

# Constraining local primordial non-Gaussianity using long-wavelength modulation of local small-scale statistics

Fundamental Physics from Future Spectroscopic Surveys  
LBL, Berkeley

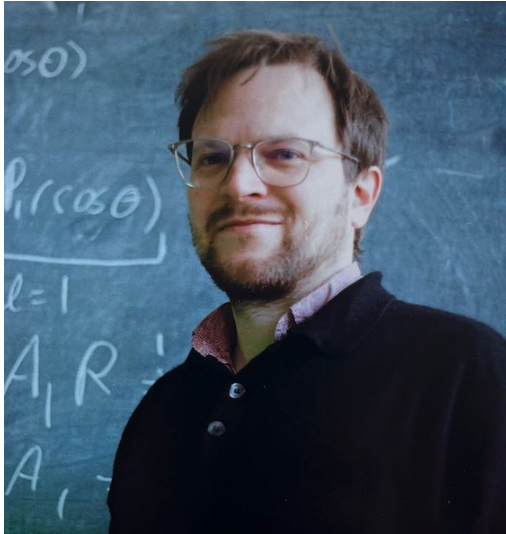


**WISCONSIN**  
UNIVERSITY OF WISCONSIN-MADISON

05/06/2023

**Utkarsh Giri (UW-Madison)**

# Work done in collaboration with



Moritz Münchmeyer  
UW-Madison



Kendrick Smith  
Perimeter Institute



# Introduction/Gist

I will present two simulation based approaches for tightening constraints on  $f_{NL}$

1a. A non-perturbative approach sensitive to  $f_{NL}$  via higher  $N$ -point functions.

- A *neater* formulation+use of position-dependent power spectrum approach (Chiang, Wagner, Schmidt, Komatsu 2014, Smith, Loverde, Zaldarriaga 2012, de putter 2018 and others)
- *Very efficient* from computational and modelling point of view
- Allows us to tap into very non-linear scales, outperforms some recent forecasts.

1b. A machine learning enhanced version of the above formalism.

- In principle sensitive to  $f_{NL}$  information in soft limits of all higher  $N+1$  point functions

We harness large-small mode coupling feature of PNG universe + LSS consistency relation guarantees of Gaussian universe, to **derive a large-scale field  $\pi^f$  composed out of very non-linear density field, which has a clean  $f_{NL}/k^2$  bias similar to halos**

# $f_{NL}$ from LSS: Current status and motivation

The most promising method has been via the

PNG induced scale-dependent bias

$$P_{hh}(k_L) = \left( b_G + \frac{b_{NG} f_{NL}}{k_L^2} \right)^2 P_{mm}(k_L)$$

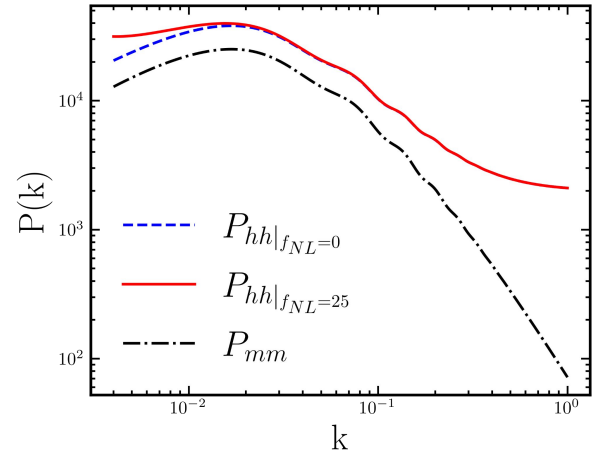
Additionally,  $B_{hhh}$  contains term  $\propto \frac{f_{NL}}{k_L^2}$

However, they are:

- Hard to model analytically beyond  $k \sim 0.2$  h/Mpc
- Simulation-based approaches are computationally very challenging

SOTA:  $P_k + B_k$  constraints:  $\sigma(f_{NL}) = \mathcal{O}(20 - 30)$  [Cabass et al. 22, D'Amico et al]

**Question: How much would a proper inclusion of higher N-point functions tighten constraints on  $f_{NL}$ ? How to best do it in a manageable way? (! Several recent studies report constraints saturating around  $k \sim 0.3$ !!)**



# A non-perturbative approach

**Main idea:  $1/k^2$  will appear in any field sensitive to small scale power**

(Covered nicely by Oliver Philcox and Sam Goldstein's talk earlier)

