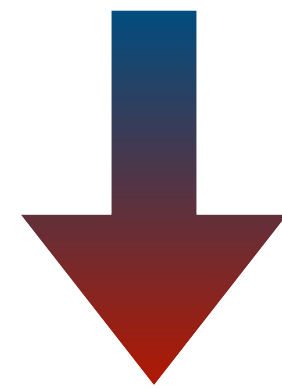


Frequency

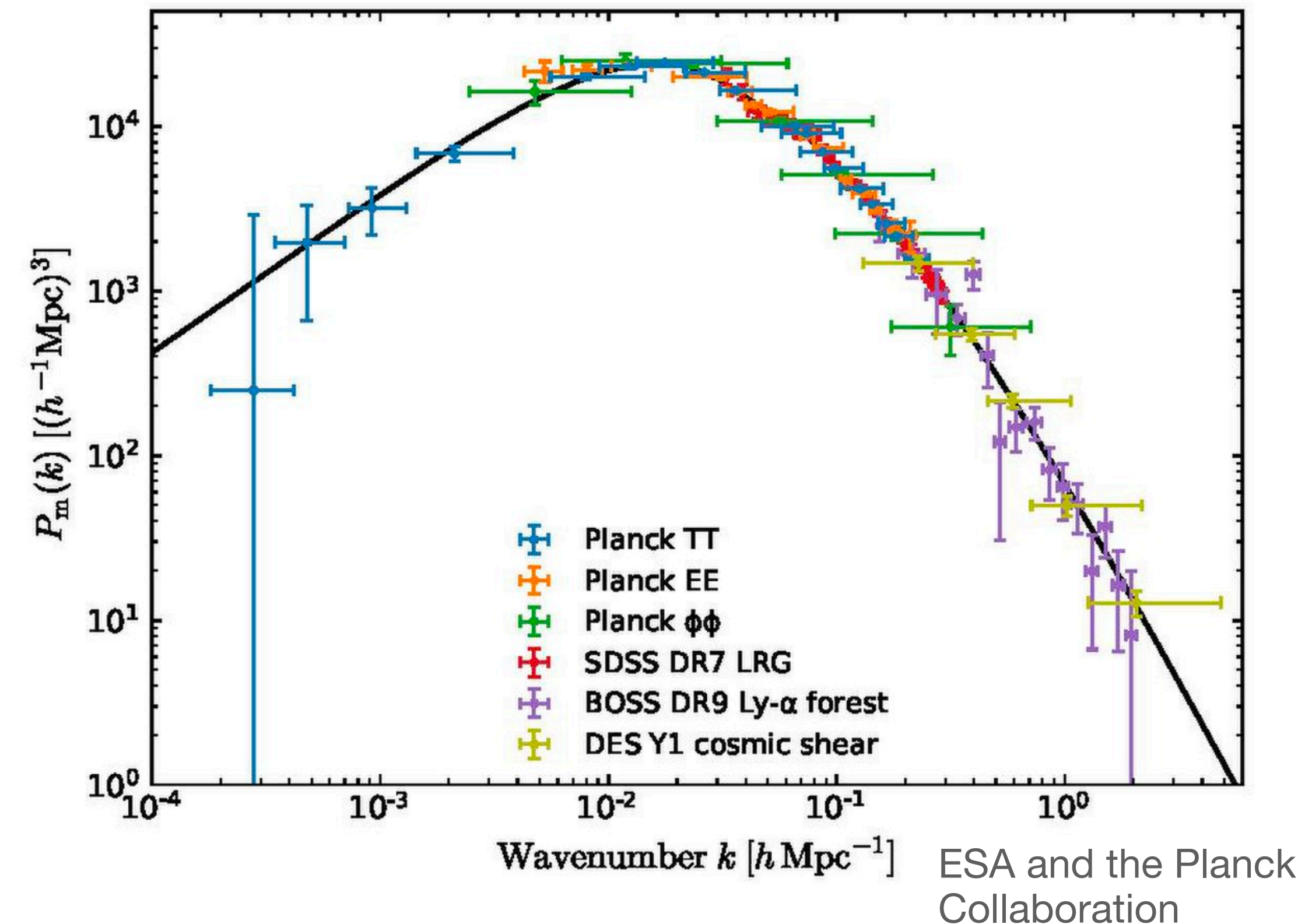
Right ascension



Observations on the sky

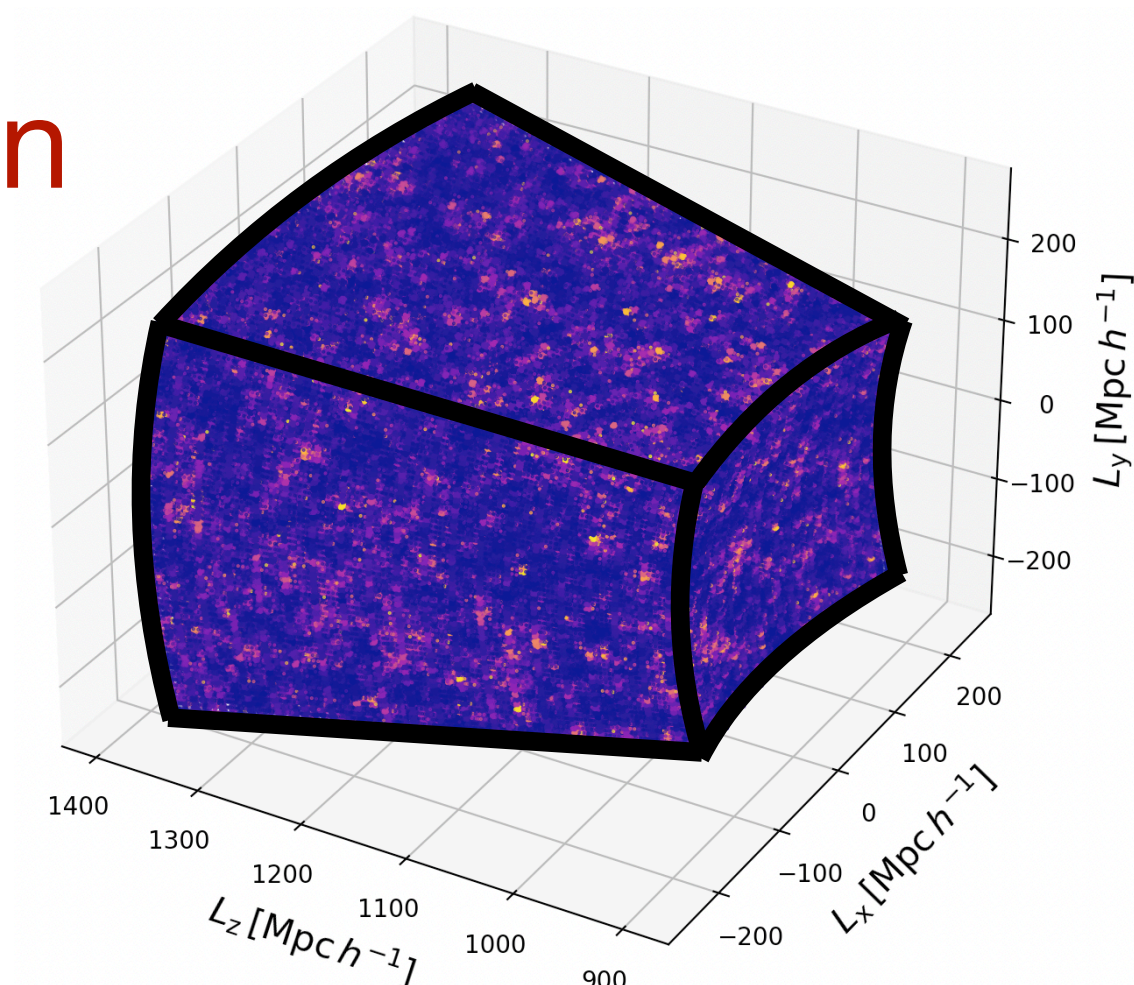


Analysis in Cartesian space

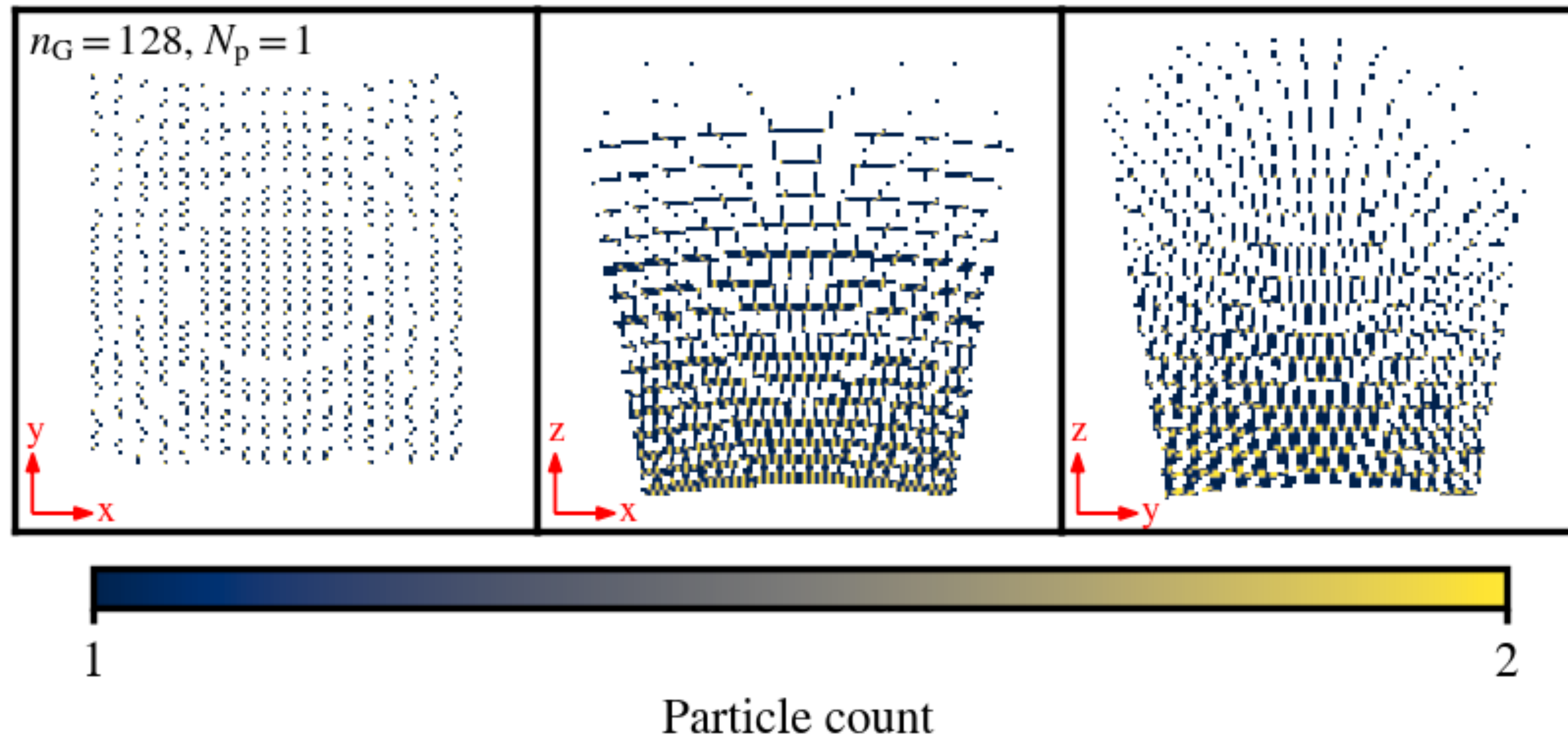


3D **Fourier-space** clustering analysis require observations in **Cartesian comoving** (Mpc/h) space

➔ Require transformation of voxel intensities

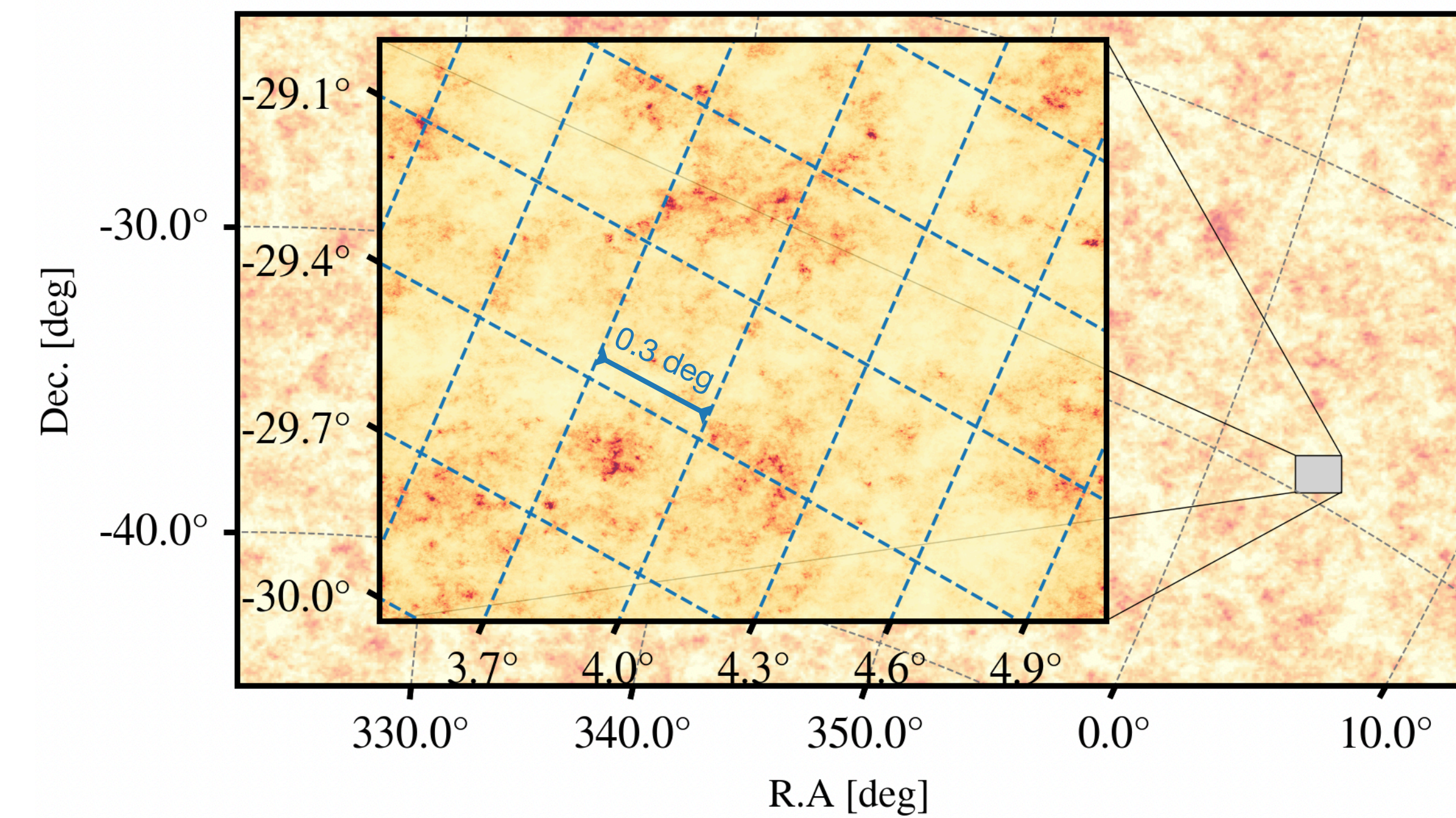


Transforming each voxel's sky coordinate into Cartesian space is trivial but **holes in the coverage** will arise unless very coarse grid is chosen

