

# Field level model for HI: mocks and applications

[Andrej Obuljen \(University of Zurich\)](#)

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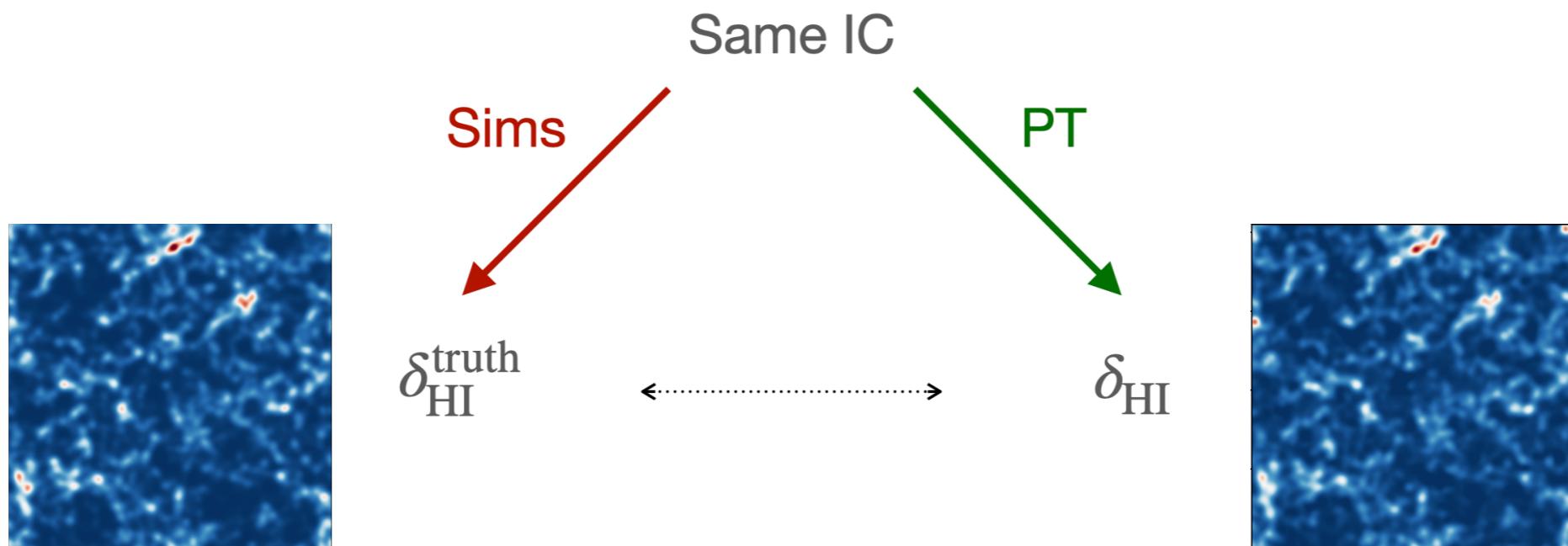
# Motivation

- 21cm IM surveys mainly sensitive to perturbative scales
- We showed we can model HI at the field level using PT+bias models (Schmittfull+18)
- HI noise properties → reconsider cosmological analysis
- Generate fast and accurate 3D mocks: **Hi-Fi mocks**
- Application of Hi-Fi mocks

# Field level approach

## Advantages

- Pixel-by-pixel agreement → agreement of all summary statistics
- No overfitting
- Easy to isolate and study noise
- No cosmic variance for same IC, no need for large hydro sims.



# What is the PT model?

Hybrid Lagrangian and Eulerian scheme, bulk flows included, only linear fields

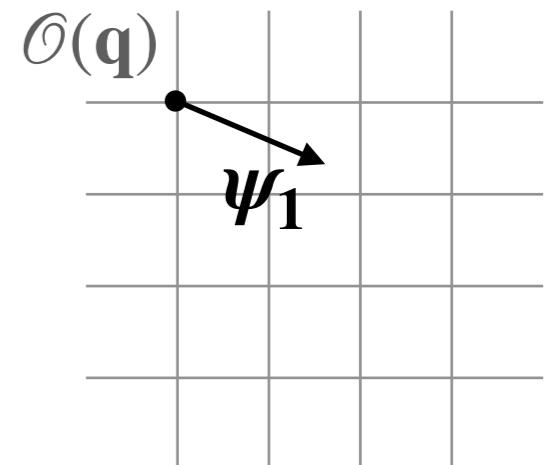
$$\delta_{\text{HI}}(\mathbf{k}) = \int d^3\mathbf{q} (1 + b_1^L \delta_1 + b_2^L (\delta_1^2 - \sigma_1^2) + b_{\mathcal{G}_2}^L \mathcal{G}_2 + \dots - i\mathbf{k} \cdot \boldsymbol{\psi}_2 + \dots) e^{-i\mathbf{k}(\mathbf{q} + \boldsymbol{\psi}_1)}$$

Zel'dovich displacement

Define *shifted bias operators* in Eulerian space:

$$\tilde{\mathcal{O}}(\mathbf{k}) = \int d^3\mathbf{q} \mathcal{O}(\mathbf{q}) e^{-i\mathbf{k}(\mathbf{q} + \boldsymbol{\psi}_1(\mathbf{q}))},$$

where  $\mathcal{O} \in \{1, \delta_1, \delta_2, \mathcal{G}_2, \delta_3, \dots\}$



$$\delta_{\text{HI}}(\mathbf{k}) = \beta_1(k) \tilde{\delta}_1(\mathbf{k}) + \beta_2(k) \tilde{\delta}_2^\perp(\mathbf{k}) + \beta_{\mathcal{G}_2}(k) \tilde{\mathcal{G}}_2^\perp(\mathbf{k}) + \dots + \text{noise}$$

Transfer functions

Matches 1-loop EFT & CLPT power spectrum

# Measure of success

## How to compare model & truth at the field level?

- Minimise mean-squared difference/residuals:

$$P_{\text{err}}(k) \equiv \langle |\delta_{\text{HI}}^{\text{truth}}(\mathbf{k}) - \delta_{\text{HI}}^{\text{model}}(\mathbf{k})|^2 \rangle$$

by doing least squares in each k-bin to obtain best-fit transfer functions:

$$\beta_i = \langle \mathcal{O}_i^\perp \delta_{\text{HI}}^{\text{truth}*} \rangle / \langle |\mathcal{O}_i^\perp|^2 \rangle$$

- Example:

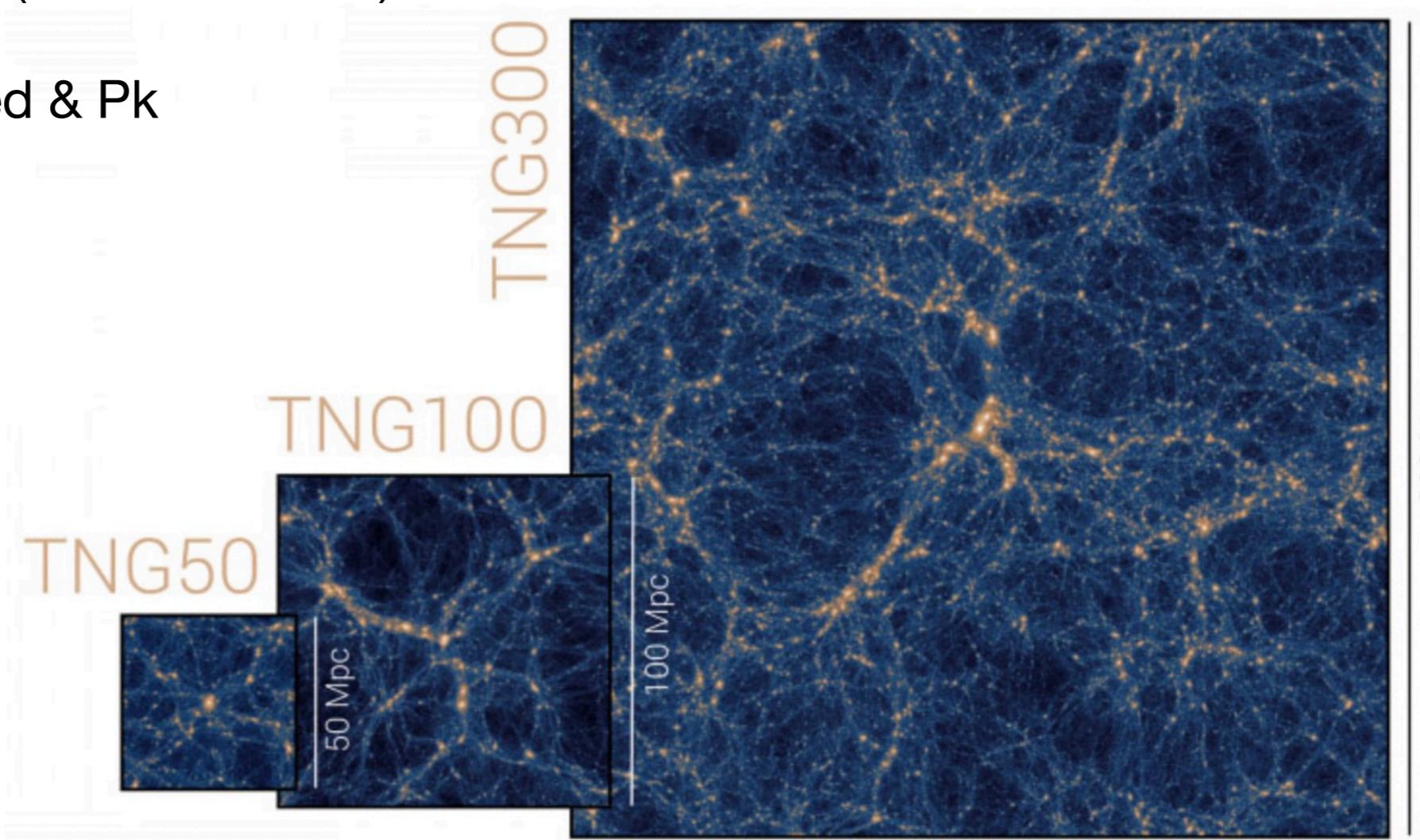
$$\delta_{\text{HI}} = b_1 \delta + \epsilon$$

