

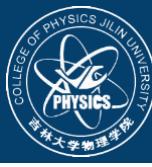
# A pair of novel statistics to improve constraints on primordial non-Gaussianity and cosmological parameters

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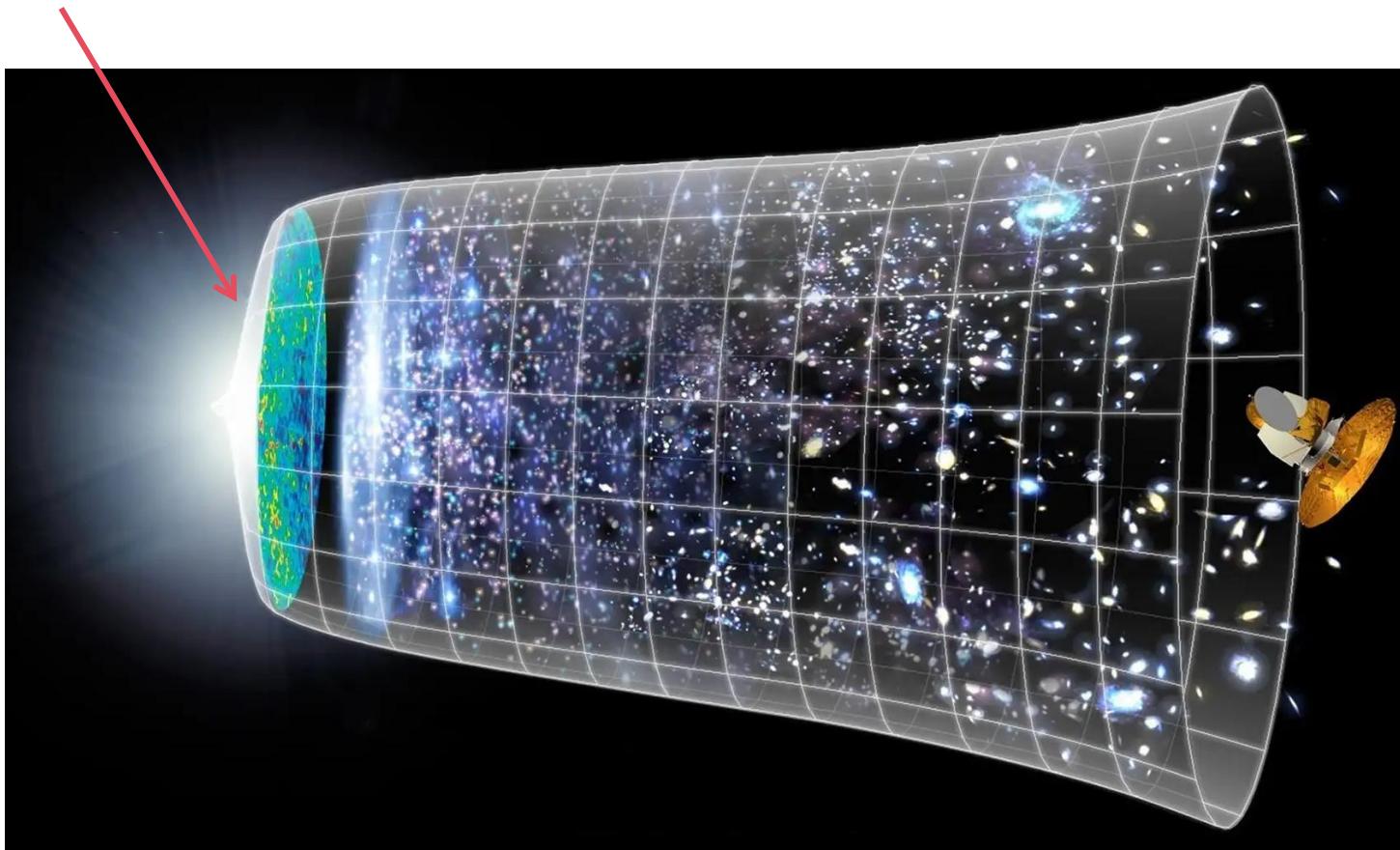


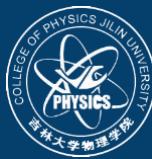
# Outline

- What is primordial non-Gaussianity (PNG)?
- Why is PNG important?
- The challenge of constraining PNG
- Our novel statistics
- Parameter forecasts

# What is Primordial non-Gaussianity (PNG)?

**Cosmic inflation:** the early Universe underwent a phase of accelerated expansion in which quantum fluctuations were stretched at cosmological scales.