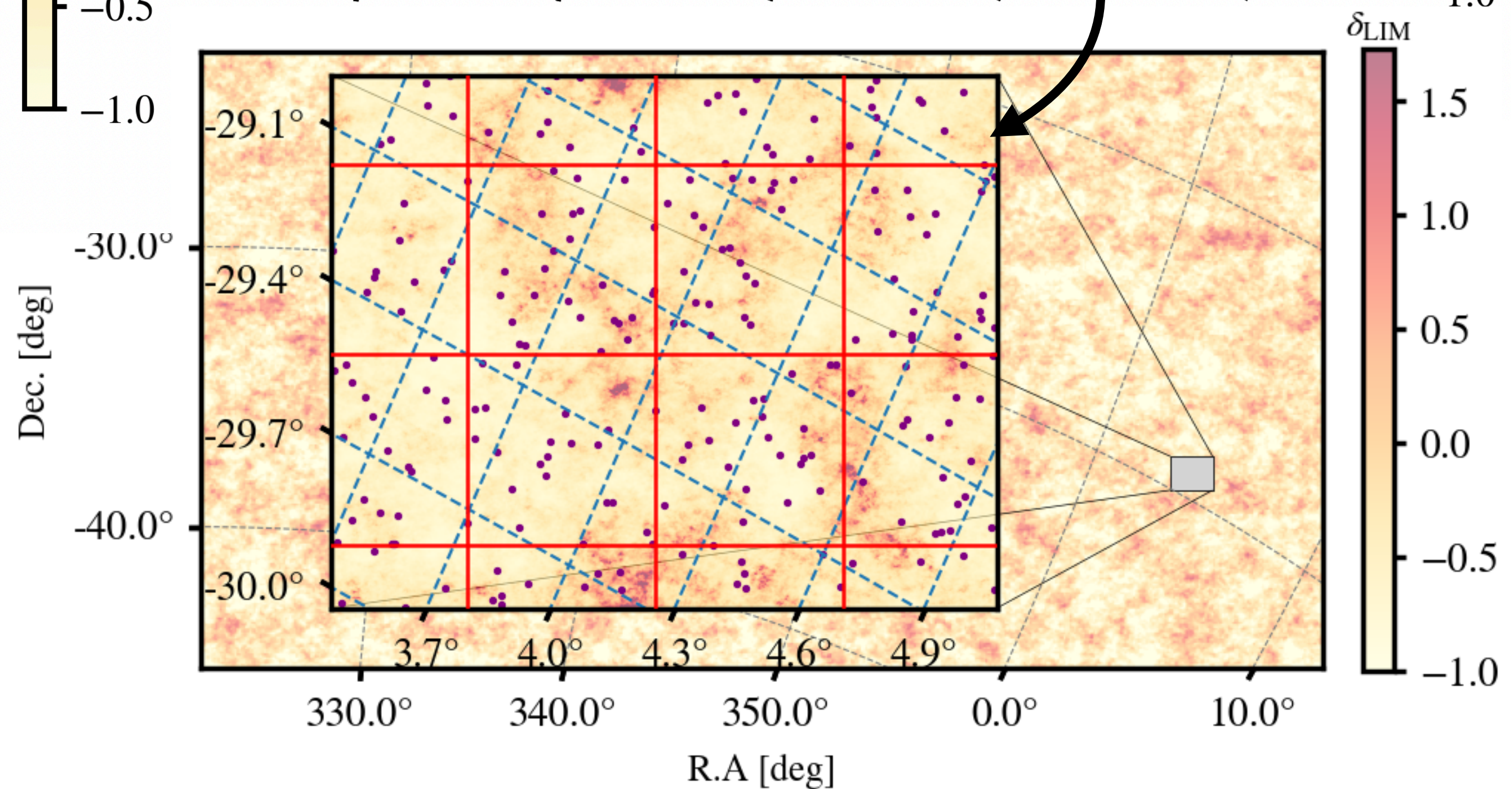
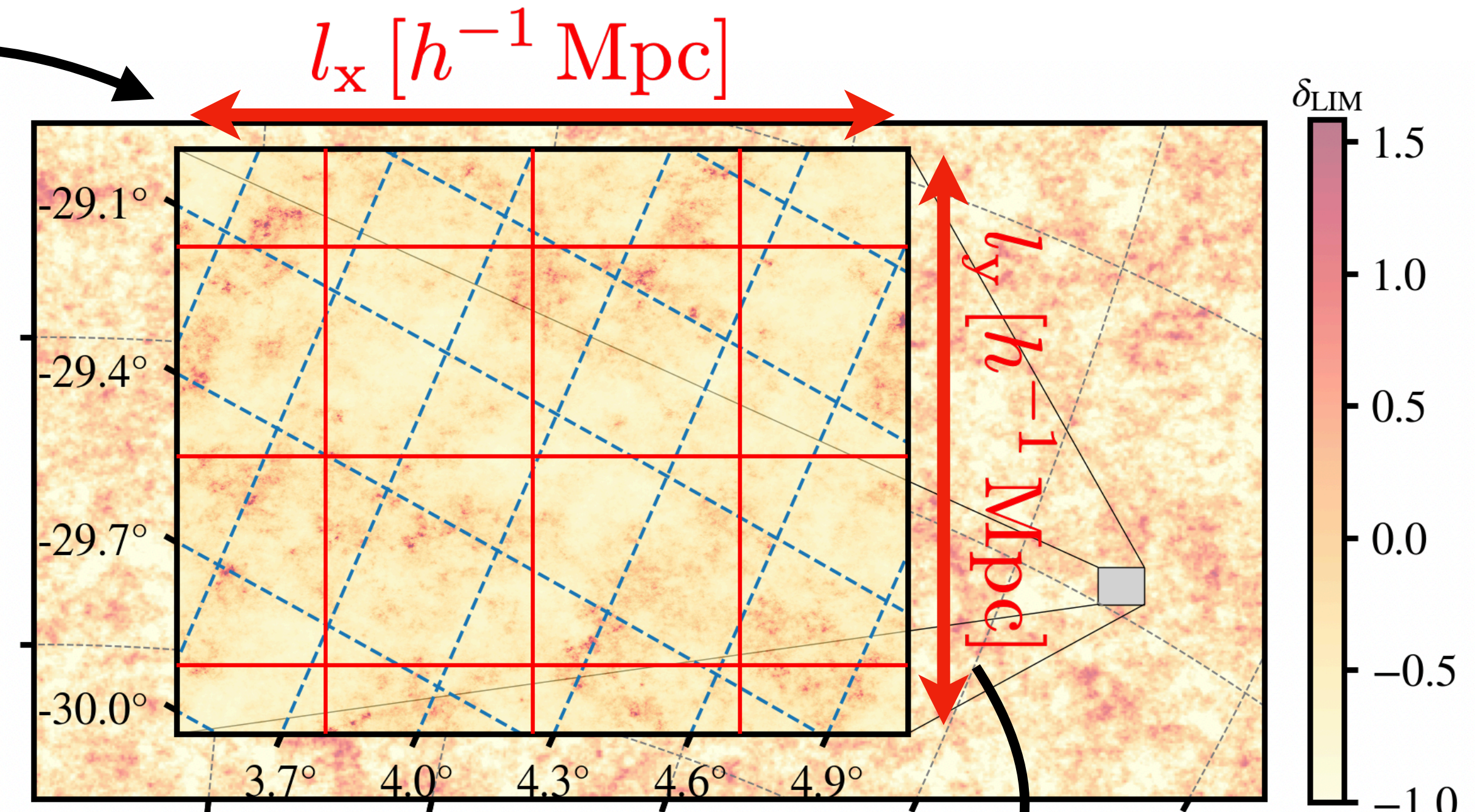
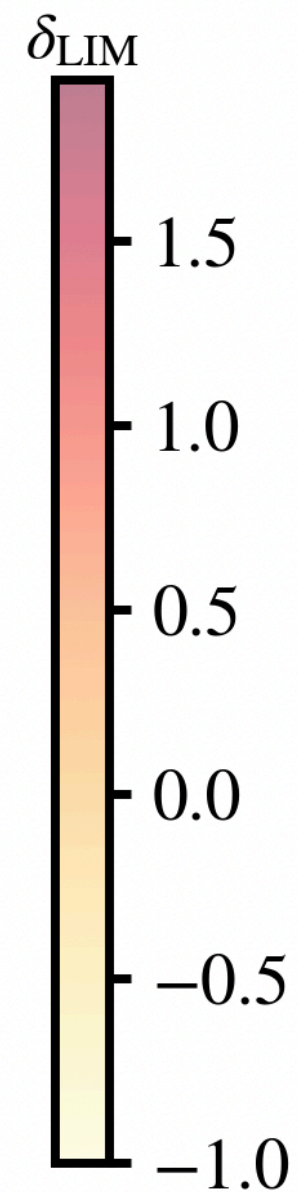


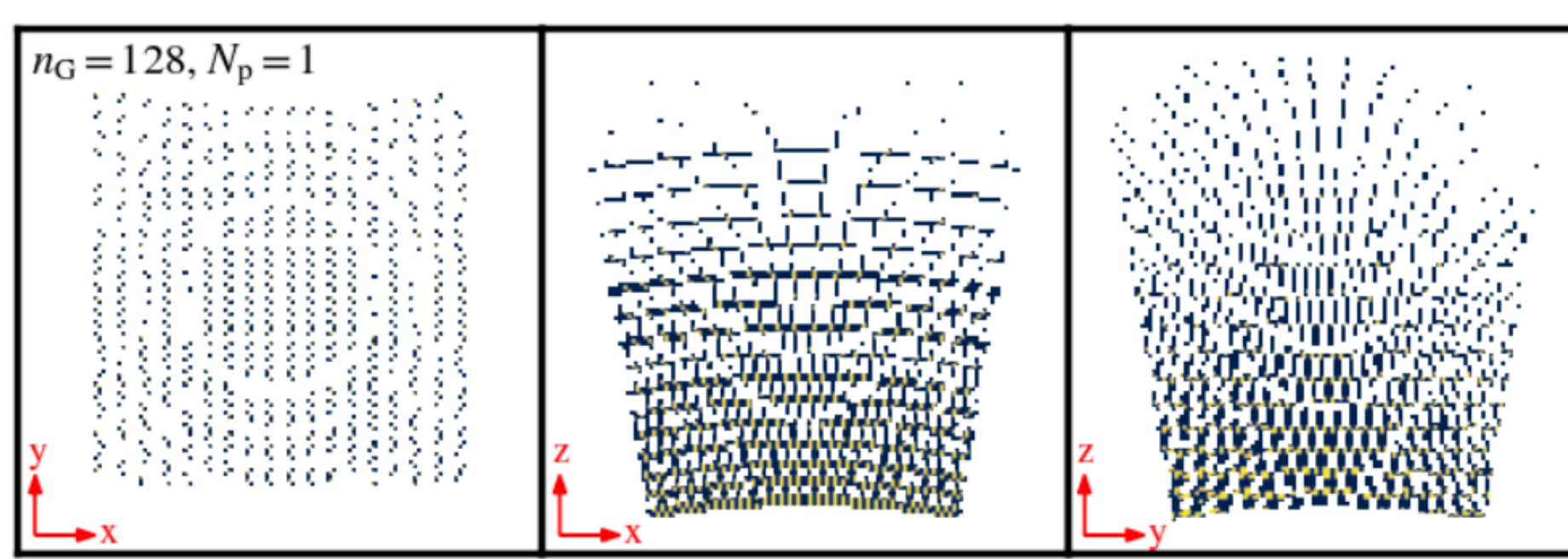
Fields not
suitable for
Fourier
transform