Define a field  $\pi^f$  which is a measure of local power/variance for observable  $f$

$$\pi^f(\mathbf{x}) = \left( \int \frac{d^3\mathbf{k}}{(2\pi)^3} W(k) \rho_f(\mathbf{k}) e^{i\mathbf{k}\cdot\mathbf{x}} \right)^2 \quad W^i(\mathbf{k}) = \begin{cases} 1 & \text{if } k_{\min}^i < |\mathbf{k}| < k_{\max}^i \\ 0 & \text{elsewhere} \end{cases}$$

$f$  could be late time matter field  $\delta_m$  or a tracer like the halo number density  $n_h$

$$P_{m\pi^m}(\mathbf{k}_L) = \left( b_\pi + 2\beta_\pi \frac{f_{NL}}{k^2} \right) P_{mm}(\mathbf{k}_L)$$

$$P_{\pi^{m'}\pi^m}(\mathbf{k}_L) = \left( b_{\pi'} + 2\beta_{\pi'} \frac{f_{NL}}{k^2} \right) \left( b_\pi + 2\beta_\pi \frac{f_{NL}}{k^2} \right) P_{mm}(\mathbf{k}_L) + N_{\pi\pi'}$$

sensitive to **squeezed bispectrum & collapsed trispectrum**

# Testing our conjectures using simulations.

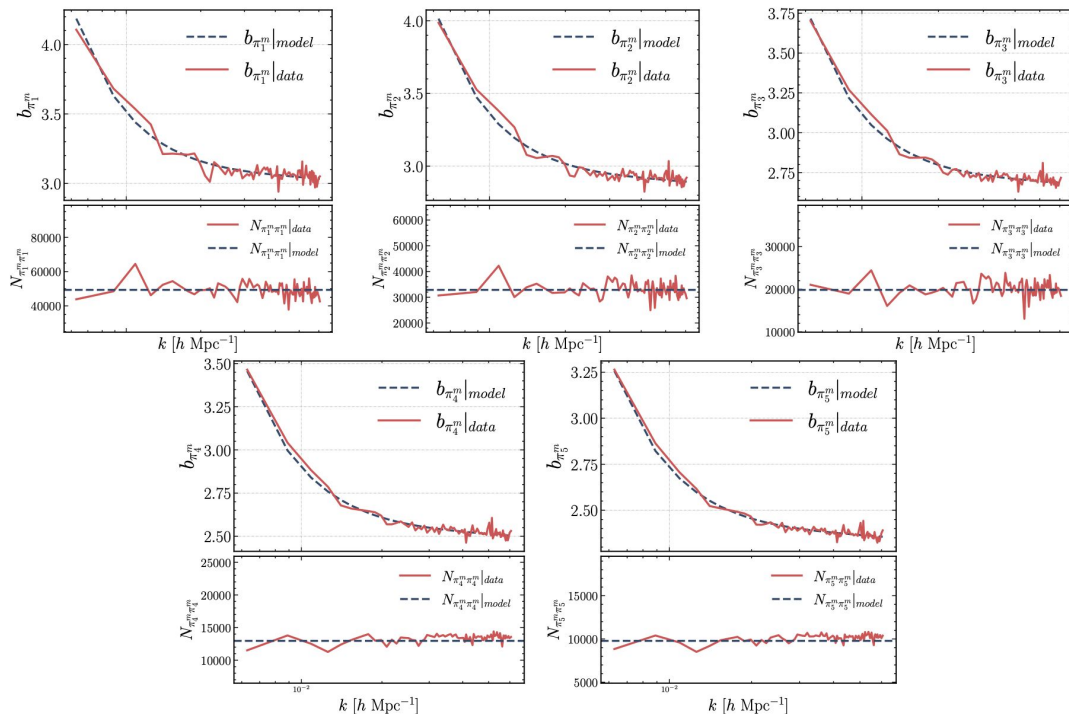
We use an ensemble of **Quijote & Quijote-PNG** (Villaesuca-Navarro et al 2019, Coulton et al 2022 Jung et al 2022) simulations

We generate  $\pi_i^m$  fields using disjoint top-hat filters  $W^i(\mathbf{k})$

$$k \in [(0.5, 1.0), (1.0, 1.5) \dots (2.5, 3.0)]$$

We demonstrate that:

- The large-scale bias shows expected  $1/k^2$  scaling.
- Noise is white as expected.