# What is Primordial non-Gaussianity (PNG)?

- There exists a broad diversity of inflationary models.
- All existing models predict tiny deviations from Gaussianity of primordial fluctuations, i.e. PNG.
- The primordial gravitational potential  $\Phi$ :

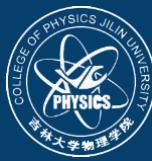
$$\Phi = \Phi_G + f_{NL}^X (\Phi_G^2 - \langle \Phi_G^2 \rangle) + \text{higher-order terms}$$

‘X’ refers to local, equilateral, or orthogonal

- The bispectrum  $B_\Phi(k_1, k_2, k_3)$  is the lowest order statistic sensitive to non-Gaussian features in the primordial potential field.

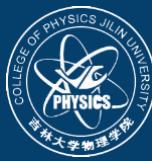
$$\langle \Phi(\mathbf{k}_1)\Phi(\mathbf{k}_2)\Phi(\mathbf{k}_3) \rangle = (2\pi)^3 \delta^D(\mathbf{k}_1 + \mathbf{k}_2 + \mathbf{k}_3) B_\Phi(k_1, k_2, k_3)$$

- Local shape:  $k_1 \ll k_2 \approx k_3$
- Equilateral shape:  $k_1 \approx k_2 \approx k_3$
- orthogonal shape:  $2k_1 \approx 2k_2 \approx k_3$  (negative amplitude)  
 $k_1 \approx k_2 \approx k_3$  (positive amplitude)



# Why is PNG important?

- Discriminating between various inflationary models.
- Providing clues about the high energy physics of the early Universe.
- Bridging the early Universe and late Universe.
- Indicating new physics beyond the standard model of cosmology.
- .....



# The challenge of constraining PNG

- The signal of PNG is faint.
- The most stringent constraints come from measurements of the cosmic microwave background (CMB) anisotropies by the Planck satellite:

$$f_{\text{NL}}^{\text{local}} = -0.9 \pm 5.1, \quad f_{\text{NL}}^{\text{equil}} = -26 \pm 47, \quad f_{\text{NL}}^{\text{ortho}} = -38 \pm 24$$