Monte Carlo sampling to a Cartesian grid

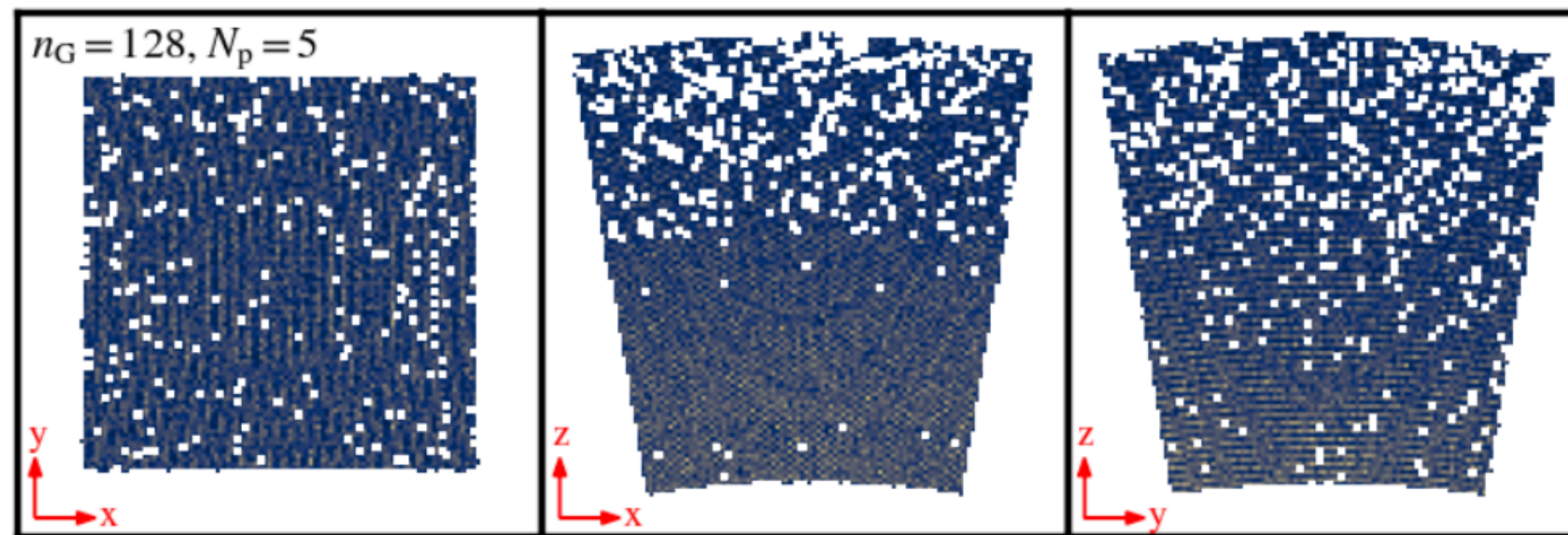


- Sky pixel boundaries
- FFT grid cell boundaries
- Sampling particles

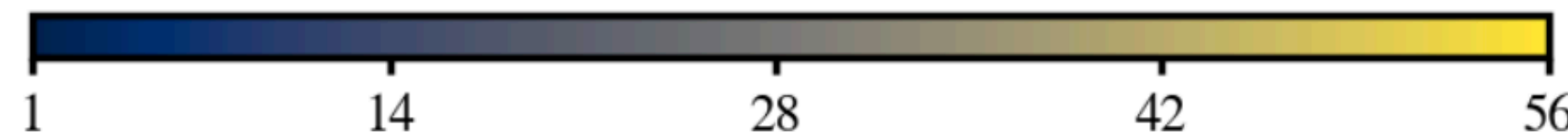
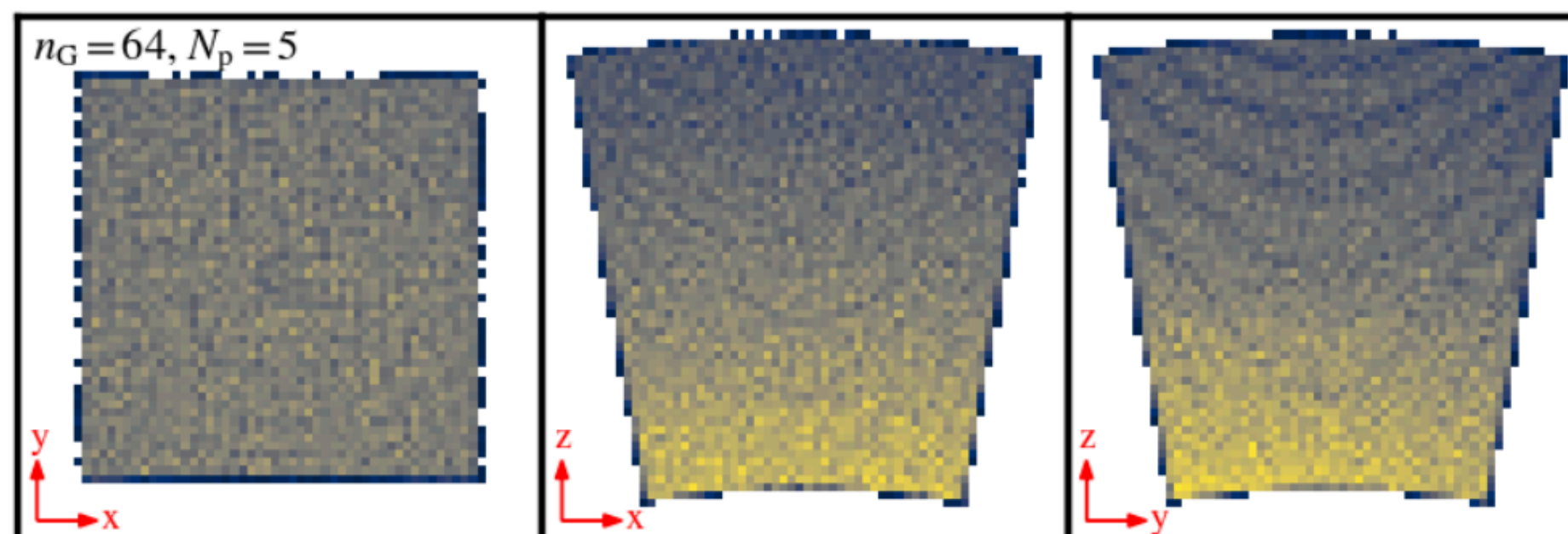




Particle count



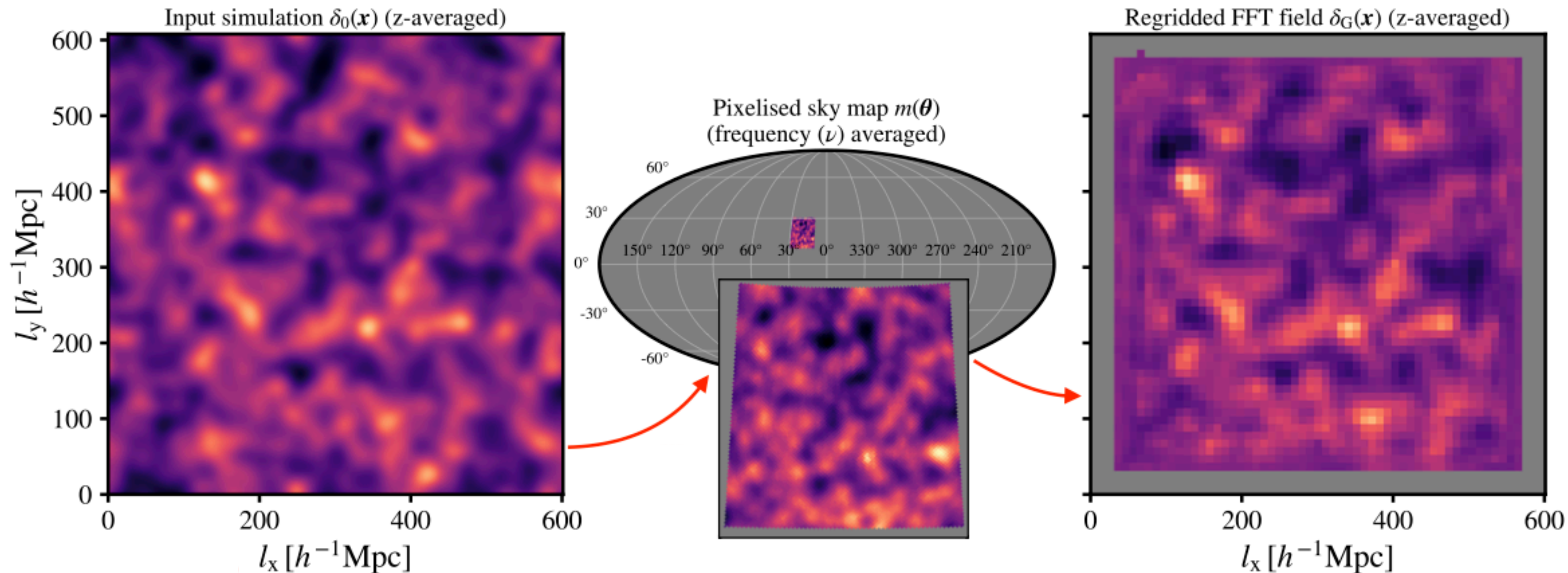
Particle count



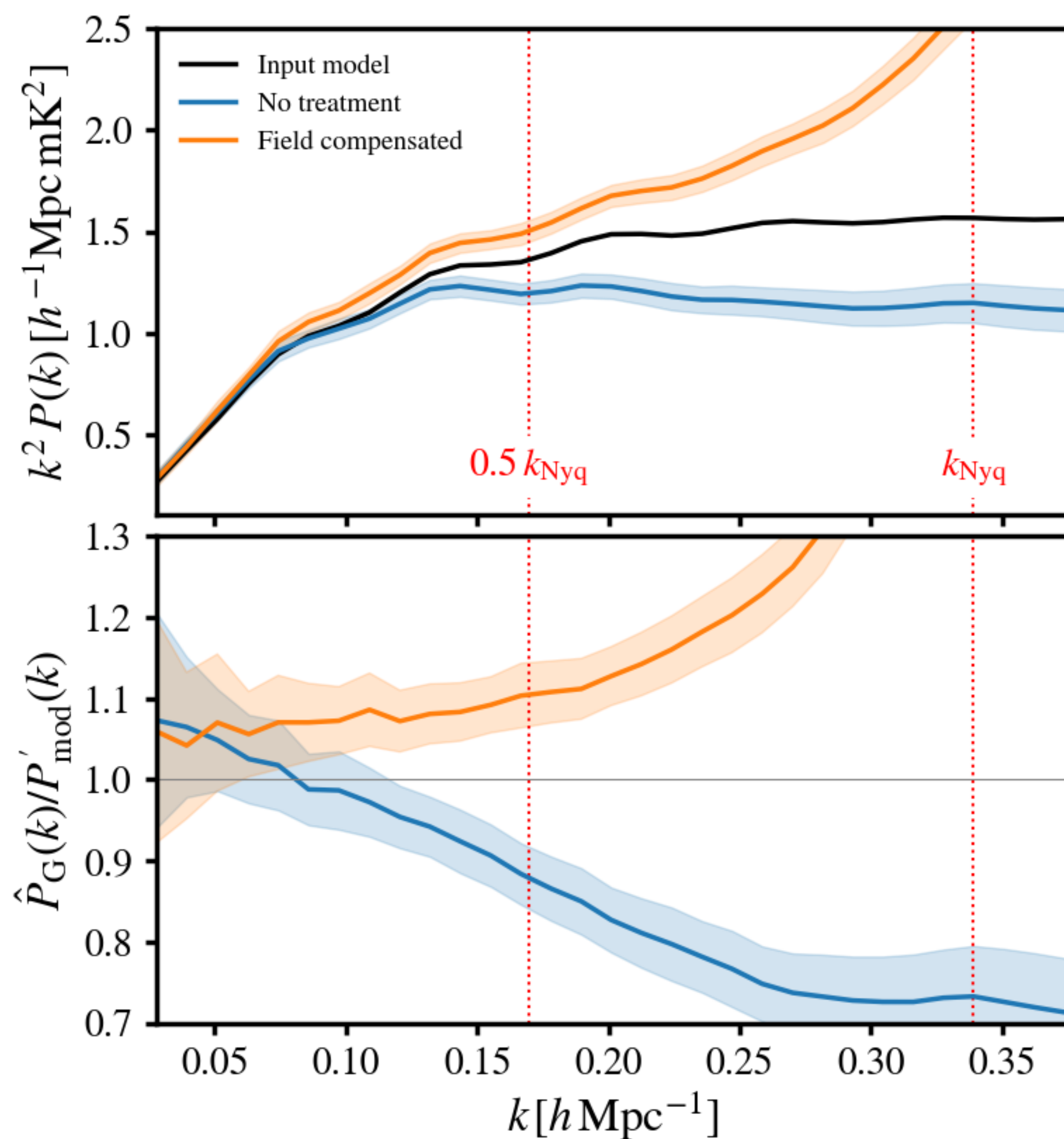
Particle count

- Enough sampling particles
- Well chosen grid resolution
- ➔ footprint in Cartesian space with no holes