$$b_1(k) = \langle \delta_{\text{HI}}^{\text{truth}} \delta^* \rangle / \langle |\delta|^2 \rangle$$

$$P_{\text{err}}(k) = \langle |\epsilon|^2 \rangle$$

# HI from IllustrisTNG

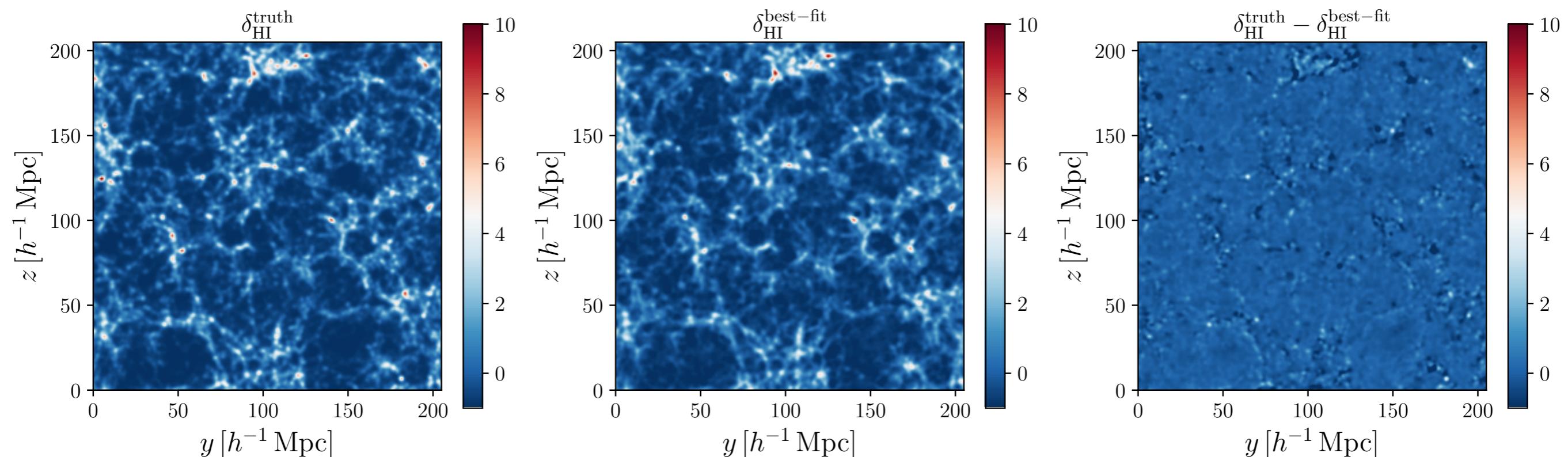
- We apply this approach to a full hydro simulation
- TNG300-1 ( $L = 205 h^{-1}\text{Mpc}$ )
- HI in post-processing (Villaescusa+18)
- Same IC: random seed & Pk



# Results – 3D fields

$z = 1$ , real space

$$\delta_{\text{HI}}(\mathbf{k}) = \beta_1(k)\tilde{\delta}_1(\mathbf{k}) + \beta_2(k)\tilde{\delta}_2^\perp(\mathbf{k}) + \beta_{\mathcal{G}_2}(k)\tilde{\mathcal{G}}_2^\perp(\mathbf{k}) + \beta_3(k)\tilde{\delta}_3^\perp(\mathbf{k}) + \dots + \text{noise}$$