# Constraining power

We perform likelihood analysis combining  $\delta_m$  with  $\pi_i^m$  derived using disjoint tophat k-space filters in the range  $k \in [(0.5, 1.0), (1.0, 1.5) \dots (2.5, 3.0)]$

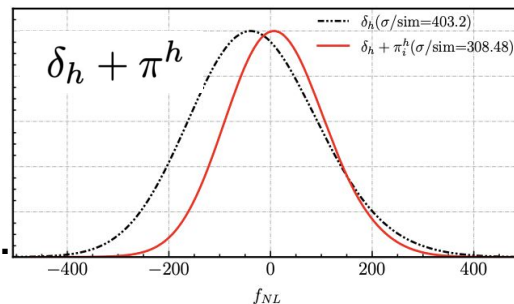
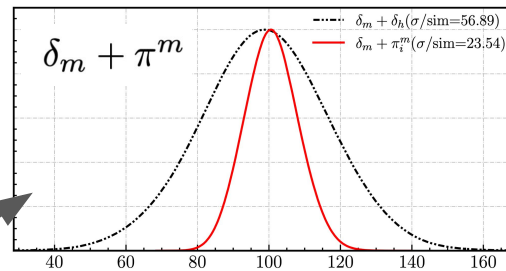
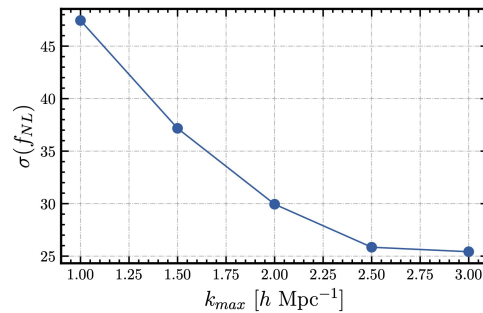
## Relative improvement in $\sigma(f_{NL})$

We fix  $b_{NG_i}$  and marginalize over  $b_{G_i}$  and  $N_{\pi_i \pi_j}$   
 $\delta_m + \pi^m$  constraints are **2.2x** times that of  $\delta_m + \delta_h$

## Constraints improve until very non-linear scales!!

$\delta_h + \pi^h$  gives modest improvements over a halo only analysis.

At  $M_{min}$  of  $\sim 10^{13} M_\odot$ , our non-perturbative analysis only improves by 20%

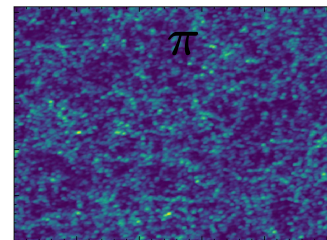
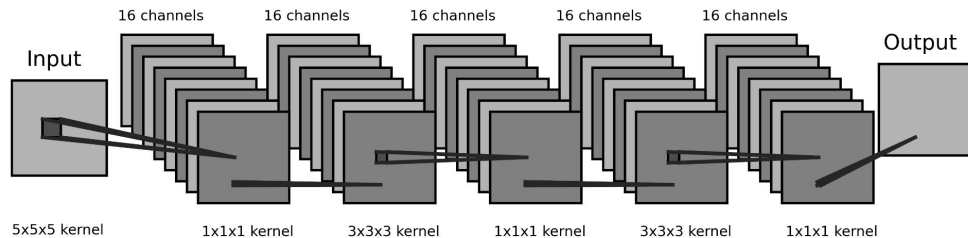
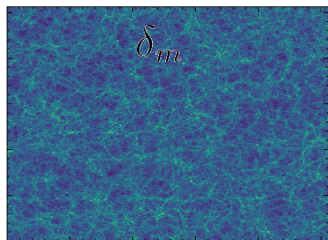


## Part II: A Neural Network enhanced approach:

- Neural Networks (NNs) give very strong constraints on parameters like  $\sigma_8$  and can potentially tap into the higher order information encoded in the density field.
- $\pi^f$  modulation is not optimal, information has leaked into higher-order moments

We design a NN with small receptive field to learn  $\pi^{NN}$  field which locally estimates  $\sigma_8$

$$\mathcal{L} = \left[ \left( \frac{1}{N_{\text{voxels}}} \sum_{\text{voxels } \mathbf{x}} \pi(\mathbf{x}) \right) - \sigma_8^{\text{true}} \right]^2$$