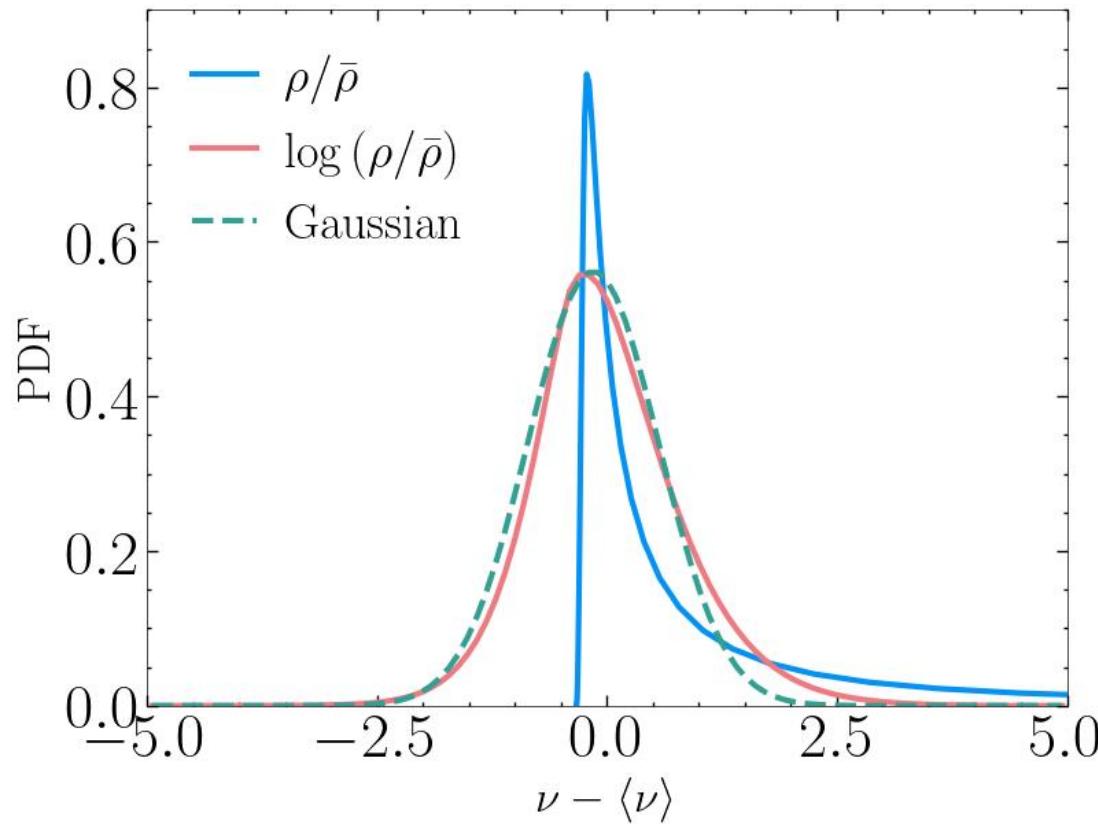
- **Obstacles:** 2D, diffusion damping at small scales
- The current and future large-scale structure (LSS) surveys hold promise for offering enhanced sensitivity to PNG
  - **Advantages:** they can map a huge 3D volume of the Universe
  - **Challenge:** the late-time non-Gaussianity
- Advanced methods:
  - marked power spectrum, power spectra in cosmic web environments, one-point probability distribution, neural network, persistent homology, ...

# Our novel statistics

- The crucial features of the late-time matter distribution:

- the PDF of density field is nearly log-normal

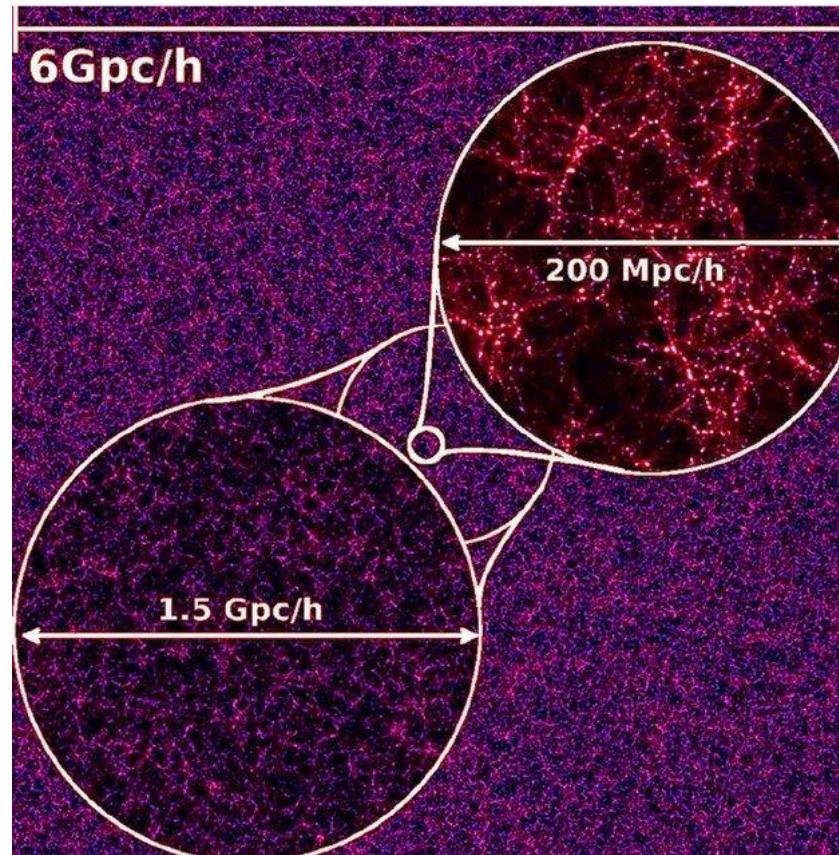
(Hamilton 1985, Coles & Jones 1991,  
Neyrinck et al. 2009, Wang et al. 2011)