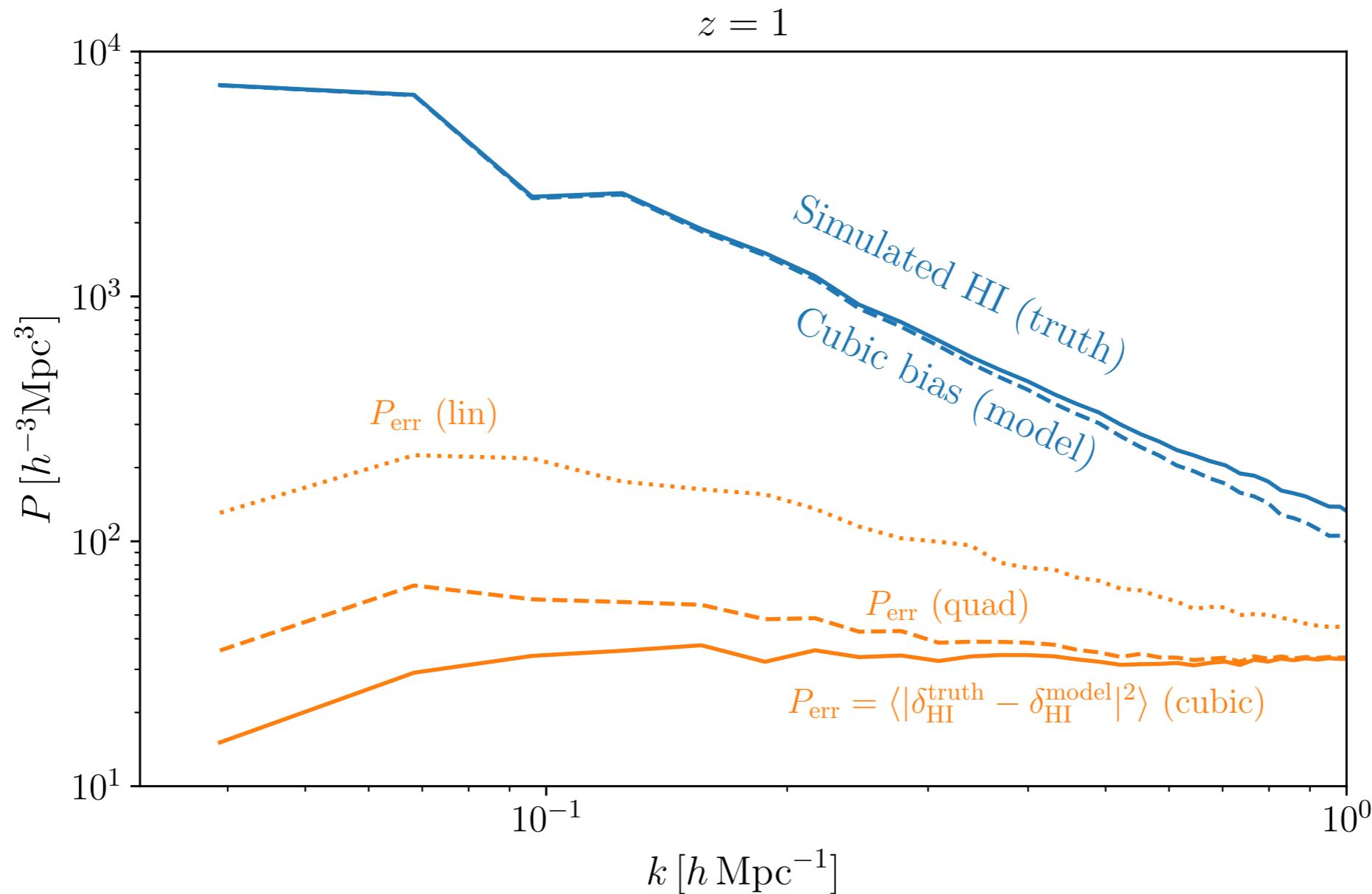


Slice depth 20 Mpc/h, smoothed 1 Mpc/h Gaussian

# Results – power spectrum

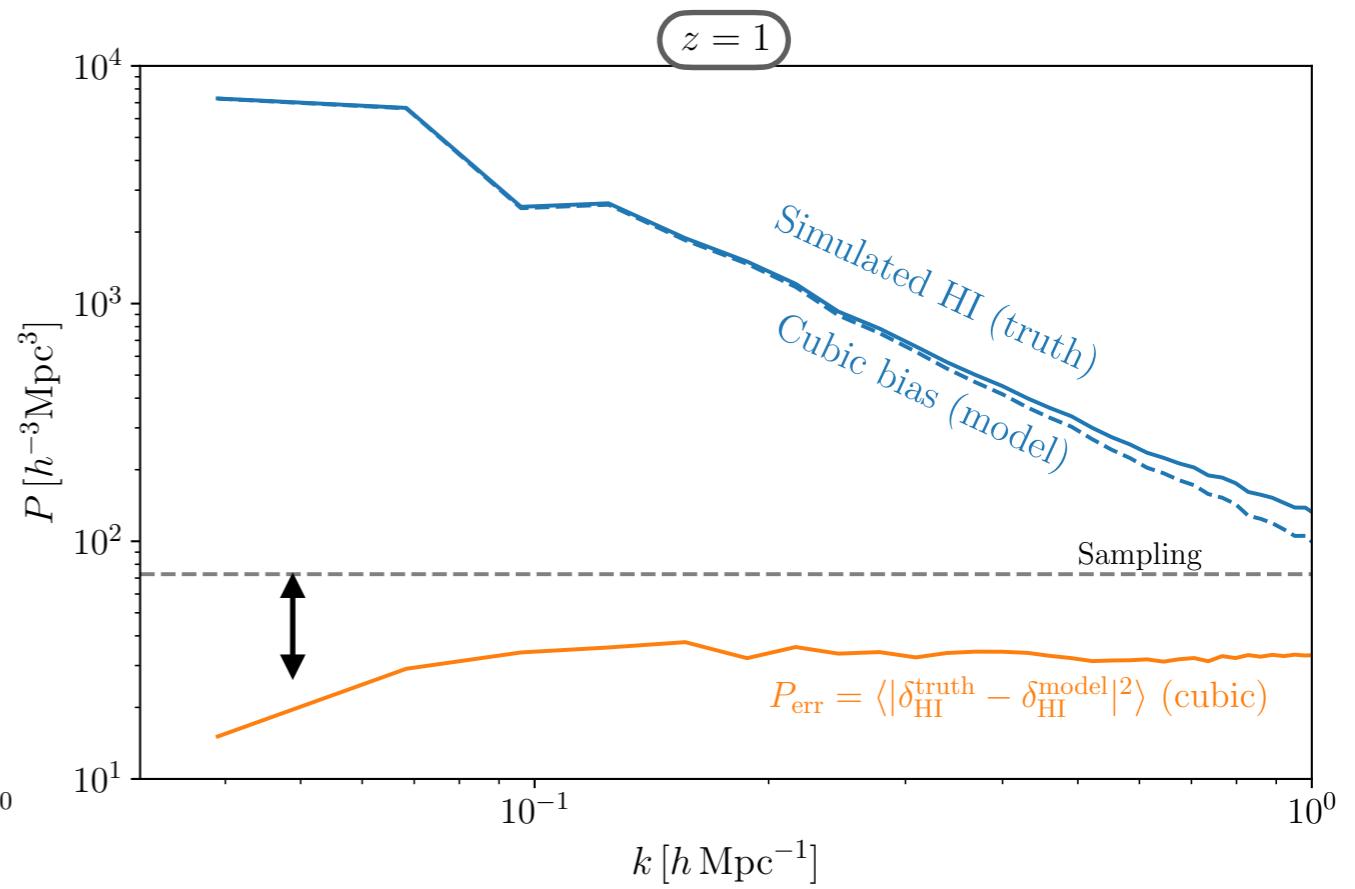
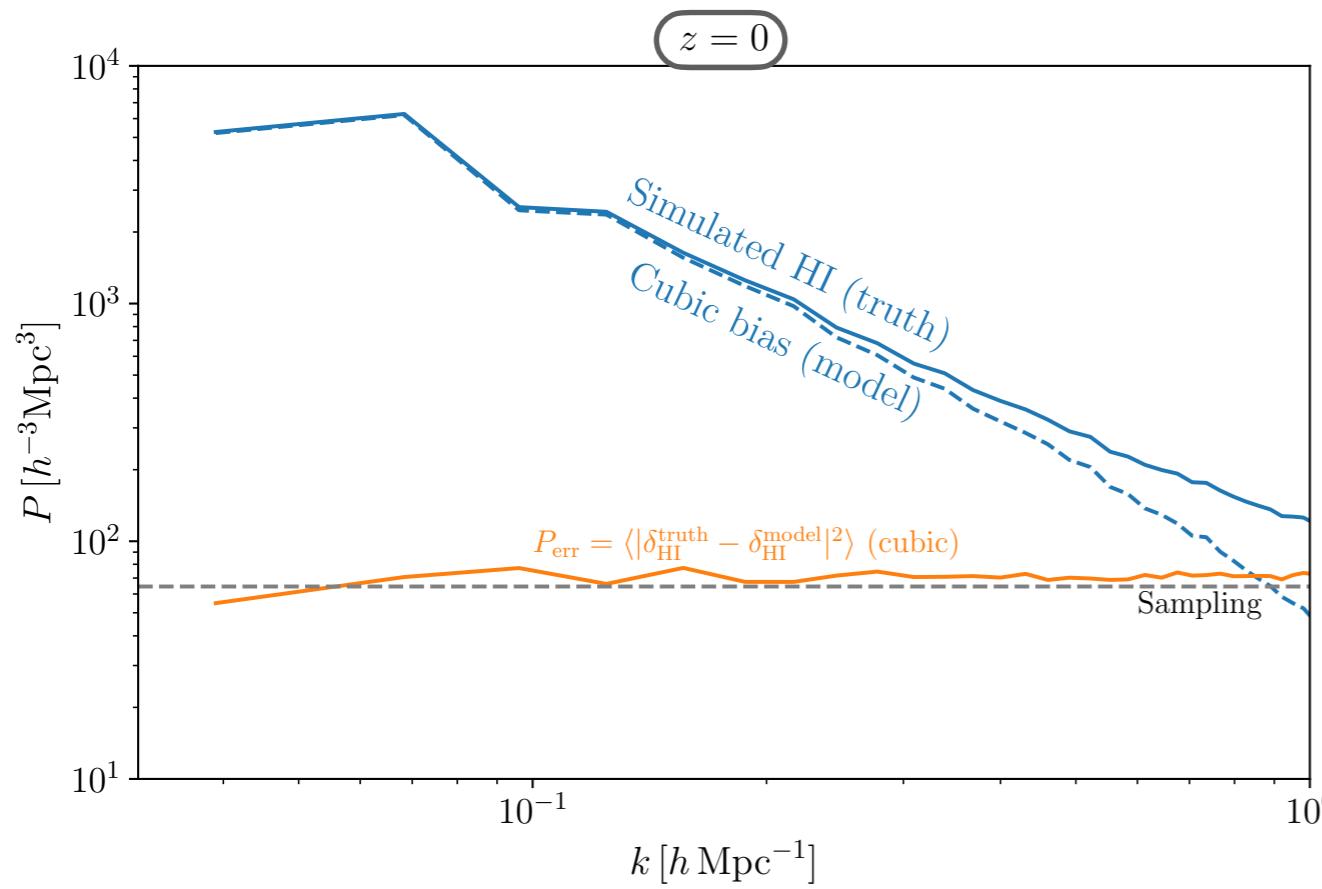
$z = 1$ , real space

$$\delta_{\text{HI}}(\mathbf{k}) = \beta_1(k)\tilde{\delta}_1(\mathbf{k}) + \beta_2(k)\tilde{\delta}_2^\perp(\mathbf{k}) + \beta_{\mathcal{G}_2}(k)\tilde{\mathcal{G}}_2^\perp(\mathbf{k}) + \beta_3(k)\tilde{\delta}_3^\perp(\mathbf{k}) + \dots + \text{noise}$$



# HI noise properties

- $P_{\text{err}}$  flat after including higher order terms
- $P_{\text{err}}$  amplitude comparable to sampling noise ( $\sim 1/n_{\text{bar}}$ ), not equal!