Testing with simulations



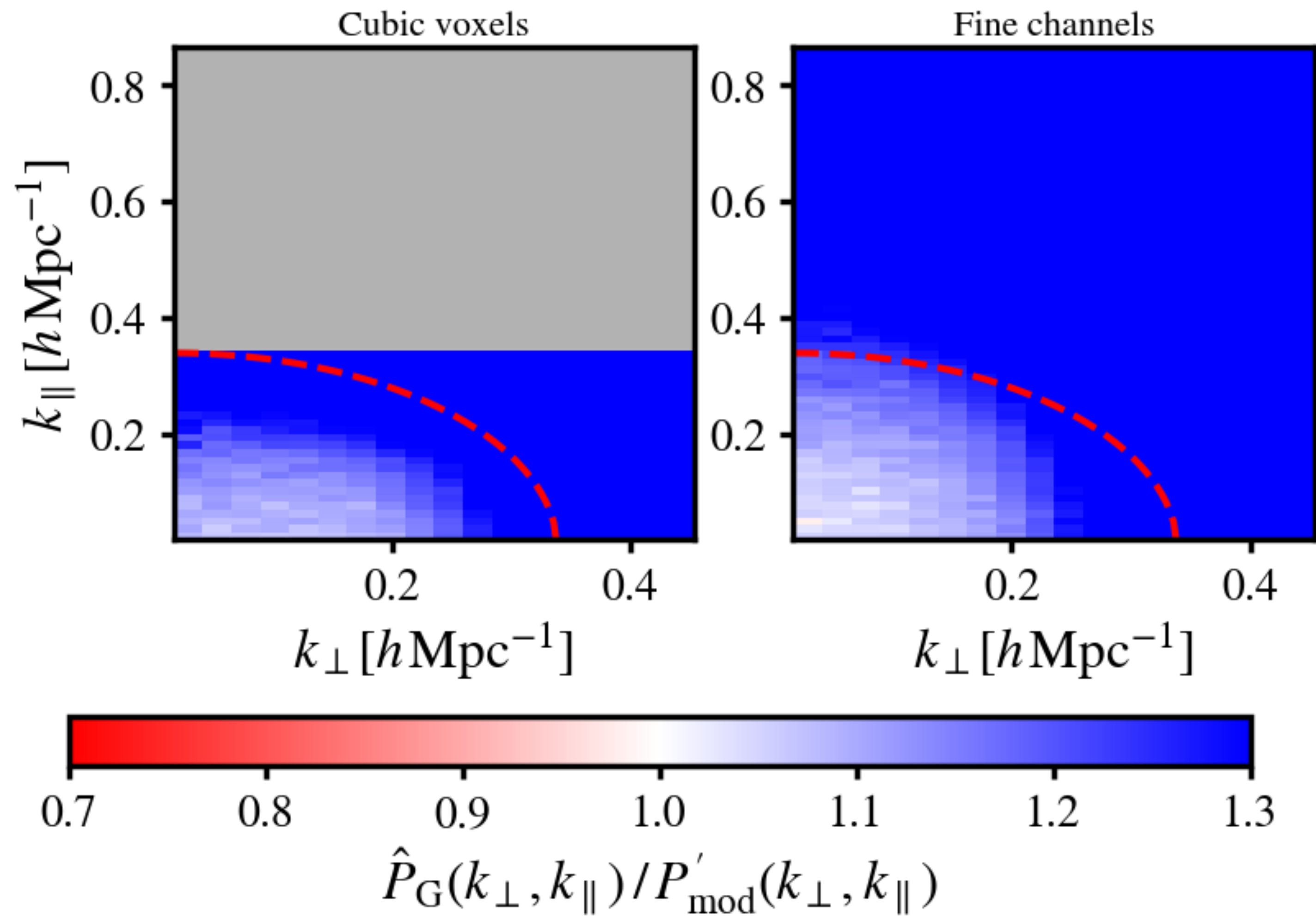
→ <https://github.com/stevecunnington/gridimp>



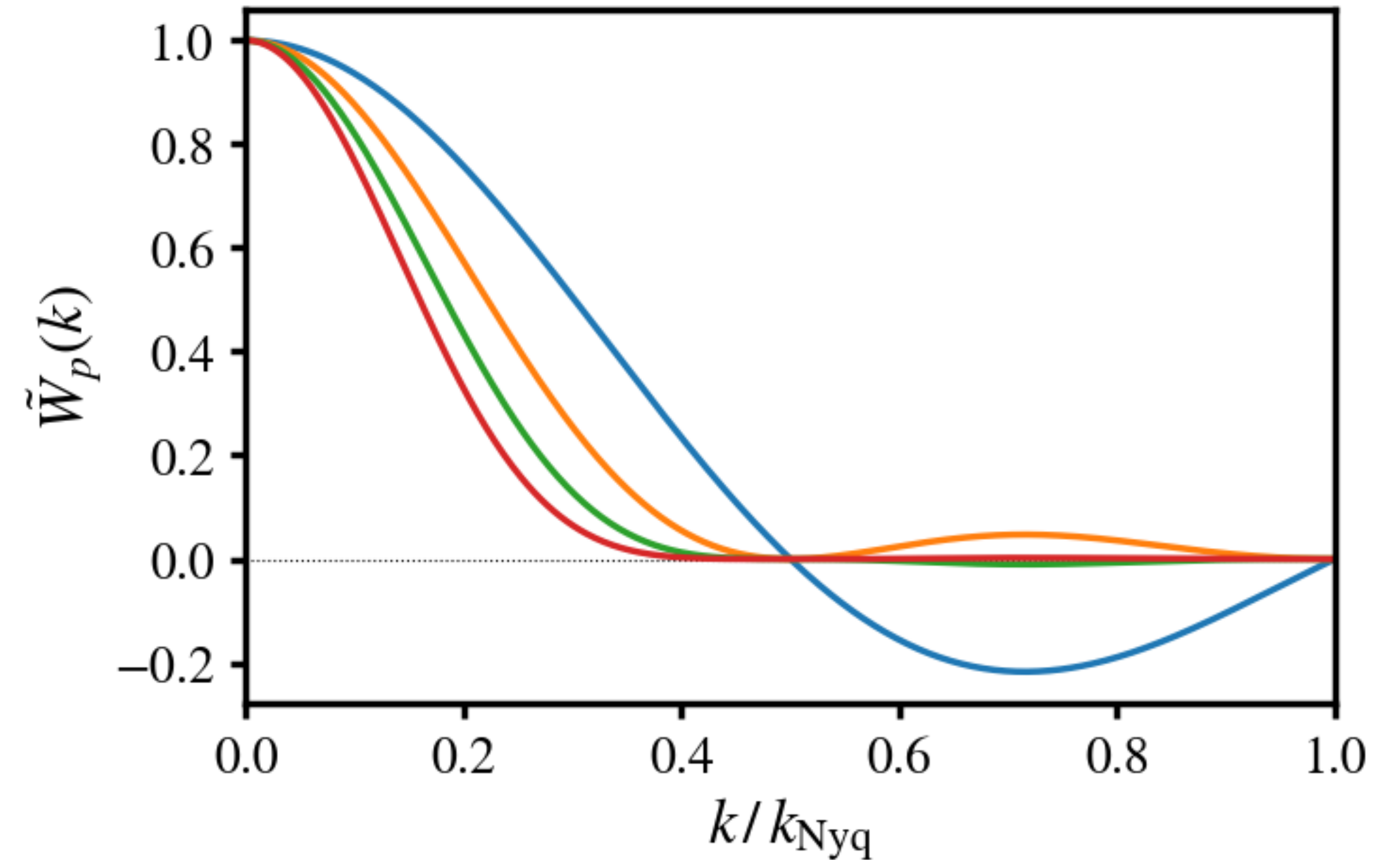
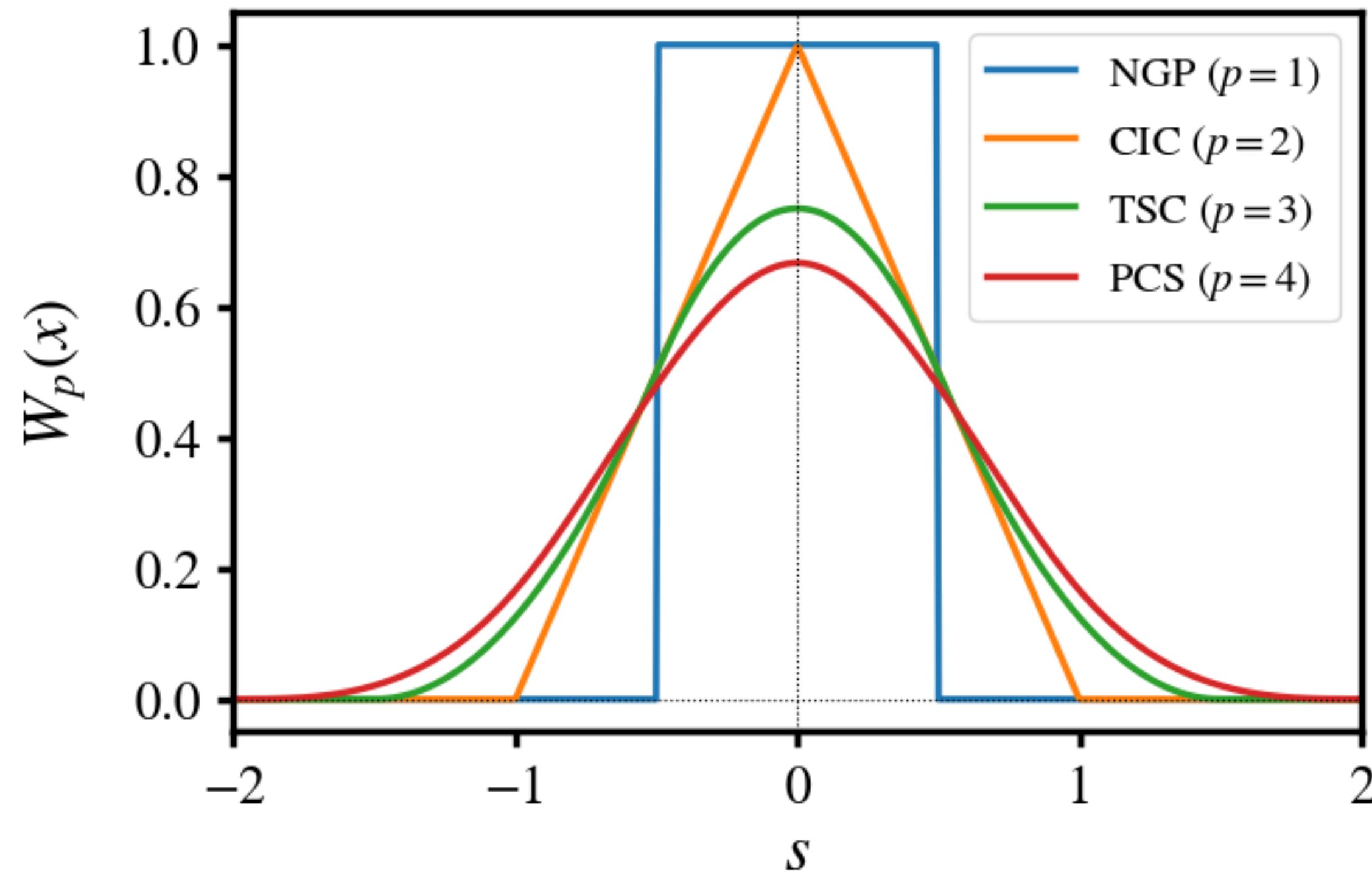
To **compensate** for smoothing from resampling, the Cartesian field is deconvolved with the Fourier transform of **top-hat** functions

$$\tilde{W}_{\text{ngp}}(\mathbf{k}) = \text{sinc}\left(\frac{k_x H_x}{2}\right) \text{sinc}\left(\frac{k_y H_y}{2}\right) \text{sinc}\left(\frac{k_z H_z}{2}\right)$$

High resolution channels

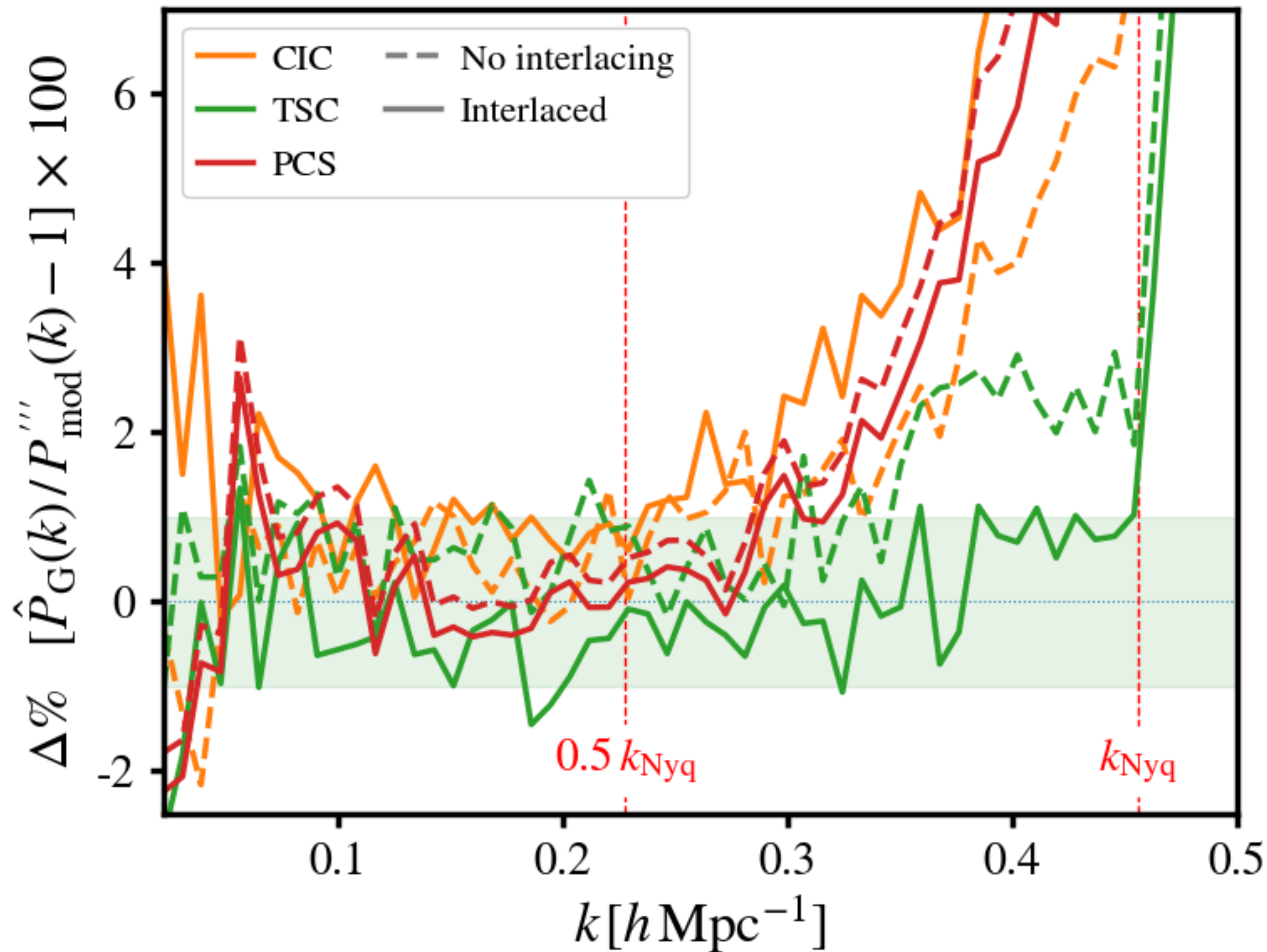


Can we mitigate aliasing with higher-order assignment schemes



$$\tilde{W}_p(\mathbf{k}) = \left[\text{sinc}\left(\frac{k_x H_x}{2}\right) \text{sinc}\left(\frac{k_y H_y}{2}\right) \text{sinc}\left(\frac{k_z H_z}{2}\right) \right]^p$$

Can we mitigate aliasing with higher-order assignment schemes



Also includes some modelling of:

- power damping from HEALPix pixelisation
- power damping from frequency channels

<https://github.com/stevecunnington/gridimp>

stevecunnington / gridimp

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gridimp Public

main 1 Branch 0 Tags

Go to file Add file Code

stevecunnington added arXiv number 93260b9 · last month 26 Commits

| | | |
|-----------------|---------------------------------|--------------|
| data | param file update | 2 months ago |
| examples | axes labels | 2 months ago |
| gridimp | removed CAMB option | last month |
| scripts | fixed some path inconsistencies | last month |
| README.md | added arXiv number | last month |
| environment.yml | .yml file | 2 months ago |
| setup.py | fixing setup.py steps | 2 months ago |

README

gridimp

gridded intensity mapping power

Getting started:

Cartesian lognormal mock:

Generate a fast cubic lognormal mock over $n=128^3$ cells with size $l=500^3(\text{Mpc}/h)^3$:

```
from gridimp import cosmo
from gridimp import mock

cosmo.SetCosmology(z=1) # set cosmology
Pmod = cosmo.GetModelPk(z=1) # obtain matter power spectrum

l = 500 # length of box down one side [Mpc/h]
n = 128 # number of cells down one side
dims = [l,l,l,n,n,n,0,0,0] # lengths,cells,origins for each dimension

delta_0 = mock.Generate(Pmod,dims) # generate LIM mock (default b=1,T=1)
```

HEALPix sky maps for simulated survey:

By defining input simulated survey parameters, a lognormal mock can be generated onto a Cartesian grid calculated to fully enclose the survey footprint.