# Our novel statistics

- The crucial features of the late-time matter distribution:
  - the density field is manifested in a hierarchical web-like structure

(Sheth & van de Weygaert 2004,  
Sheth 2004, Shen et al. 2006)

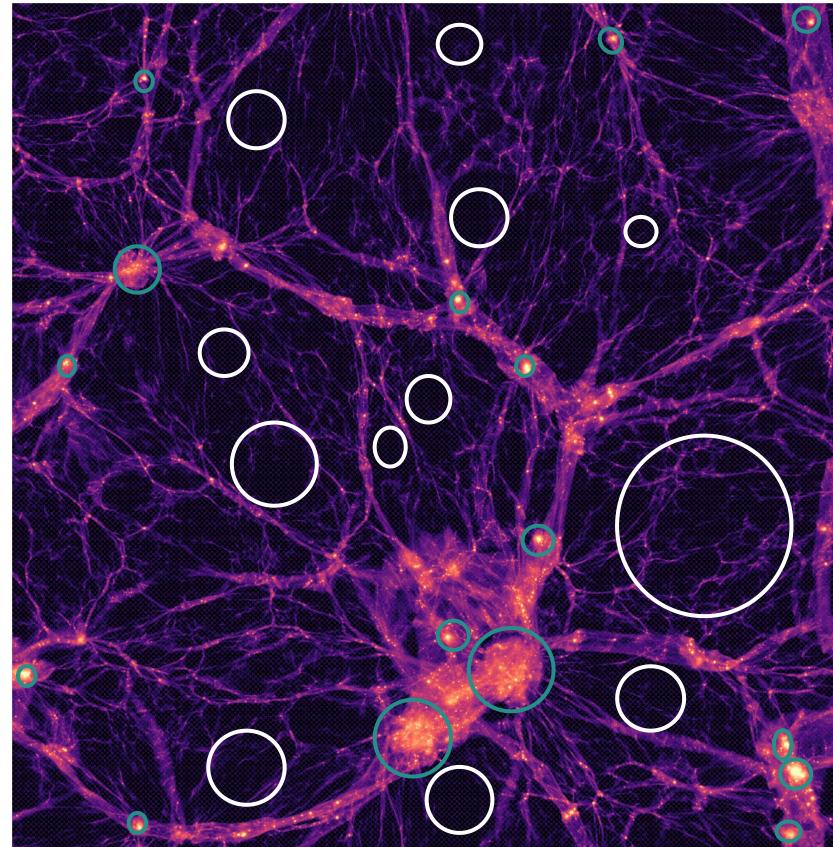


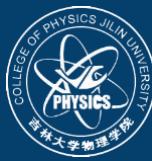
(credit: William A. Watson 2014)

# Our novel statistics

- The crucial features of the late-time matter distribution:
  - the local extrema of density field are particularly sensitive to the PNG

(Dalal et al. 2008, Chan et al. 2019)





# Our novel statistics

- Logarithmic transform

$$\rho_{\ln}(\mathbf{x}) = \ln[\rho(\mathbf{x})/\bar{\rho}]$$

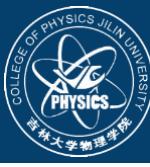
- Continuous wavelet transform (CWT)

$$\tilde{\rho}_{\ln}(w, \mathbf{x}) = \int \rho_{\ln}(\mathbf{x}') \Psi(w, \mathbf{x} - \mathbf{x}') d^3 \mathbf{x}'$$