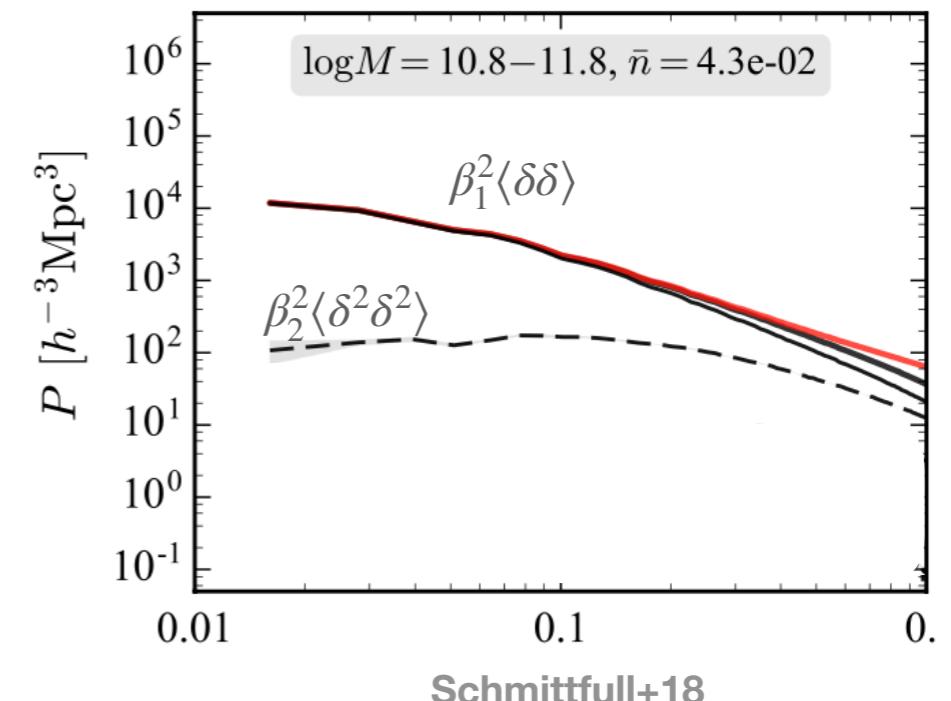
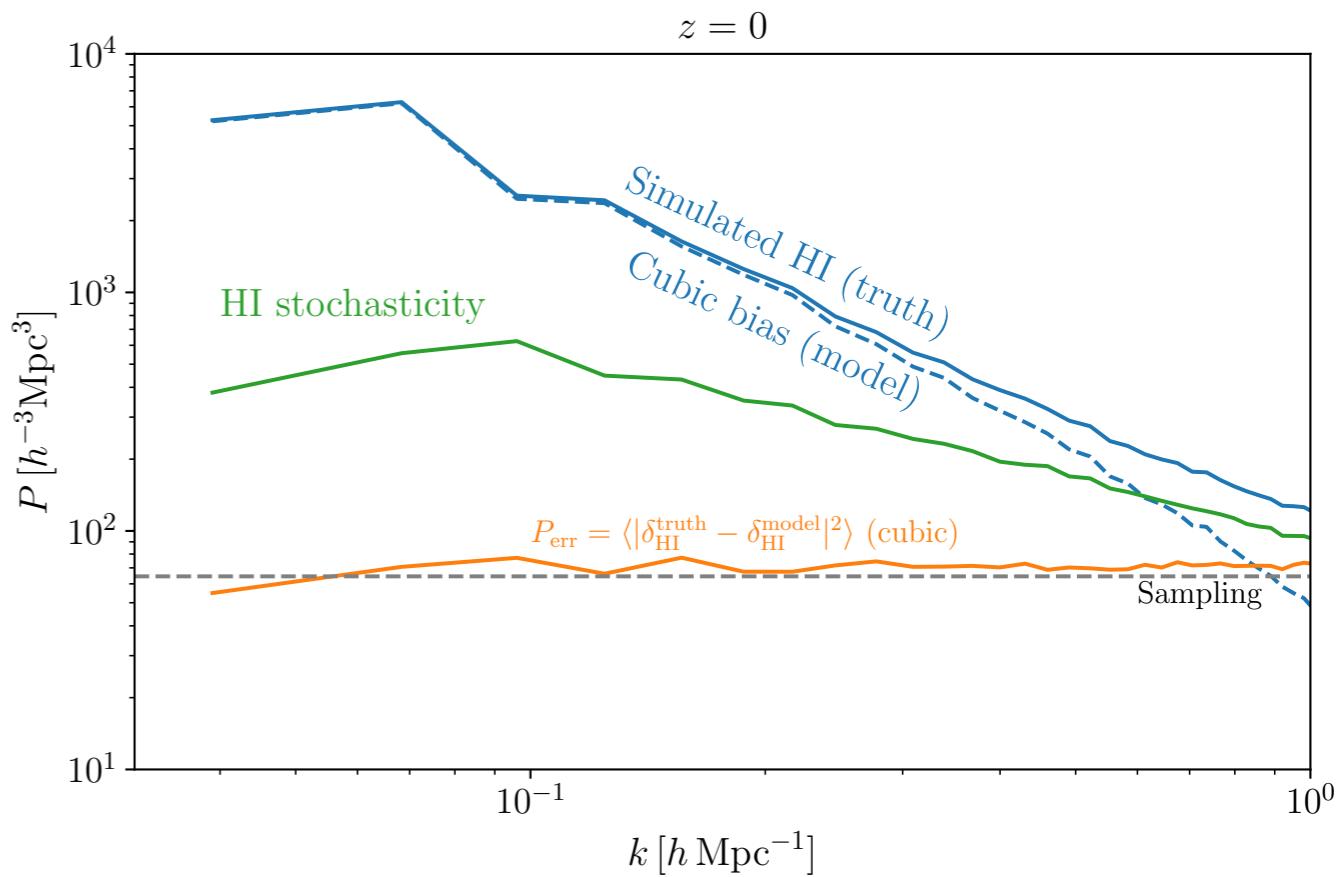


# HI noise properties

- HI stochasticity:  $\langle |\delta_{\text{HI}}^{\text{truth}} - b_1 \delta_m|^2 \rangle$
- $P_{\text{err}}$  lower than stochasticity