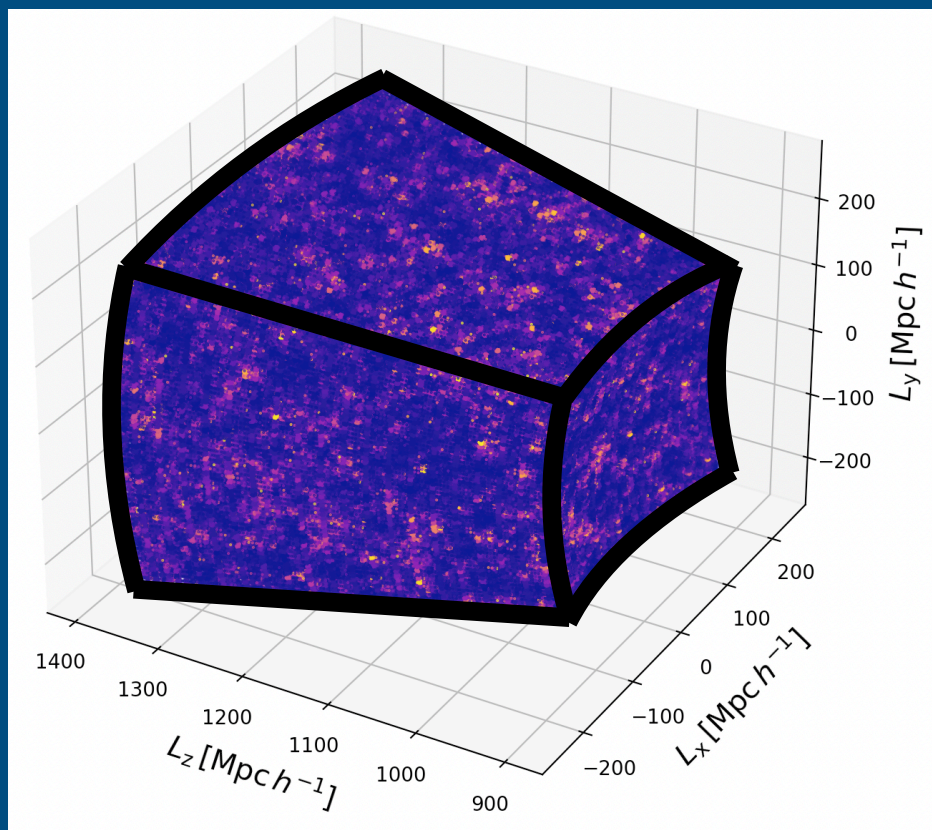
```
### Define survey parameters:
ramin,ramax = 10,30 # [deg]
decmin,decmax = 10,30 # [deg]
numin,numax = 900,1110 # [MHz]
nnu = 60 # number of frequency channels
nside = 128 # HEALPix resolution: determines pixel size
n0 = 256 # n0^3 will be number of cells for input grid cube

### Initialise healpix and get sky coordinates for voxels covering survey:
from gridimp import grid
ra,dec,nu,dims_0 = grid.init_healpix(nside,ramin,ramax,decmin,decmax,numin,numax,nnu,n0)

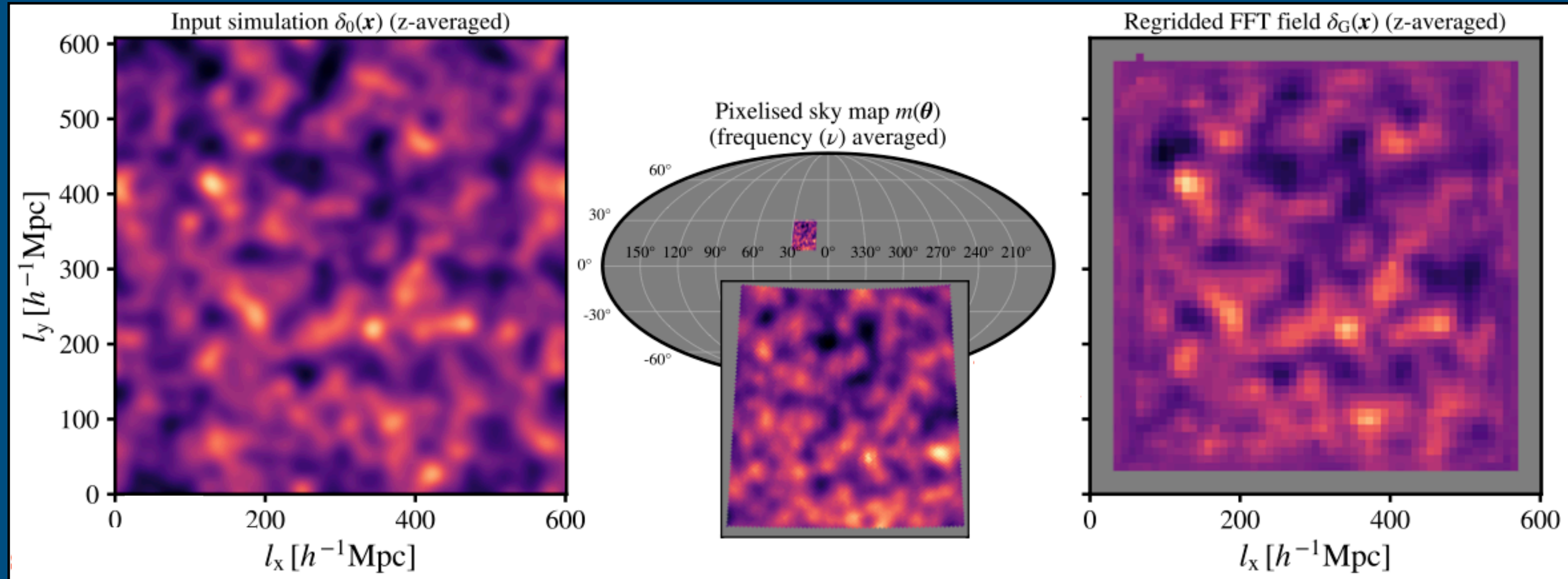
delta_0 = mock.Generate(Pmod,dims_0) # generate LIM mock (default b=1,T=1)

# Create healpy sky maps for each frequency channel:
m = grid.lightcone_healpy(delta_0,dims_0,ra,dec,nu,nside)
```

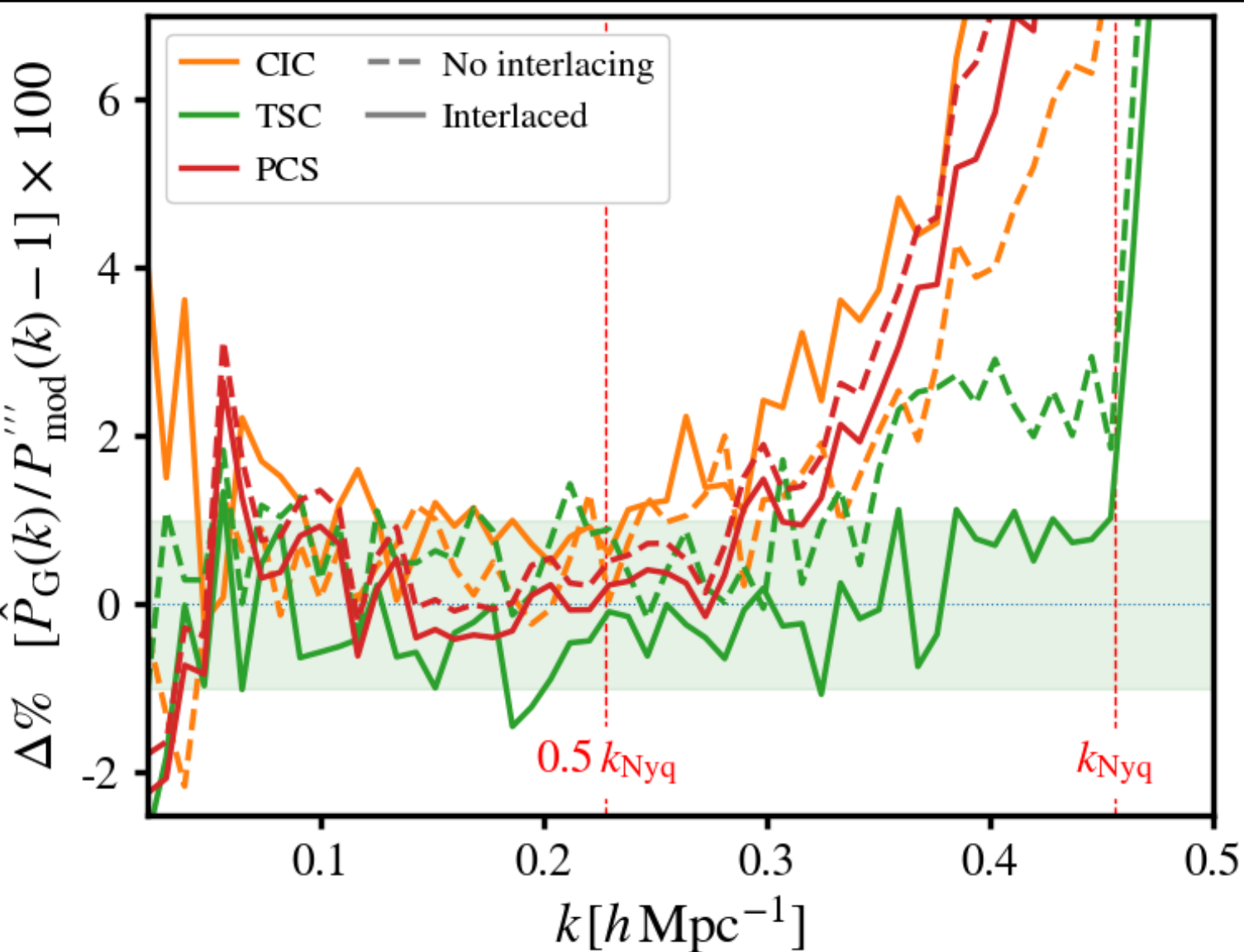
In Summary...



Need to robustly validate methods for transforming intensity maps to Cartesian spaces for clustering analysis



Using a suite of lognormal simulations we tested a Monte Carlo field sampling technique as a means for regridding the observational data

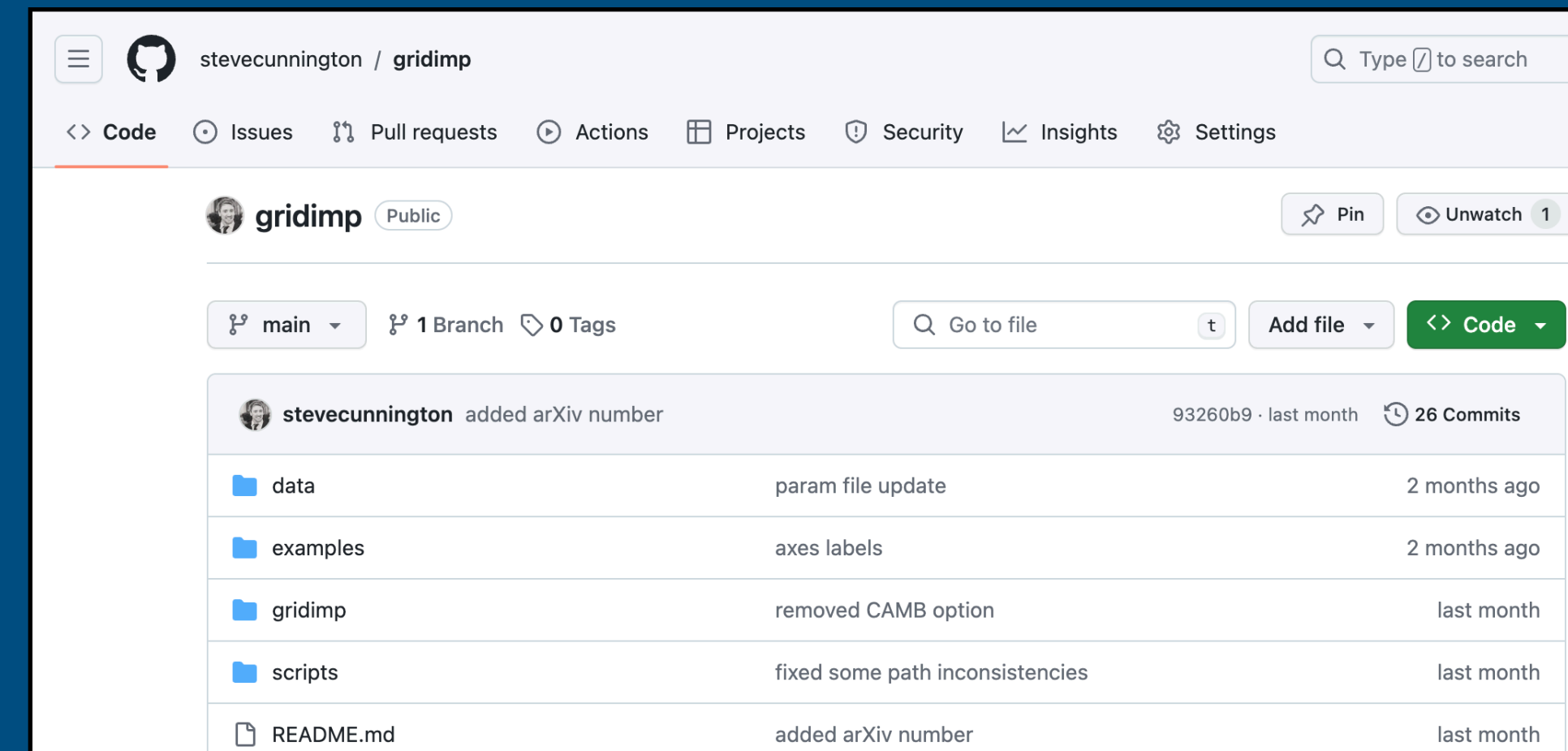


With additional modelling, good accuracy is achieved. Higher-order particle assignment can also help control aliasing