- the rescaled wavelet:  $\Psi(w, \mathbf{x}) = w^{3/2} \Psi(w \mathbf{x})$
- $w \in \{w_0, w_0 + \Delta w, w_0 + 2\Delta w, \dots, w_0 + i\Delta w, \dots\}$
- the isotropic Gaussian-derived wavelet (GDW):

$$\Psi(\mathbf{x}) = C_N (6 - |\mathbf{x}|^2) e^{-|\mathbf{x}|^2/4}$$

(Wang & He 2021, Wang et al. 2022)



# Our novel statistics

- Detecting peaks (valleys) of  $\tilde{\rho}_{\ln}(w, \mathbf{x})$  by locating coordinates with values above (below) their neighbors.
- The **scale-dependent peak height function** (scale-PKHF) is the number density of CWT peaks with heights falling in the bin  $[\nu_{\text{pk}} - d\nu_{\text{pk}}/2, \nu_{\text{pk}} + d\nu_{\text{pk}}/2]$ :

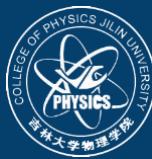
$$n_{\text{pk}}(w, \nu_{\text{pk}}) = \frac{d\mathcal{N}_{\text{pk}}(w)}{d\nu_{\text{pk}}}$$

- The **scale-dependent valley depth function** (scale-VLYDF) is the number density of CWT valleys with depths falling in the bin  $[\nu_{\text{vly}} - d\nu_{\text{vly}}/2, \nu_{\text{vly}} + d\nu_{\text{vly}}/2]$ :

$$n_{\text{vly}}(w, \nu_{\text{vly}}) = \frac{d\mathcal{N}_{\text{vly}}(w)}{d\nu_{\text{vly}}}$$

- The correspondence:  $w = c_w k$ , with  $c_w = 2/\sqrt{7}$  for the isotropic GDW

(Wang & He 2024)



# Parameter forecasts · Fisher analysis

- **Parameter vector:**  $\theta = \{f_{\text{NL}}^{\text{local}}, f_{\text{NL}}^{\text{equil}}, f_{\text{NL}}^{\text{ortho}}, \Omega_m, \Omega_b, \sigma_8, n_s, h\}$
- **Statistic vector:**  $S = \left\{ n_{\text{vly}}(k_0, \nu_{\text{vly},0}), n_{\text{vly}}(k_0, \nu_{\text{vly},1}), n_{\text{vly}}(k_0, \nu_{\text{vly},2}), \dots, n_{\text{vly}}(k_1, \nu_{\text{vly},0}), n_{\text{vly}}(k_1, \nu_{\text{vly},1}), n_{\text{vly}}(k_1, \nu_{\text{vly},2}), \dots, n_{\text{vly}}(k_2, \nu_{\text{vly},0}), n_{\text{vly}}(k_2, \nu_{\text{vly},1}), n_{\text{vly}}(k_2, \nu_{\text{vly},2}), \dots, n_{\text{pk}}(k_0, \nu_{\text{pk},0}), n_{\text{pk}}(k_0, \nu_{\text{pk},1}), n_{\text{pk}}(k_0, \nu_{\text{pk},2}), \dots, n_{\text{pk}}(k_1, \nu_{\text{pk},0}), n_{\text{pk}}(k_1, \nu_{\text{pk},1}), n_{\text{pk}}(k_1, \nu_{\text{pk},2}), \dots, n_{\text{pk}}(k_2, \nu_{\text{pk},0}), n_{\text{pk}}(k_2, \nu_{\text{pk},1}), n_{\text{pk}}(k_2, \nu_{\text{pk},2}), \dots, P(k_0), P(k_1), P(k_2), \dots \right\}$