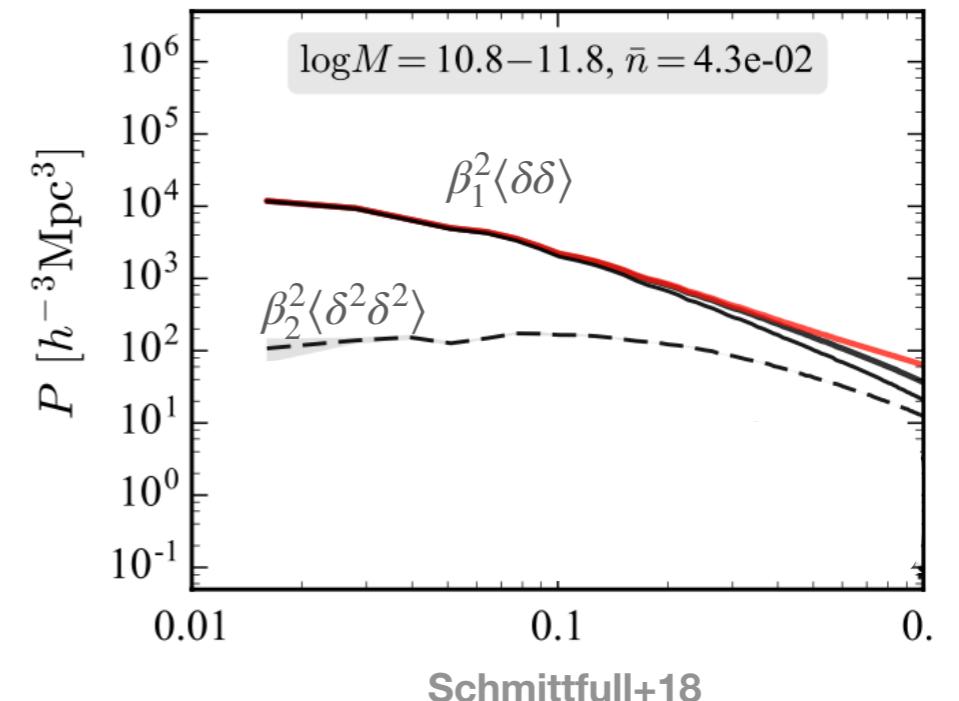
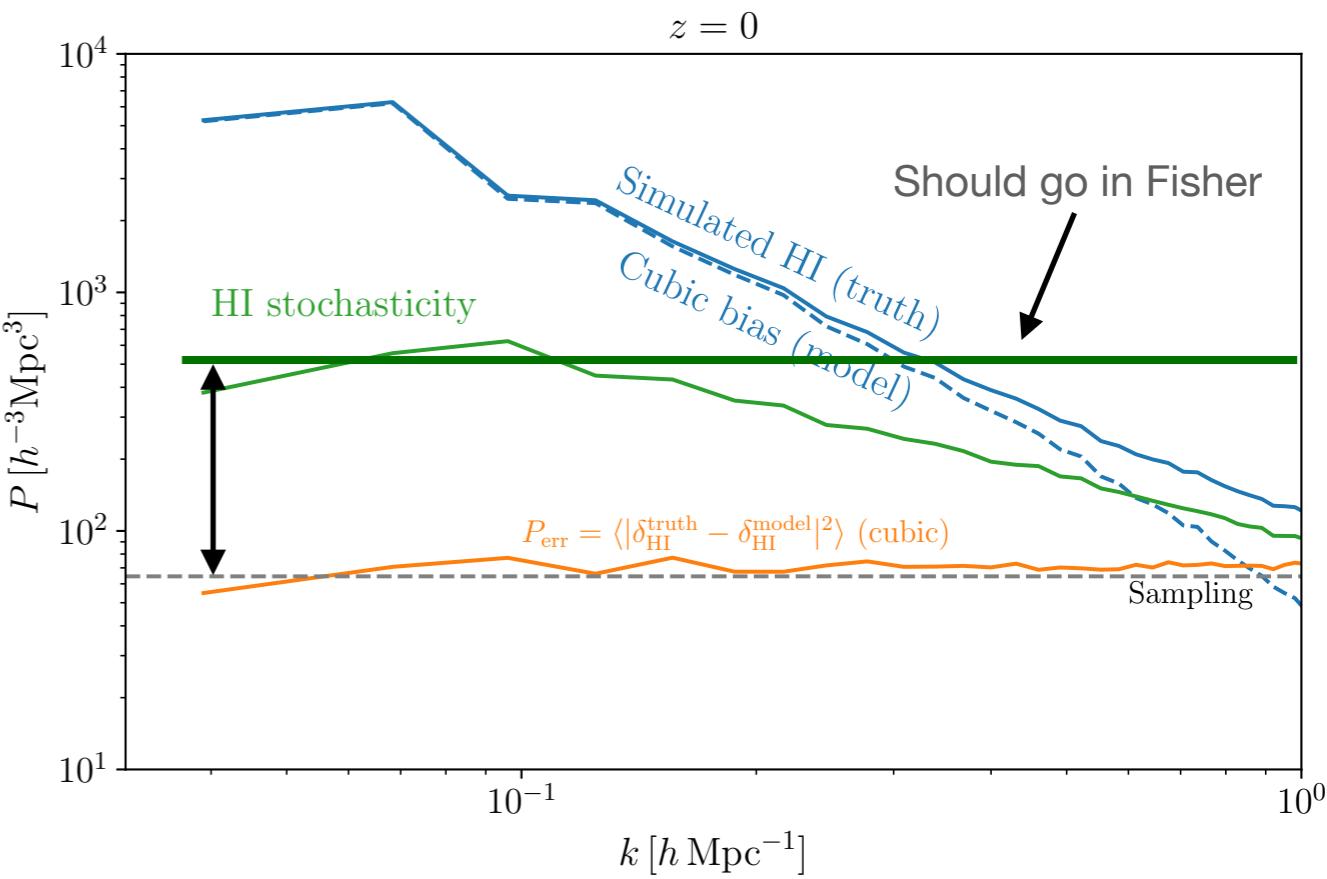
$$\delta_{\text{HI}}(\mathbf{k}) - \beta_1 \tilde{\delta}_1(\mathbf{k}) = \boxed{\beta_2 \tilde{\delta}_2^\perp(\mathbf{k}) + \beta_{\mathcal{G}_2} \tilde{\mathcal{G}}_2^\perp(\mathbf{k}) + \dots + \epsilon}$$

- In contrast to galaxy surveys, higher order terms dominate  $P_{\text{err}}$



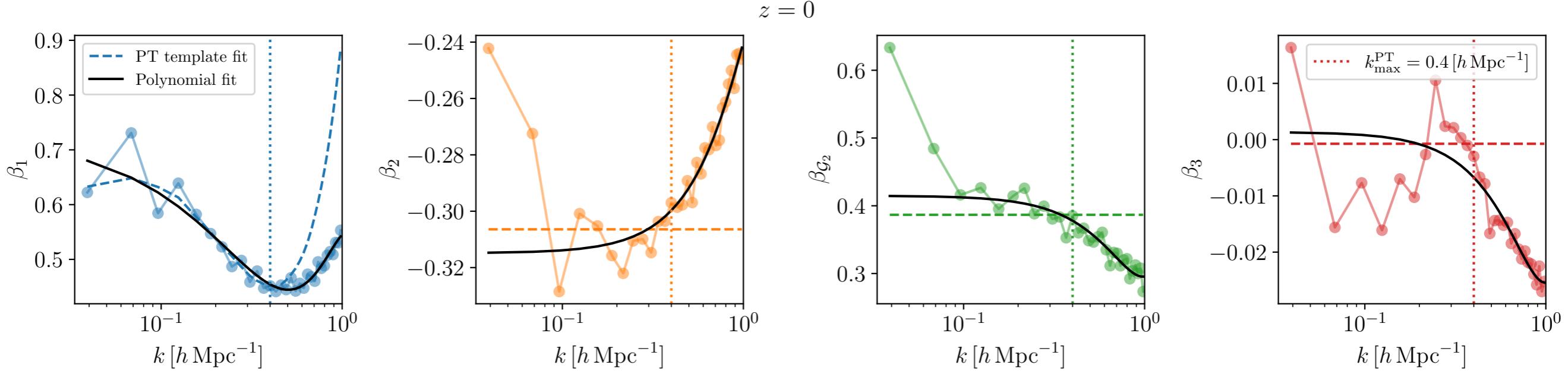
# HI noise properties

- For  $P_k$ , higher order terms flat & degenerate with noise
- Most Fisher forecasts for  $P_k$  assume sampling noise (optimistic)
- Field level may do better!
- For current noise levels, improvement modest...



# Transfer functions

$z = 0$ , real space



$$\delta_{\text{HI}}(\mathbf{k}) = \beta_1(k)\tilde{\delta}_1(\mathbf{k}) + \beta_2(k)\tilde{\delta}_2^\perp(\mathbf{k}) + \beta_{G_2}(k)\tilde{\mathcal{G}}_2^\perp(\mathbf{k}) + \beta_3(k)\tilde{\delta}_3^\perp(\mathbf{k}) + \dots + \text{noise}$$