Thank you

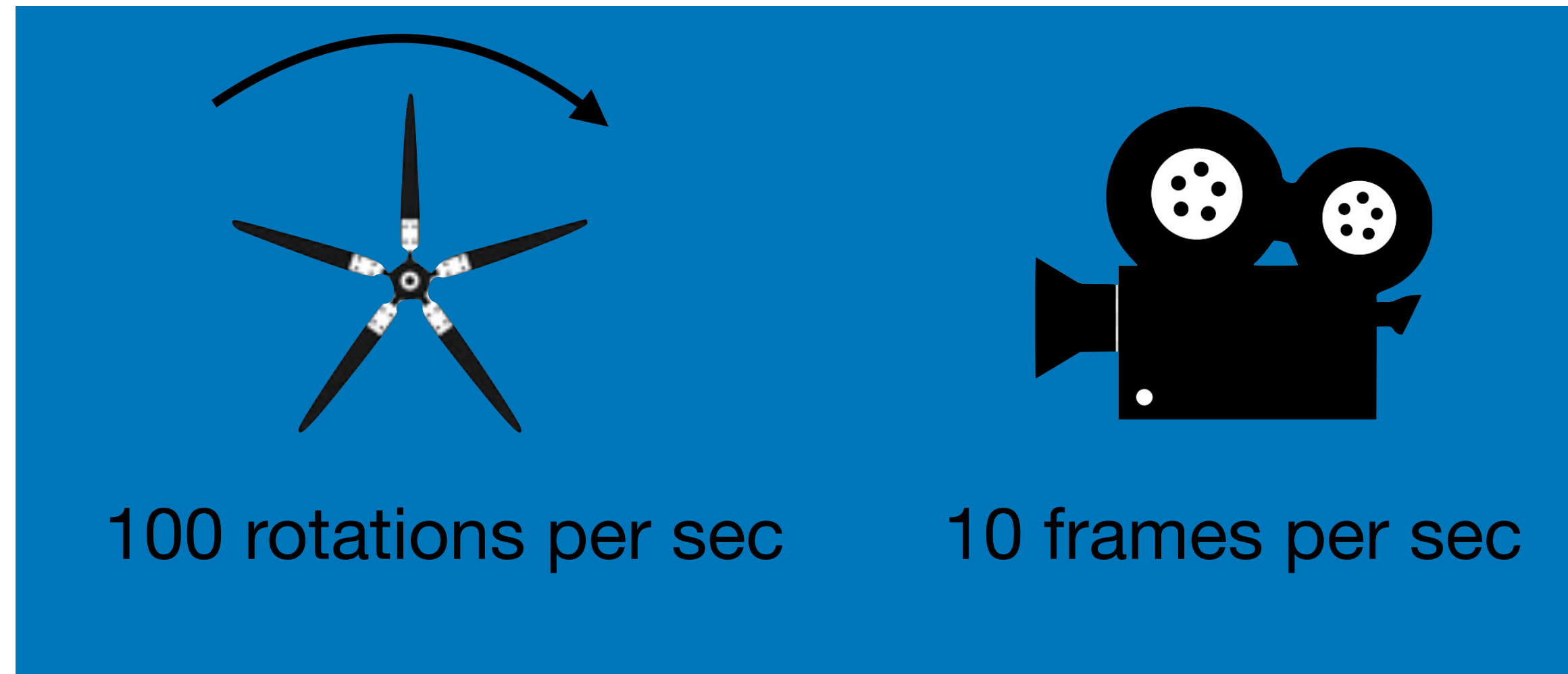
Public code available to generate intensity maps with survey geometry and correct for the distorting effects

github.com/stevecunnington/gridimp

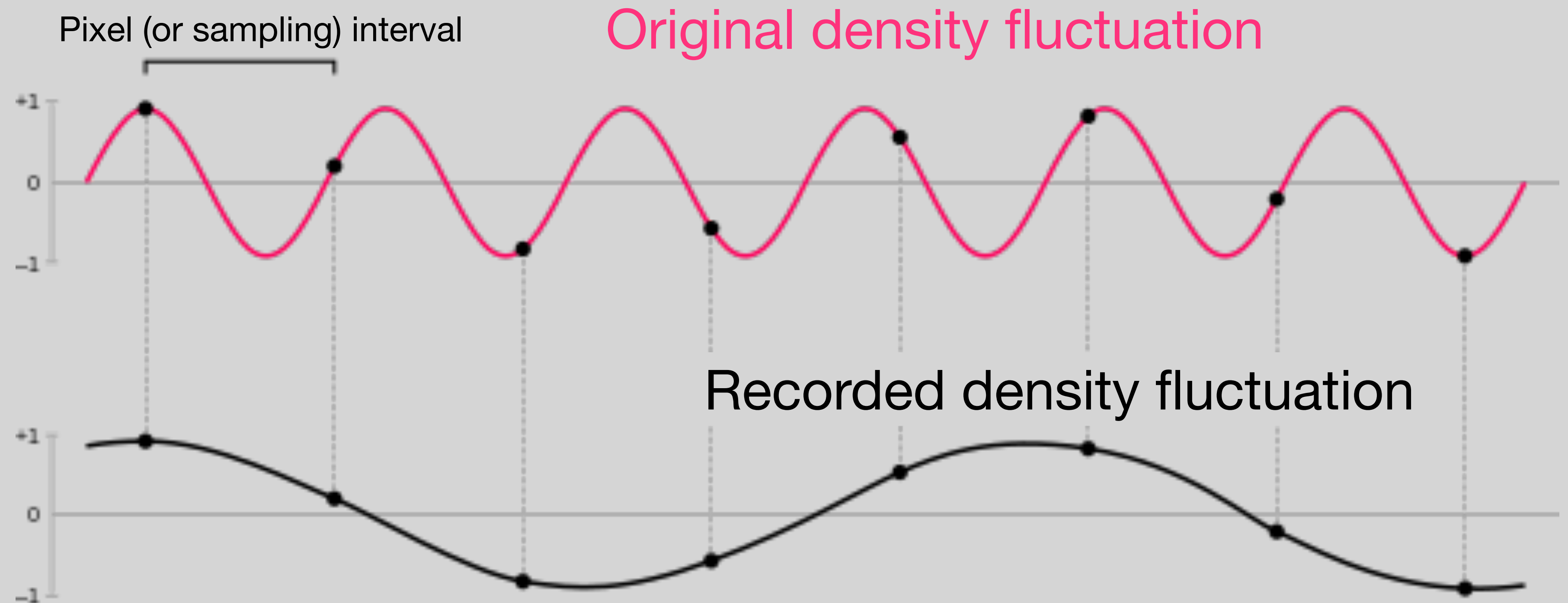


Backup Slides

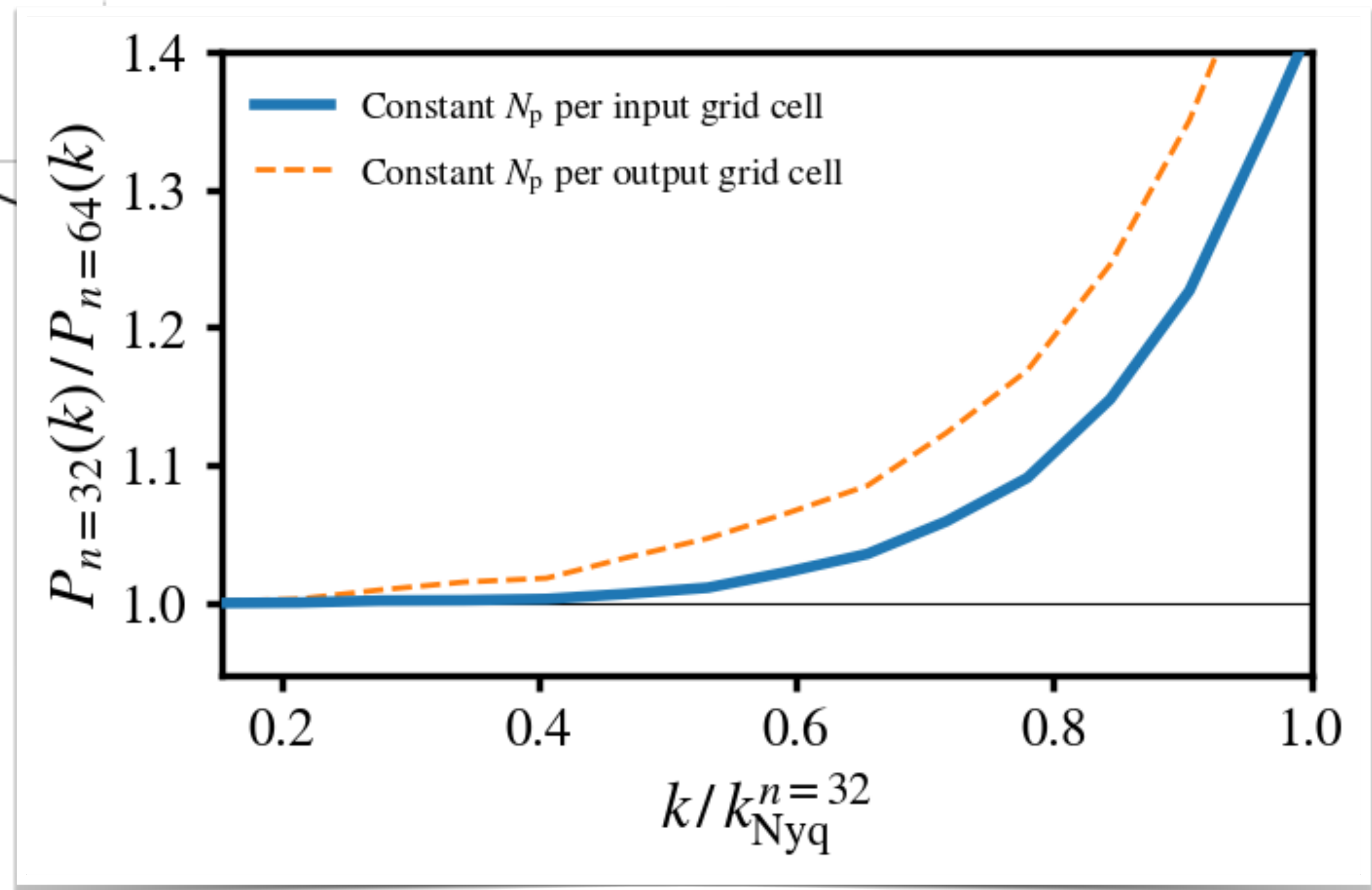
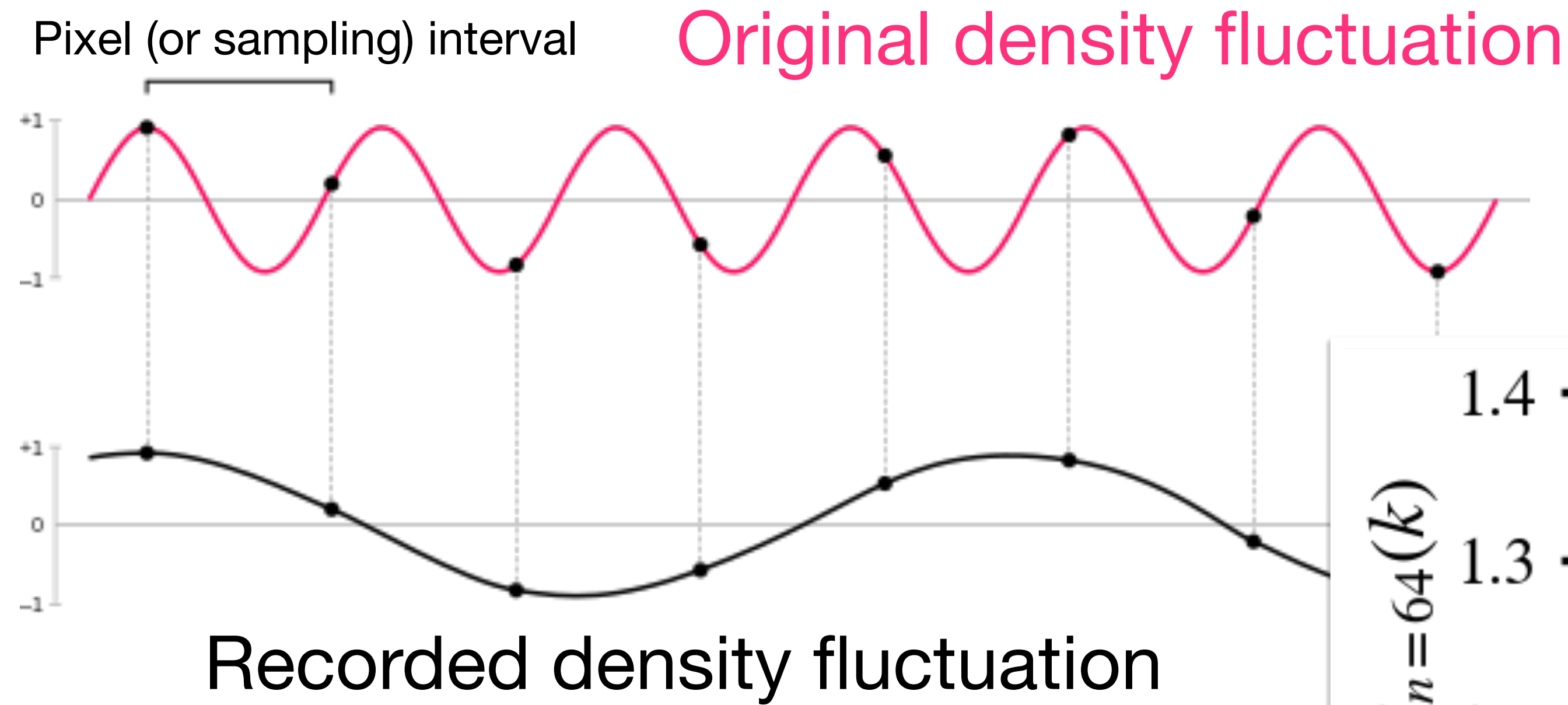
Aliasing from resampling