- The  $1-\sigma$  marginalized error on parameters:

$$\sigma(\theta_i) \geq \sqrt{(\mathcal{F}^{-1})_{ii}}$$

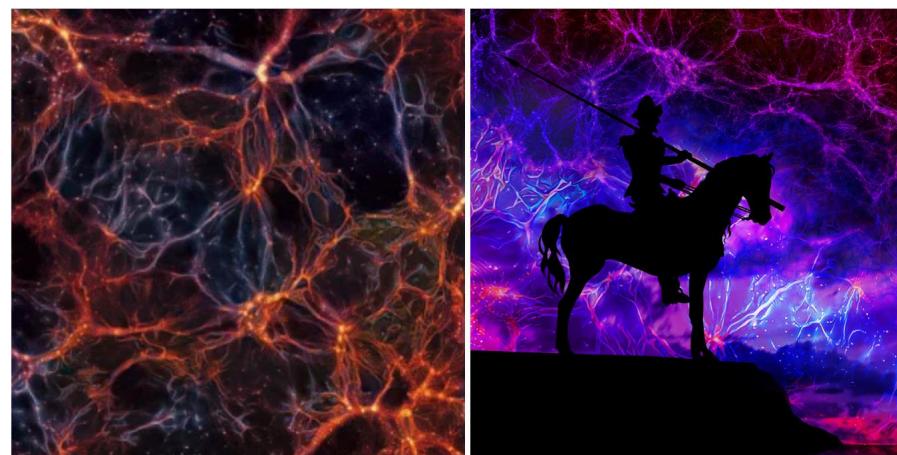
- Fisher matrix:

$$\mathcal{F}_{ij} = \left( \frac{\partial S}{\partial \theta_i} \right) \mathcal{C}^{-1} \left( \frac{\partial S}{\partial \theta_j} \right)^T$$

# Parameter forecasts · Simulations

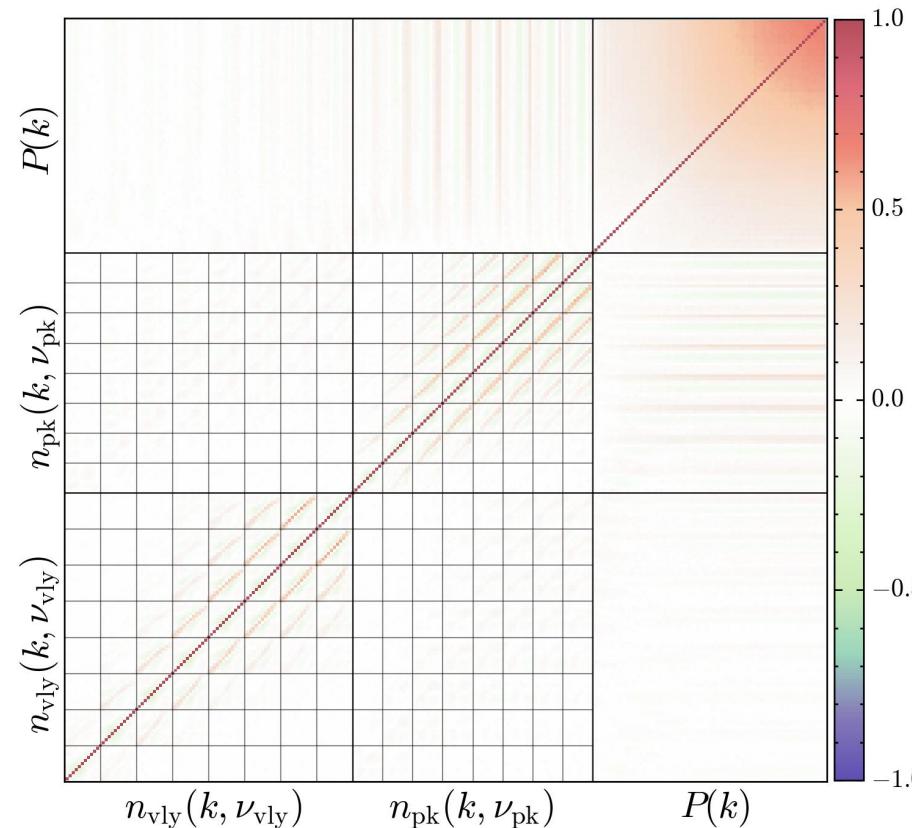
- The Quijote simulation suite: (<https://quijote-simulations.readthedocs.io/en/latest/index.html>)