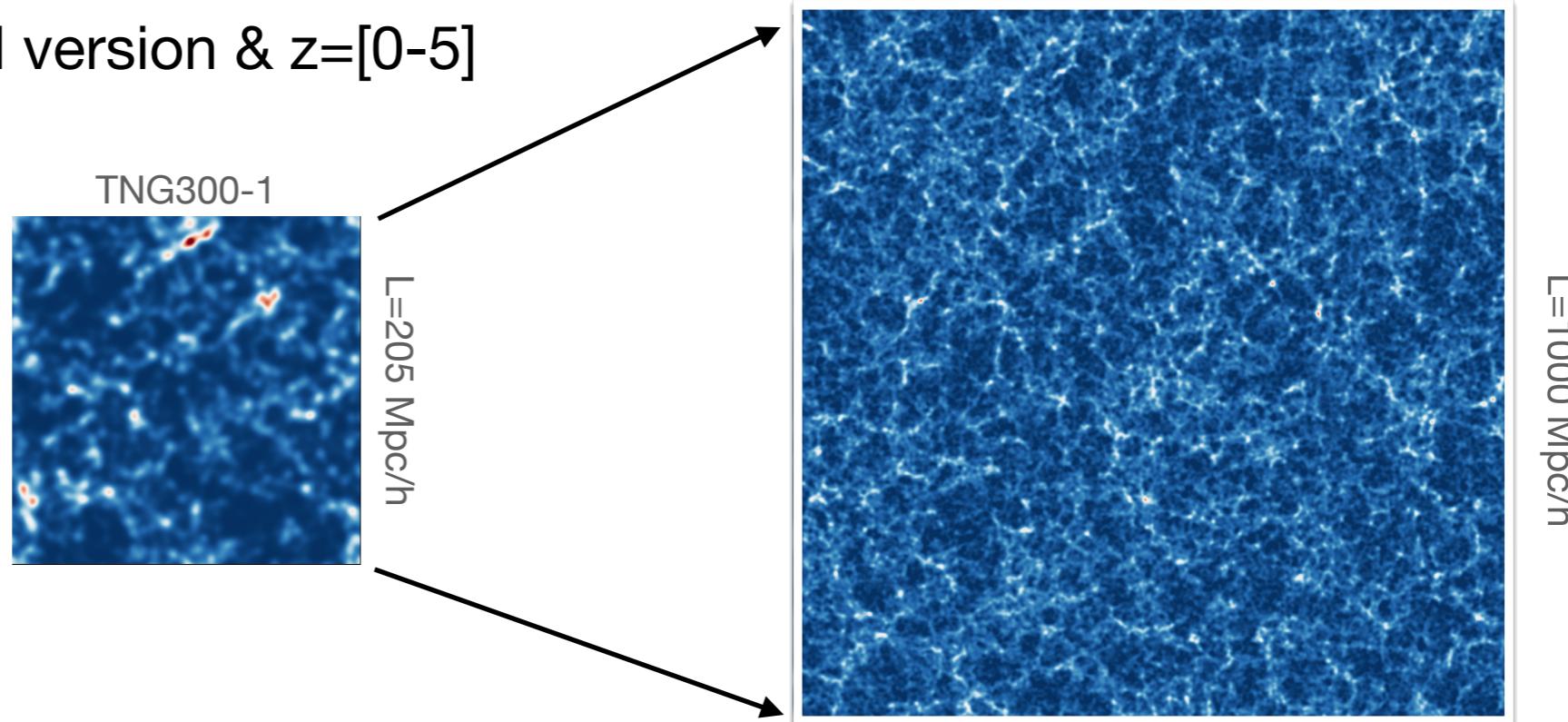
$$\beta_1^{\text{real}}(k) = a_0 + a_1 k + a_2 k^2 + a_4 k^4,$$

$$\beta_{i \neq 1}^{\text{real}}(k) = a_0 + a_2 k^2 + a_4 k^4$$

# Hi-Fi mocks

[https://github.com/andrejobuljen/Hi-Fi\\_mocks](https://github.com/andrejobuljen/Hi-Fi_mocks)

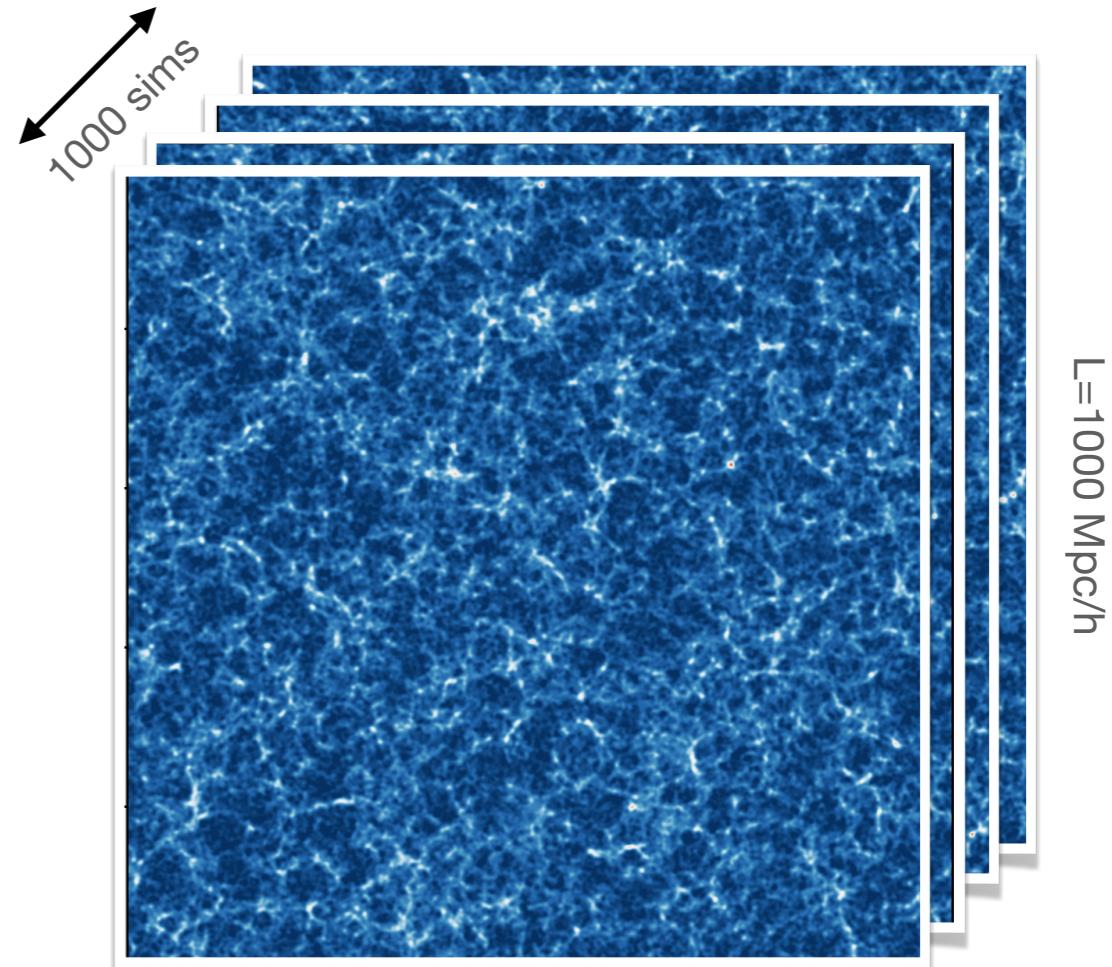
- Generate fast 3D HI field (Hi-Fi) level mocks
- Tuned to TNG HI clustering
- Extendable to any volume!
- Real & redshift space
- Publicly available, give it a try!
- Very soon parallelised version &  $z=[0-5]$