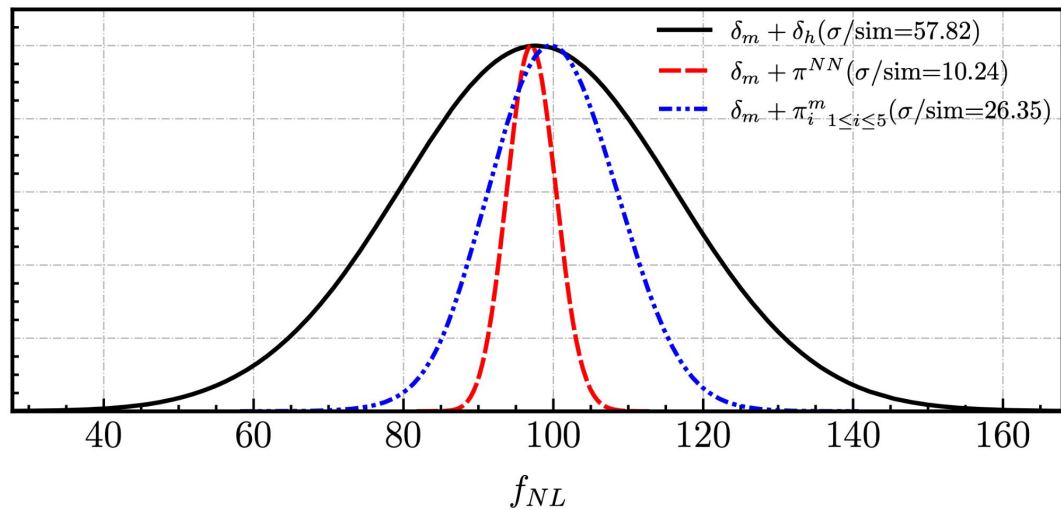


We use Quijote simulations with fixed cosmology but with varying  $\sigma_8$  for training.



# Results:

The  $\delta_m + \pi^{NN}$  based analysis gives a **factor of 5.5** better constraint than  $\delta_m + \delta_h$



# Conclusion:

- We have presented a neat method for extracting information from soft limits of density fields, in particular  $f_{NL}$  information from squeezed bispectrum and collapsed trispectrum.
  - We find significant  $f_{NL}$  sensitivity on *very* non-linear scales\*.
  - Future surveys with **large area + low shot-noise** can exploit higher N-point functions in combination with power spectrum to further tighten constraints on  $f_{NL}$
  - Several interesting applications!
-



# Interpretation and Validation

The bias model for  $\pi$  similar to that of  $\delta_h$

$$\pi^{NN}(\mathbf{k}_L) = b_\pi(\mathbf{k}_L)\delta_m(\mathbf{k}_L) + \epsilon$$

With

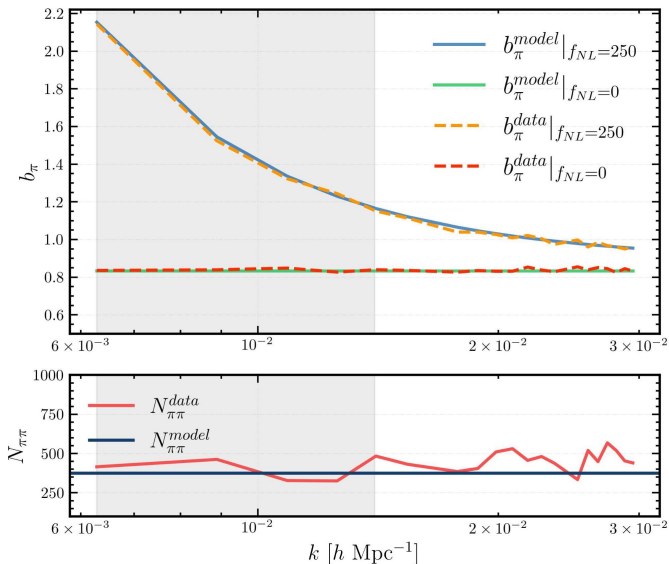
$$b_\pi(k) = b_\pi^G + \boxed{b_\pi^{NG} \frac{f_{NL}}{\alpha(k, z)}}$$

We evaluate this on “**unseen, non-gaussian**” sims

- o Recover  $1/k^2$  scaling, constant noise for  $k \rightarrow 0$
- o Find  $\sim 100\%$  correlation with matter field

**This is not a "black box" approach and is physics informed.** We can do several field level null-tests; cross-correlate with noise maps. Also with other cosmological fields.

**Robust  $1/k^2$  scale dependence, can't be faked!**



# Origin of scale-dependent bias in halos

To motivate our formalism, we first look at origin of scale-dependent halo bias

$f_{NL} = 0$

$f_{NL} \neq 0$

## Large-scale behaviour of $n_h(\mathbf{x})$

$n_h(\mathbf{x}) = F[\delta_m(\mathbf{x})^n]$  modulated only by  $\delta_l$

$n_h(\mathbf{x}) = F[\delta_m(\mathbf{x}), \phi(\mathbf{x})]$   
 modulated by  $\delta_l$  &  $\phi_l (= \frac{\delta_l}{k^2})$

$$\delta_h(\mathbf{k}_L) = b_h(\mathbf{k}_L)\delta_m(\mathbf{k}_L) \quad ; \quad b(\mathbf{k}_L) = \frac{1}{\bar{n}_h} \frac{\partial n_h(\mathbf{x}|f_l)}{\partial f_{\mathbf{k}_l}} = b_\pi + 2\beta \frac{f_{NL}}{k^2}$$

**Main idea: Similar modulation will appear in any field sensitive to small scale power**