- Goal: quantify the information content on cosmological observables
  - 15000 fiducial simulations with a Planck cosmology
  - $\{f_{\text{NL}}^{\text{local}} = 0, f_{\text{NL}}^{\text{equil}} = 0, f_{\text{NL}}^{\text{ortho}} = 0, \Omega_m = 0.3175, \Omega_b = 0.049, \sigma_8 = 0.834, n_s = 0.9624, h = 0.6711\}$
  - 5 sets of 500 simulations varying one cosmological parameter
  - 3 sets of 500 simulations with  $f_{\text{NL}} = \pm 100$  (Quijote-PNG)
  - Each simulation box contains  $512^3$  dark matter particles and has a size of 1 Gpc/h



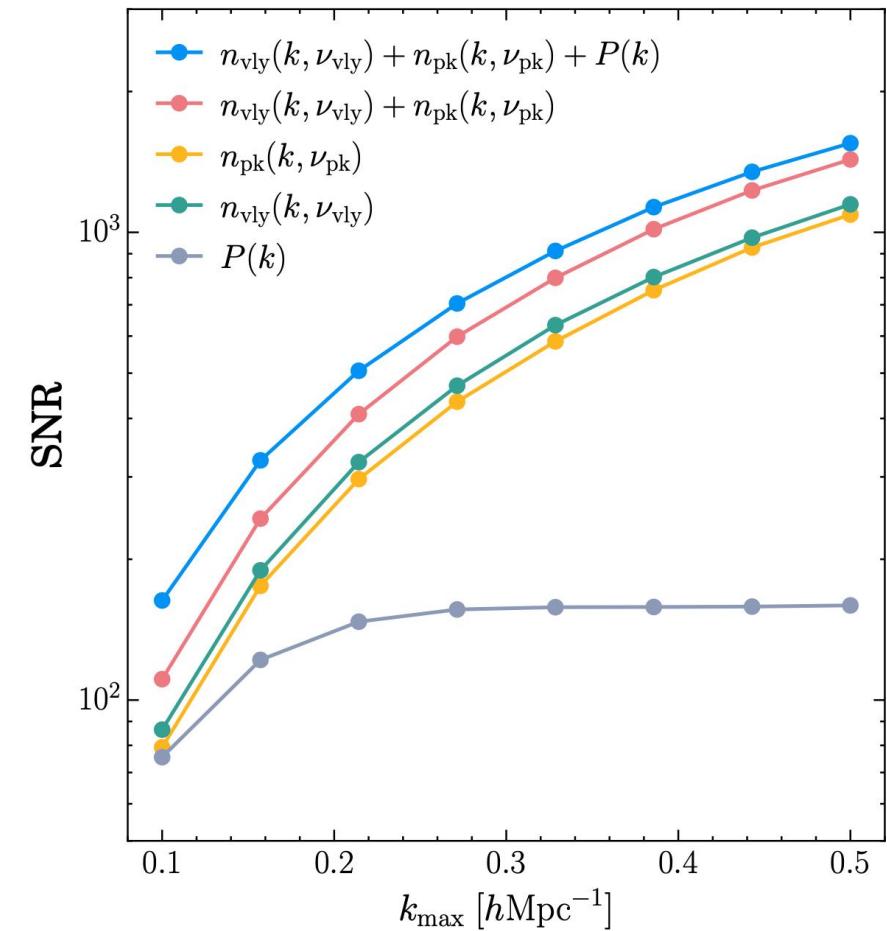
# Parameter forecasts · Results

- The correlation matrix  $r_{ij} = C_{ij}/\sqrt{C_{ii}C_{jj}}$
- The covariances of the scale-PKHF and scale-VLYDF are more diagonalized
- The scale-PKHF, scale-VLYDF and power spectrum are almost uncorrelated with each other