In Cosmology:



Aliasing from resampling

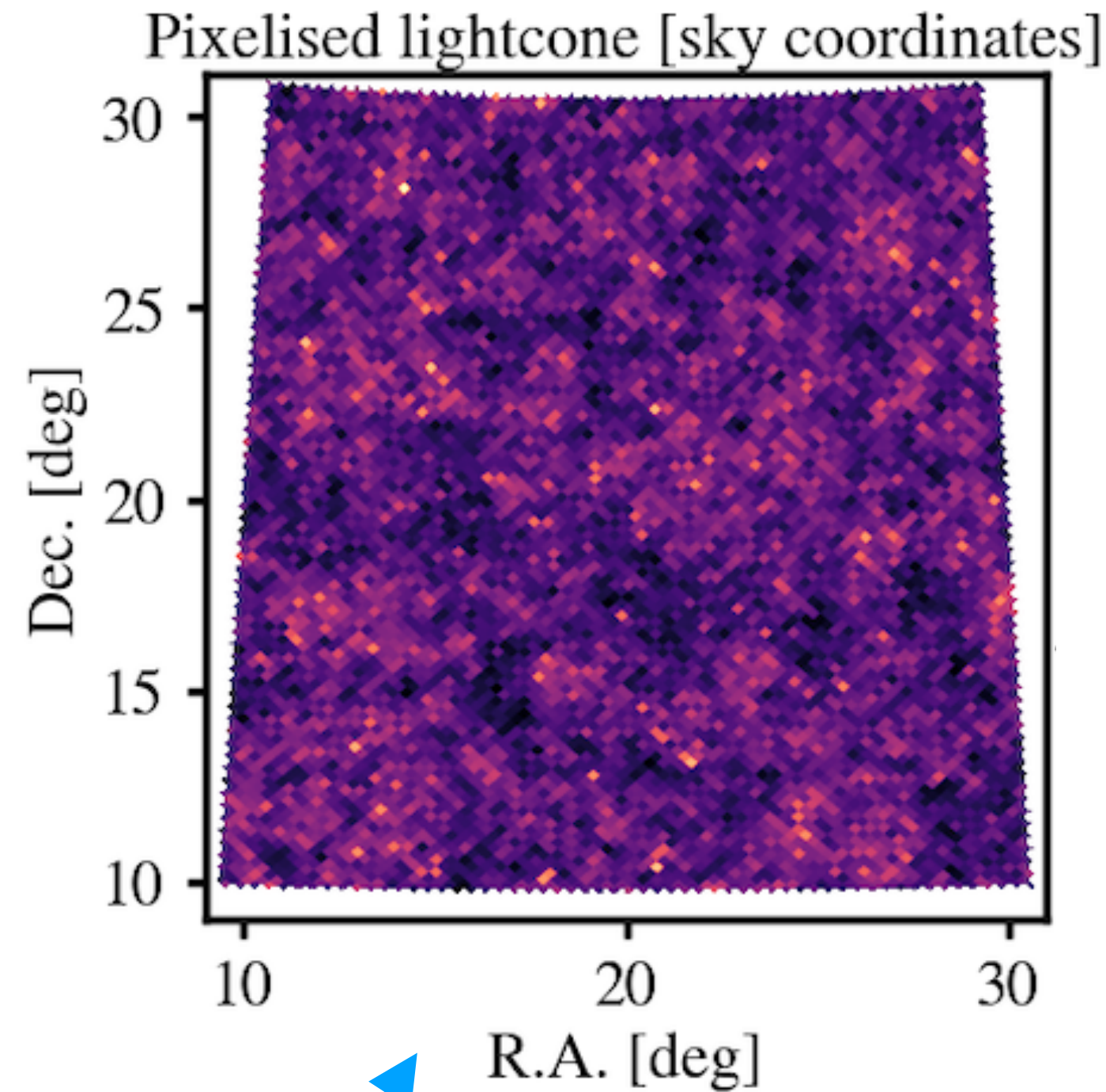
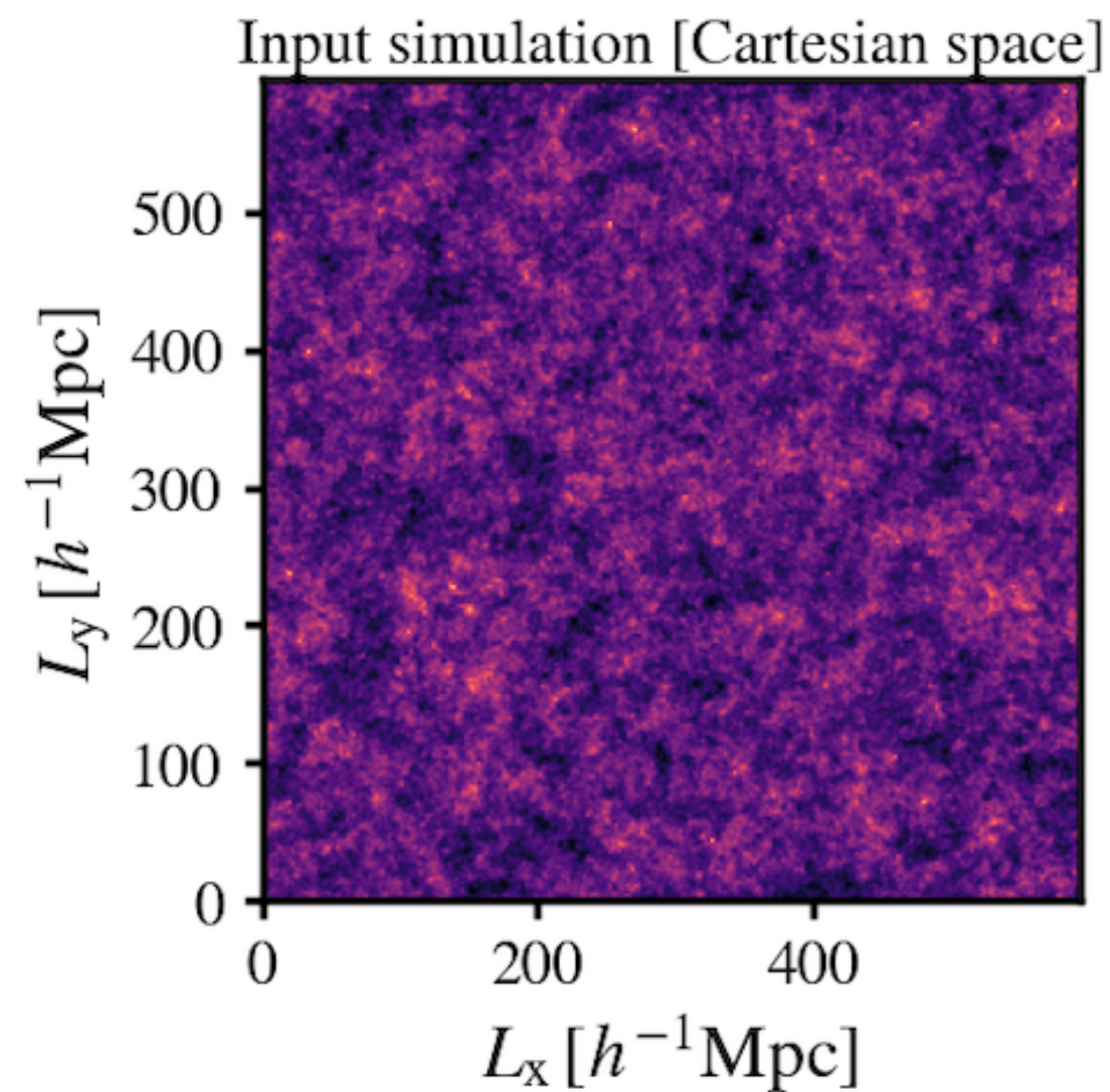


Simulating the mapping/regridding scenario

$$P_m(k)$$

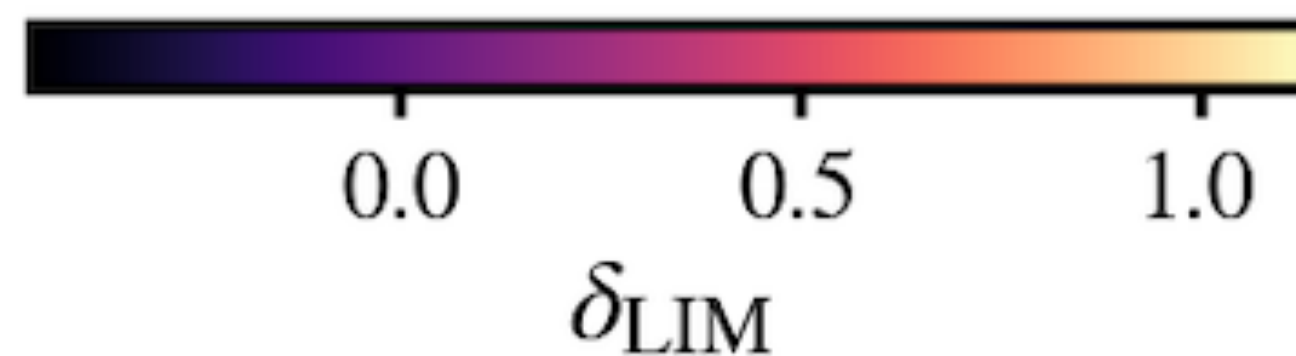
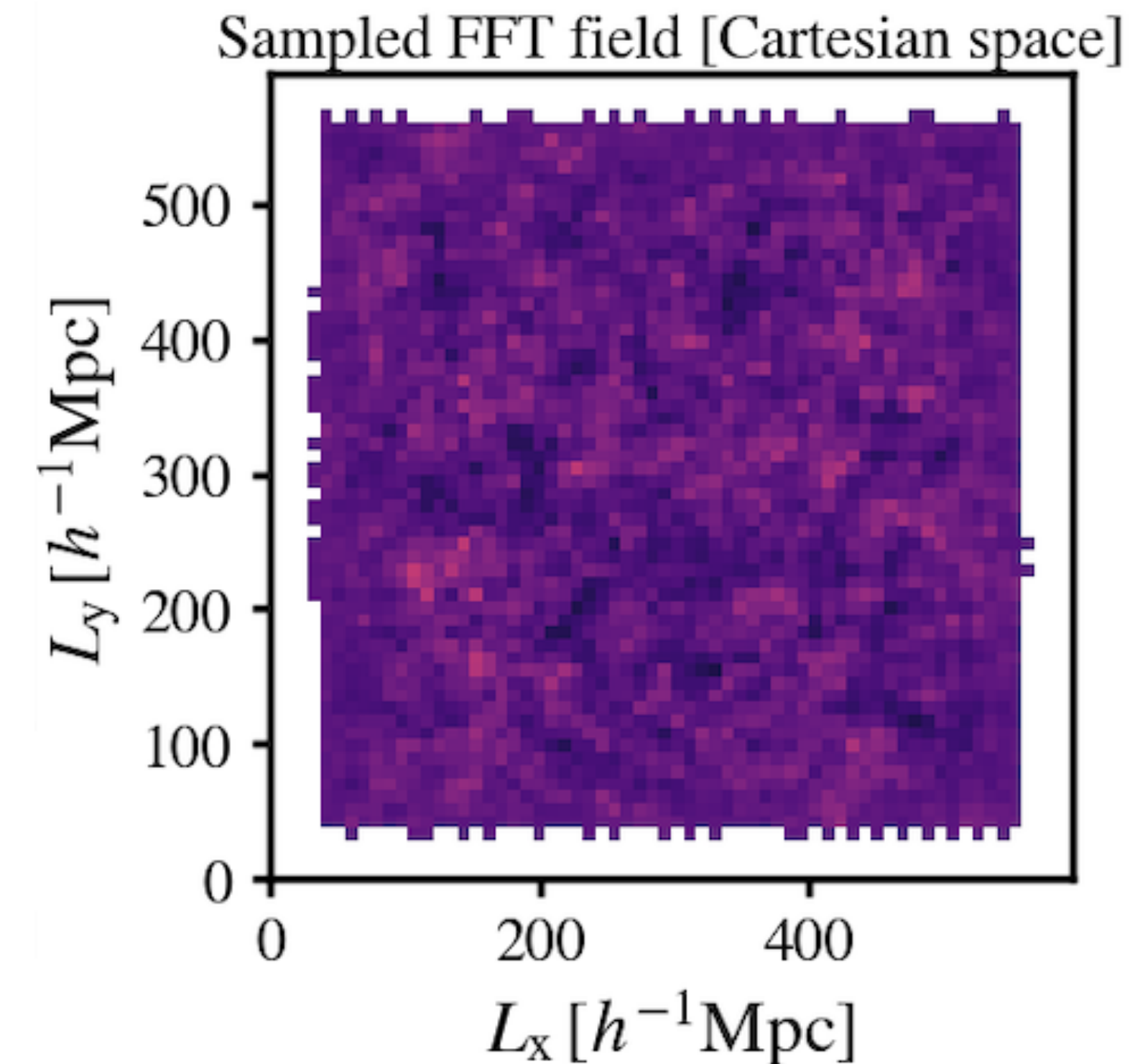
Matter power spectrum