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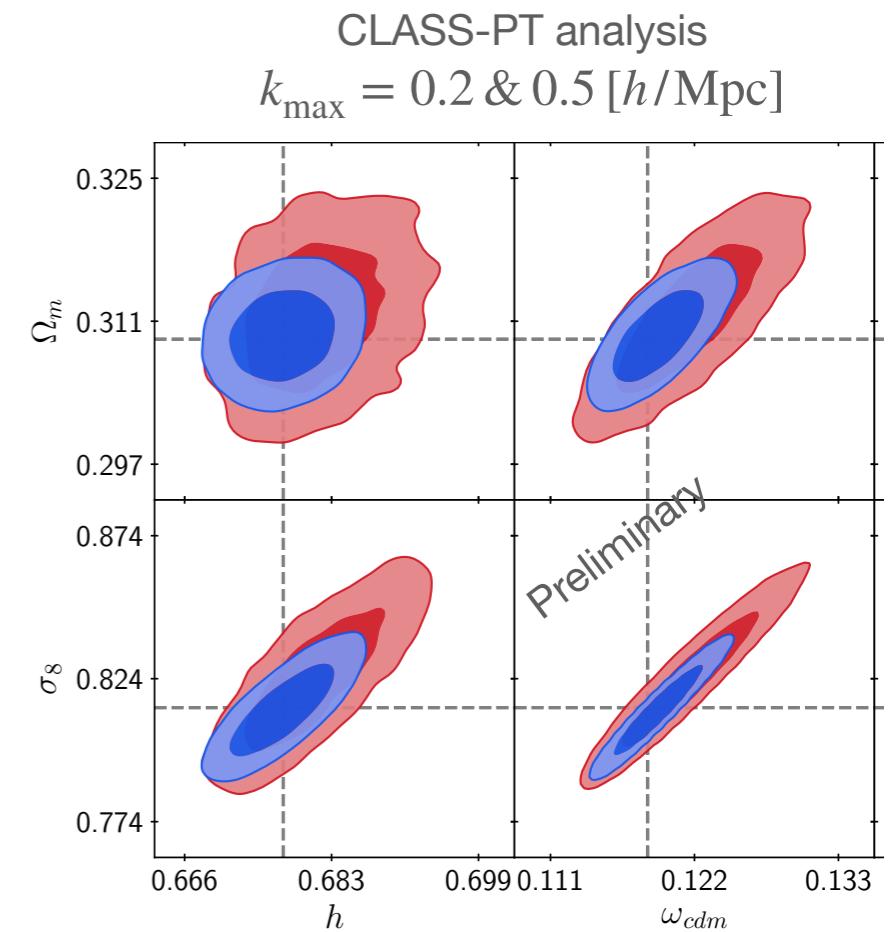
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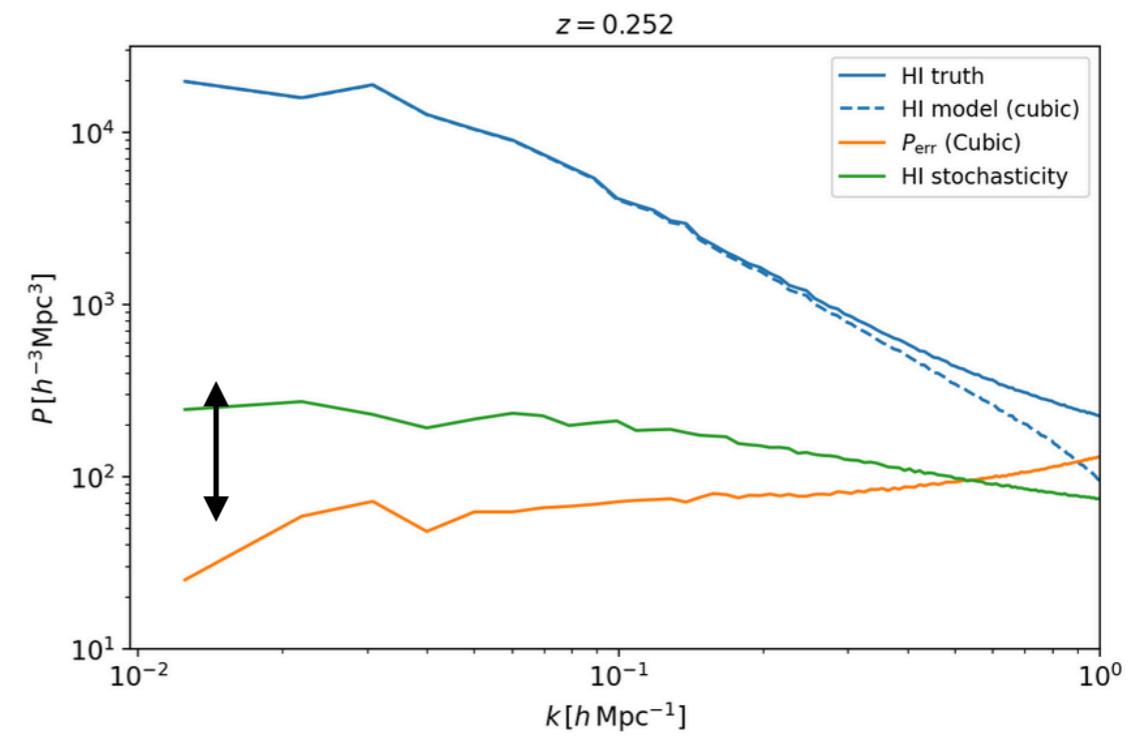
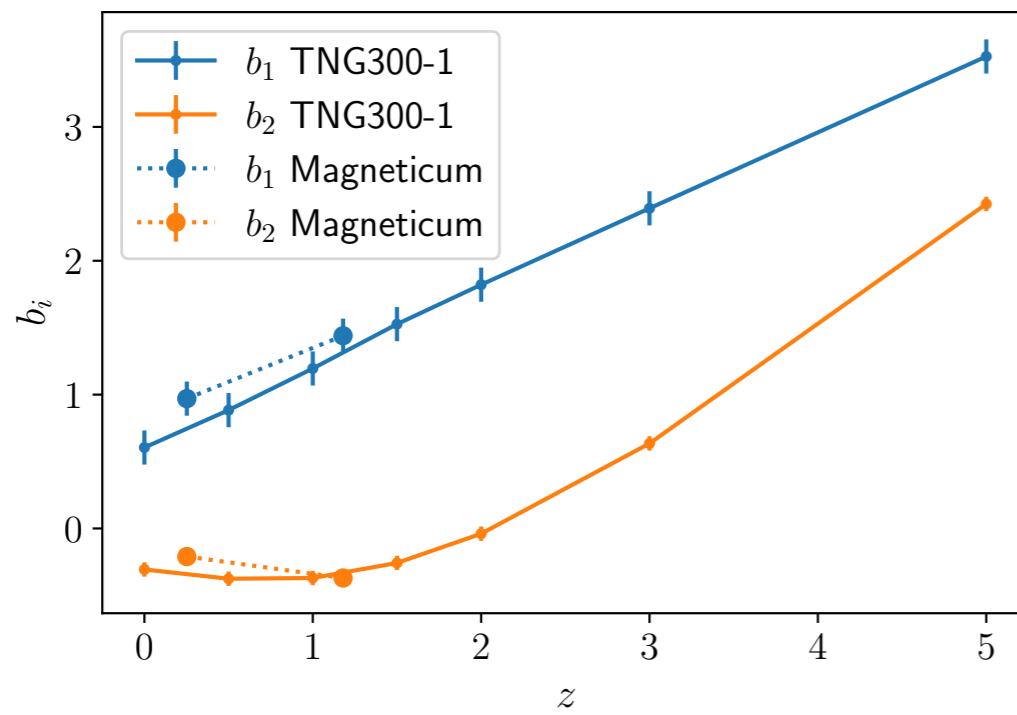
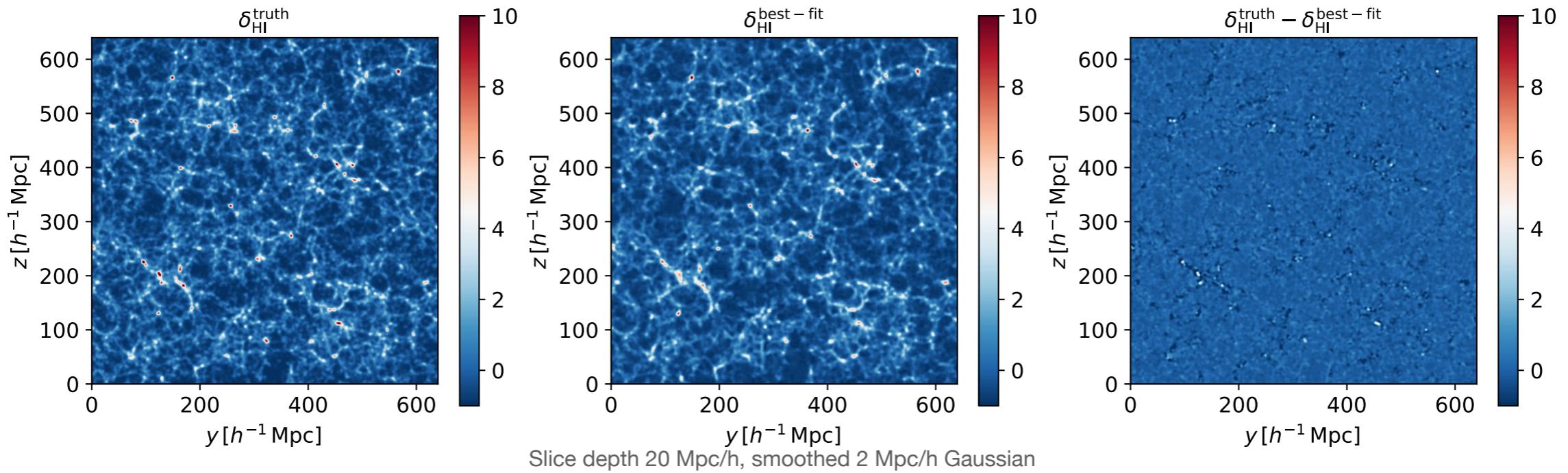
[https://github.com/andrejobuljen/Hi-Fi mocks](https://github.com/andrejobuljen/Hi-Fi_mocks)

- Preliminary!
- $10 \times \text{Gpc}/h$  HI fields for  $P_k$
- 1000 fields for covariance matrix
- $P_k$  analysis in real space
- Unbiased cosmology  $k_{\max} = 0.5$
- Coming soon:  $P_k$  in z-space, bispectrum...