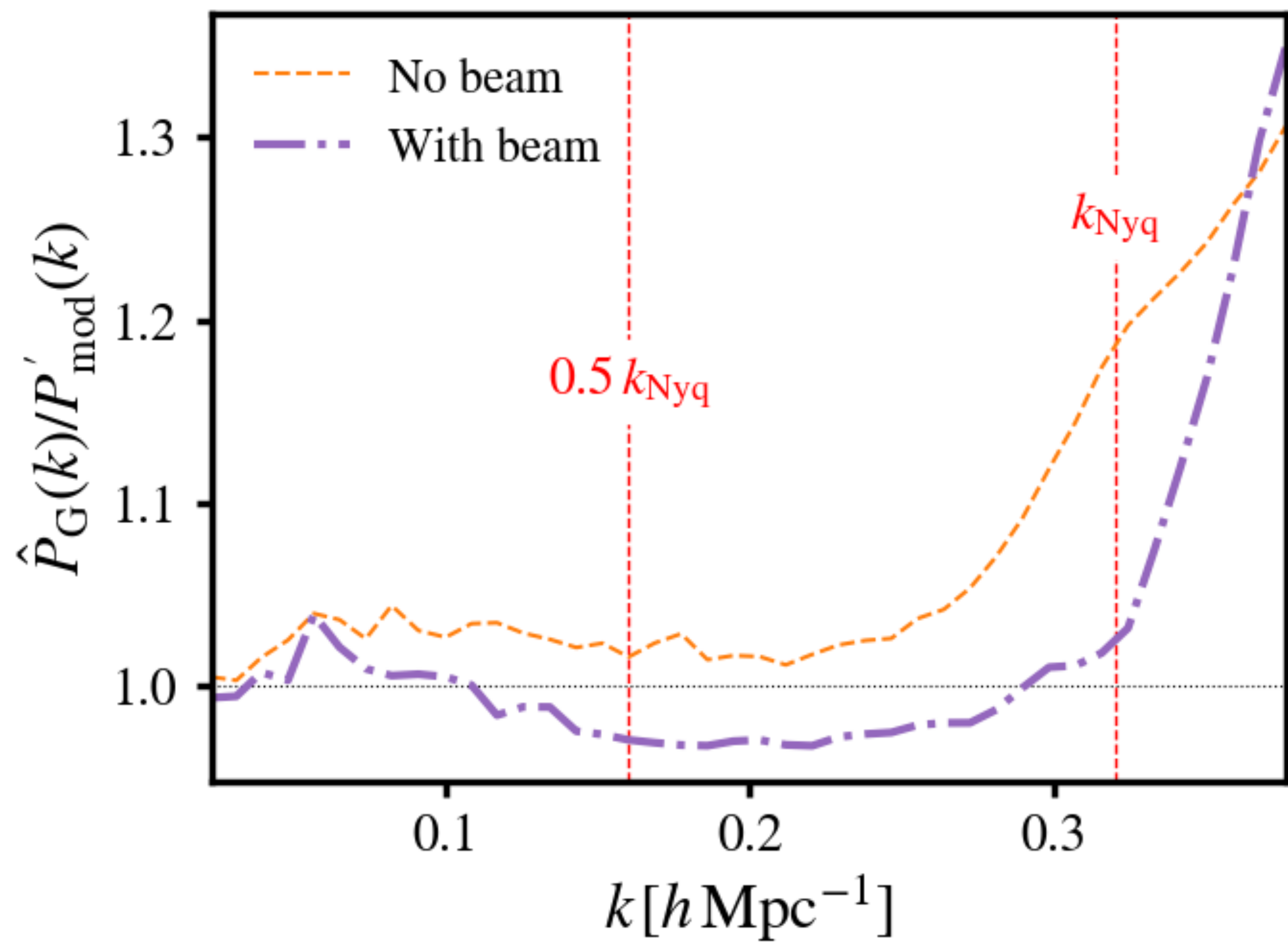
realisation



sample onto a *lightcone*
- emulates starting point of
calibrated line intensity maps

resample back to a
Cartesian grid for FFT





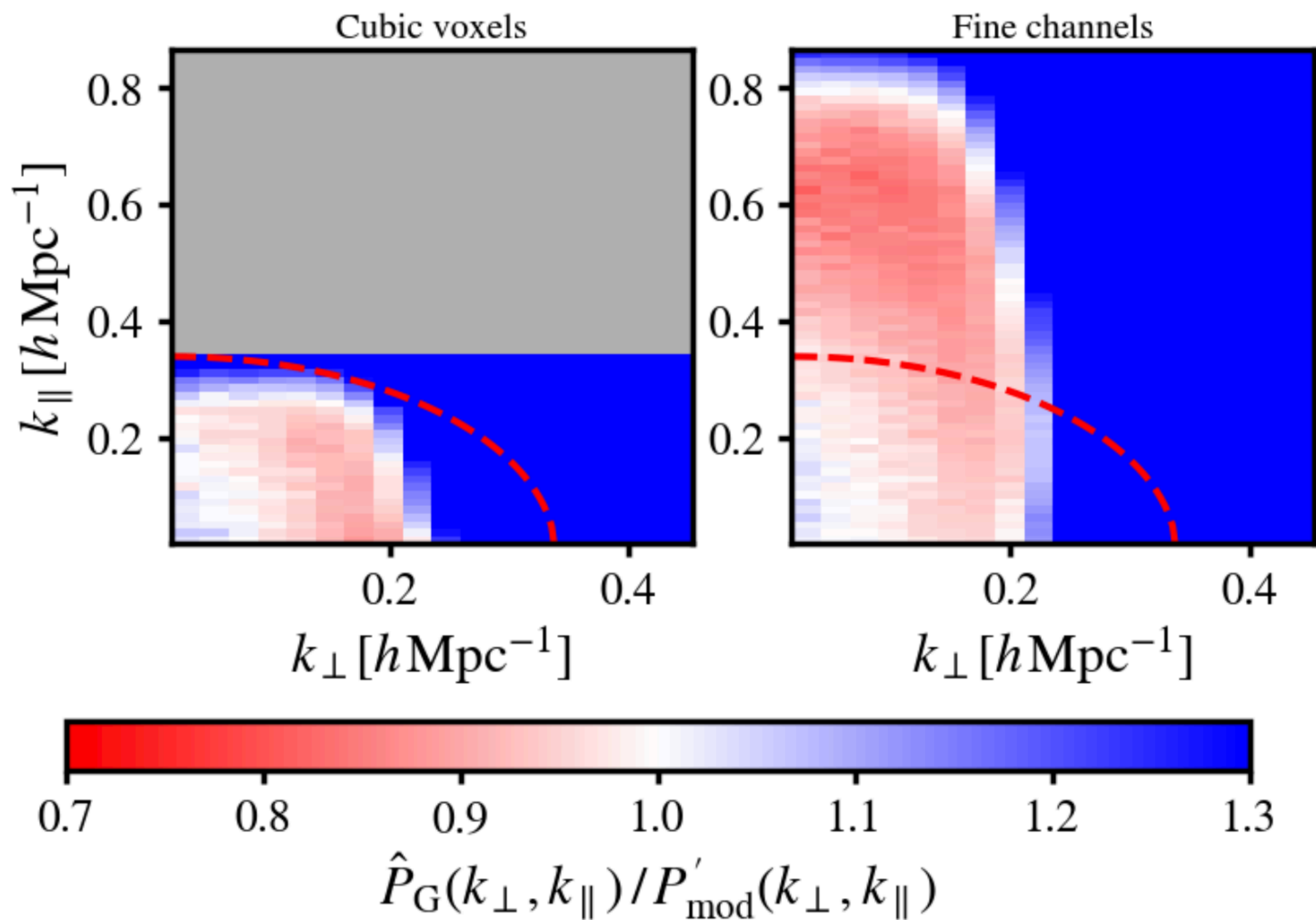


Figure C1. Cylindrical (k_\perp, k_\parallel) power spectrum comparison to model for two simulation versions. Same as Figure 5 but with smoothing effects from the telescope beam. The red dashed line shows the *minimum* Nyquist frequency calculated from the cell resolution along one of the dimensions in the perpendicular directions. Colour-bar range is fixed to avoid saturation.

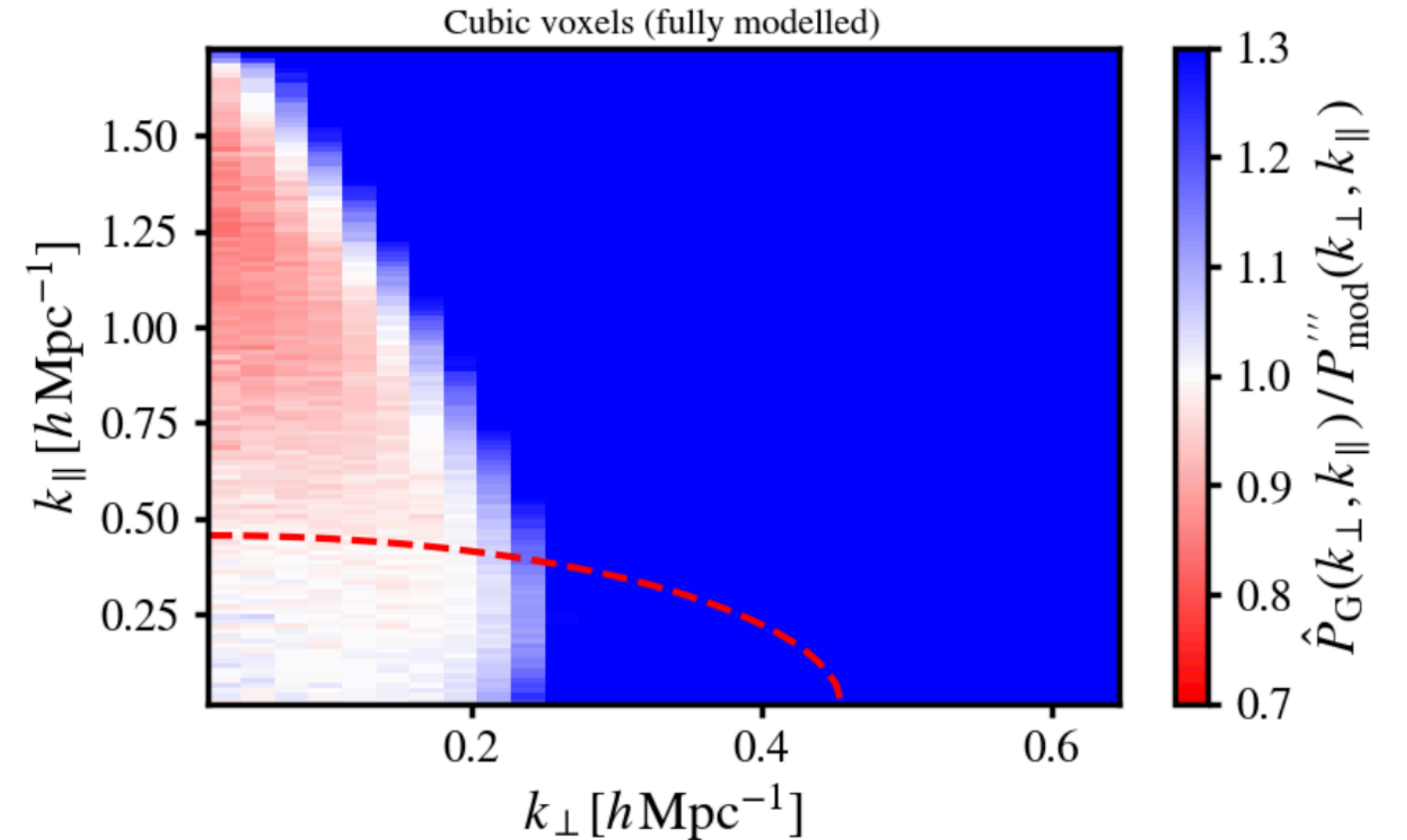


Figure C2. Cylindrical (k_\perp, k_\parallel) power spectrum for the Cubic voxels simulation (with the beam) relative to the final full model P'''_{mod} (Equation 15) including the treatments for discretisation and aliasing from the sky map voxels. This shows the 2D version of Figure 9 where PCS and interlacing has been used in the regridding of the data. The red dashed line shows the *minimum* Nyquist frequency calculated from the cell resolution along one of the dimensions in the perpendicular directions. Colour-bar range is fixed to avoid saturation.

Higher-order interpolation schemes

- Nearest Grid Point (NGP):

$$W_{\text{ngp}}(s) = \begin{cases} 1 & \text{for } |s| < \frac{1}{2} \\ 0 & \text{otherwise} \end{cases}$$

- Cloud-in-Cell (CIC):

$$W_{\text{cic}}(s) = \begin{cases} 1 - |s| & \text{for } |s| < 1 \\ 0 & \text{otherwise} \end{cases}$$

- Triangular-Shaped Cloud (TSC)

$$W_{\text{tsc}}(s) = \begin{cases} \frac{3}{4} - s^2 & \text{for } |s| < \frac{1}{2} \\ \frac{1}{2} \left(\frac{3}{2} - |s| \right)^2 & \text{for } \frac{1}{2} \leq |s| < \frac{3}{2} \\ 0 & \text{otherwise} \end{cases}$$

- Piecewise Cubic Spline (PCS)

$$W_{\text{pcs}}(s) = \begin{cases} \frac{1}{6} (4 - 6s^2 + 3|s|^3) & \text{for } 0 \leq |s| < 1 \\ \frac{1}{6} (2 - |s|)^3 & \text{for } 1 \leq |s| < 2 \\ 0 & \text{otherwise} \end{cases}$$