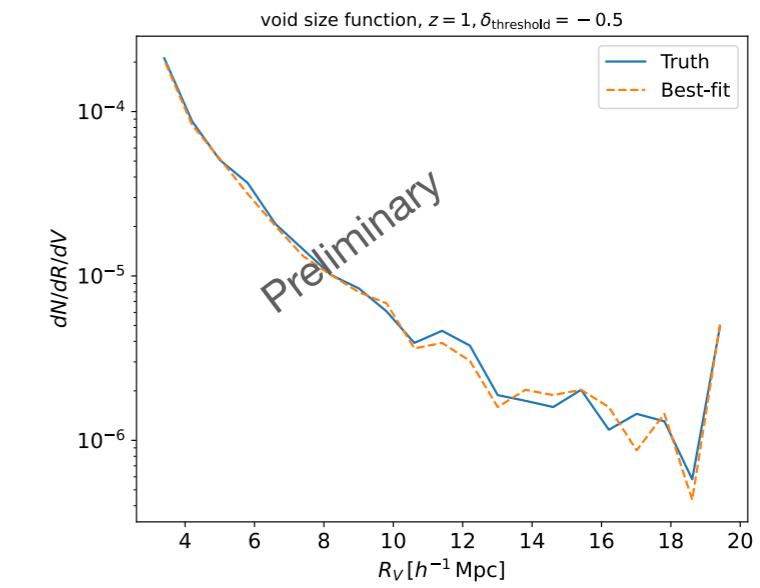
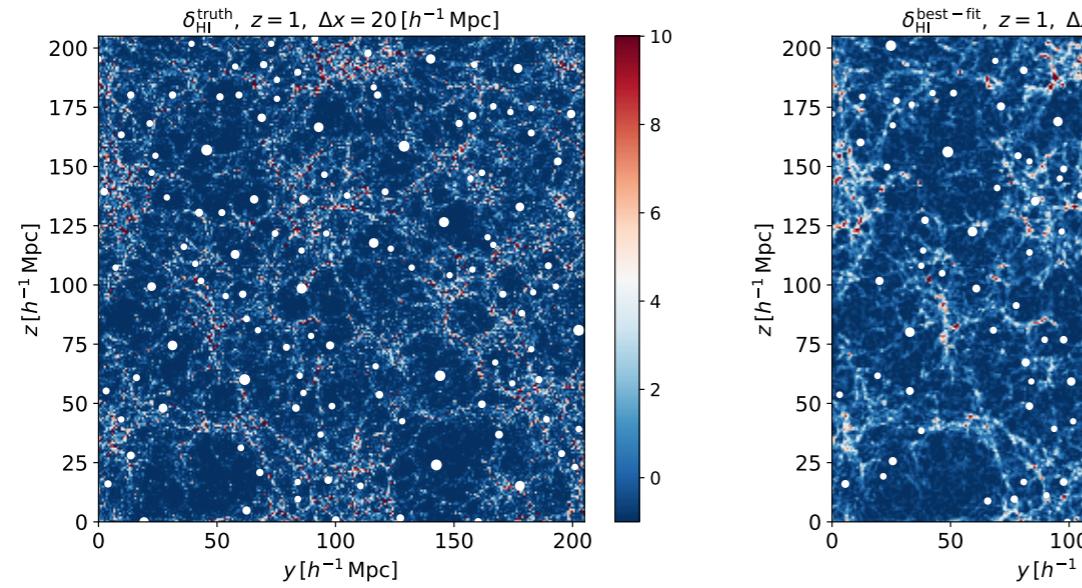
# Other hydro sims?

## Magneticum Box2bhr (640 Mpc/h)

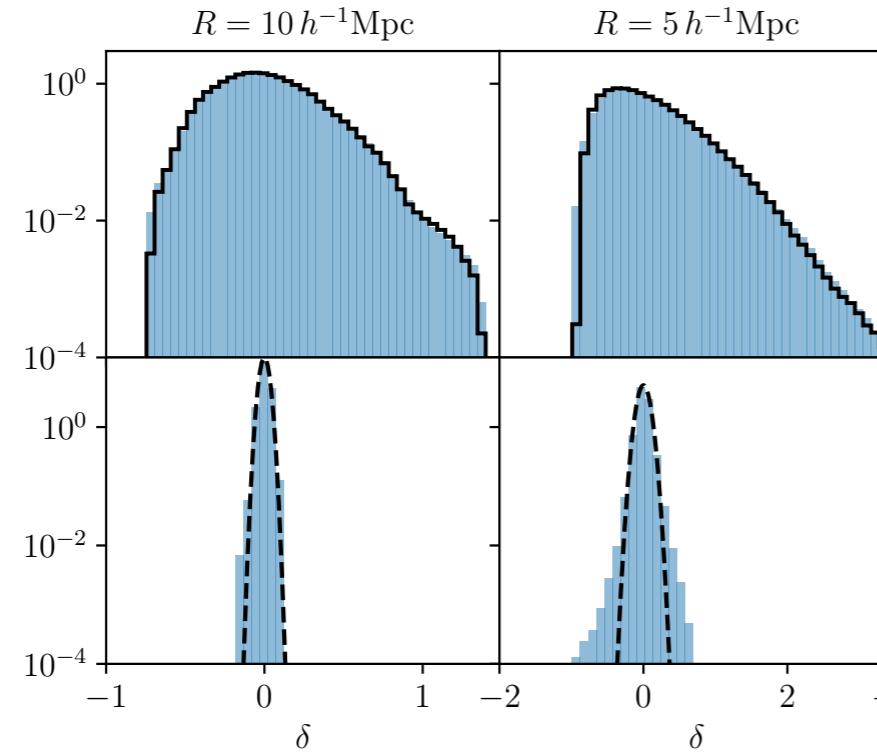


# How about other statistics?

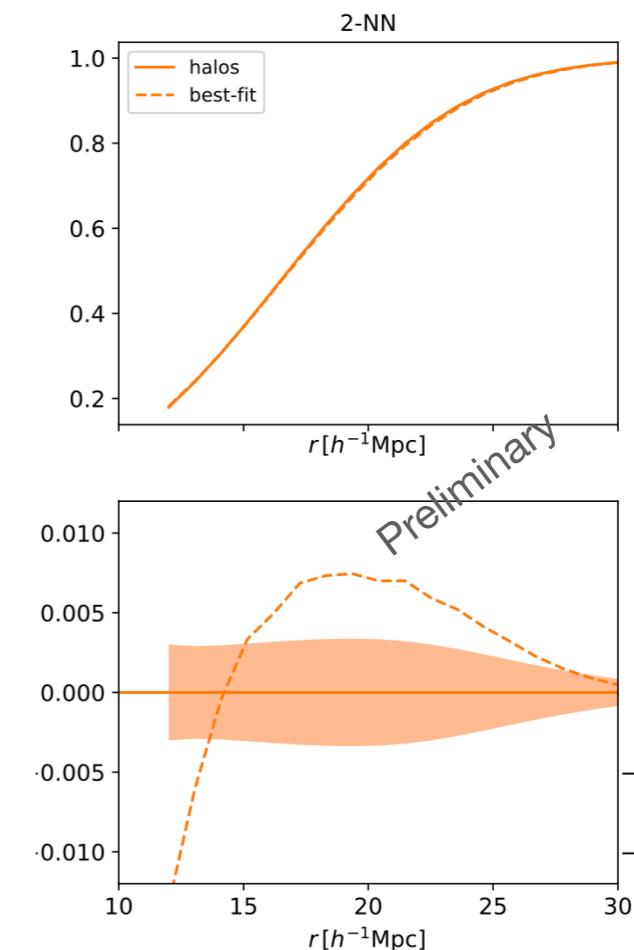
## Void size function



## 1D PDFs



## kNNs



# Looking ahead

- Easily adaptable to any biased tracer!
- Realistic Fisher forecasts for HI: Pk vs ‘field-level’ analysis (in prep)
- Consistently combining constraints from non-Pk observables (1d PDFs, void-size function, kNNs etc) using field-level mocks (ongoing)
- Differentiable forward model in the full field-level inference... (first steps)
- Use ML to describe the residuals...

# Conclusions

- PT + bias model at the field level works well for HI
- HI noise properties motivate beyond Pk analysis
- We provide code to generate fast and accurate HI field level mocks: **Hi-Fi mocks**
- Directly simulate HI field, no need to find halos -> differentiable
- Parallelised version almost ready (Jax version in the works)
- Useful for forecasts, mocks, covariances, future data analysis, KARABO?

Thank you!