# Parameter forecasts · Results

- The cumulative signal-to-noise ratio (SNR):  $\text{SNR} = \sqrt{\mathbf{S}\mathbf{C}^{-1}\mathbf{S}^T}$
- The scale-PKHF and scale-VLYDF do not show the flattening effect
- The combination of scale-PKHF and scale-VLYDF much higher SNR, up to **8.98** times than the power spectrum at  $k_{\max} = 0.5$  h/Mpc
- even **9.73** times when the power spectrum is included

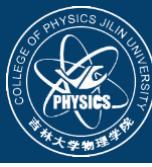


# Parameter forecasts · Results

- Improvement factors of statistics over the power spectrum:  $\sigma_P/\sigma_S$

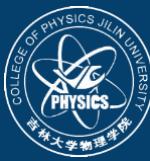
- $f_{\text{NL}}^{\text{local}}$ :  $n_{\text{vly}} + n_{\text{pk}} + P > n_{\text{vly}} + n_{\text{pk}} > P + B > n_{\text{vly}} > B > n_{\text{pk}} > P$
- $f_{\text{NL}}^{\text{equil}}$ :  $n_{\text{vly}} + n_{\text{pk}} + P > n_{\text{vly}} + n_{\text{pk}} > P + B > B > n_{\text{vly}} > n_{\text{pk}} > P$
- $f_{\text{NL}}^{\text{ortho}}$ :  $n_{\text{vly}} + n_{\text{pk}} + P > n_{\text{vly}} + n_{\text{pk}} > P + B > B > n_{\text{vly}} > n_{\text{pk}} > P$
- $\Omega_m$ :  $n_{\text{vly}} + n_{\text{pk}} + P > P + B > B > n_{\text{vly}} + n_{\text{pk}} > P > n_{\text{vly}} > n_{\text{pk}}$
- $\Omega_b$ :  $n_{\text{vly}} + n_{\text{pk}} + P > P + B > B > n_{\text{vly}} + n_{\text{pk}} > n_{\text{vly}} > n_{\text{pk}} > P$
- $\sigma_8$ :  $n_{\text{vly}} + n_{\text{pk}} + P > P + B > B > n_{\text{vly}} + n_{\text{pk}} > n_{\text{vly}} = n_{\text{pk}} > P$
- $n_s$ :  $n_{\text{vly}} + n_{\text{pk}} + P > n_{\text{vly}} + n_{\text{pk}} > n_{\text{vly}} > n_{\text{pk}} > P + B > B > P$
- $h$ :  $n_{\text{vly}} + n_{\text{pk}} + P > P + B > B > n_{\text{vly}} + n_{\text{pk}} > n_{\text{vly}} > n_{\text{pk}} > P$

Paras	$\sigma_P/\sigma_B$ [19]	$\sigma_P/\sigma_{P+B}$ [19]	$\sigma_P/\sigma_{n_{\text{vly}}}$	$\sigma_P/\sigma_{n_{\text{pk}}}$	$\sigma_P/\sigma_{n_{\text{vly}}+n_{\text{pk}}}$	$\sigma_P/\sigma_{n_{\text{vly}}+n_{\text{pk}}+P}$
$f_{\text{NL}}^{\text{local}}$	28.6	57.6	32.7	20.2	60.2	99.1
$f_{\text{NL}}^{\text{equil}}$	45.1	53.3	28.1	19.5	54.8	116.0
$f_{\text{NL}}^{\text{ortho}}$	43.5	74.9	29.8	39.3	104.4	112.4
$\Omega_m$	2.5	5.1	0.8	0.7	1.2	5.9
$\Omega_b$	2.4	3.8	1.2	1.1	1.6	4.0
$\sigma_8$	10.1	29.9	4.2	4.2	8.8	48.5
$n_s$	3.2	7.8	12.4	8.6	15.6	22.8
$h$	2.6	4.9	1.7	1.5	2.4	6.6



# Summary and Outlook

- We introduce two new summary statistics, the scale-PKHF and scale-VLYDF, for constraining PNG
- The scale-PKHF and scale-VLYDF are capable of capturing a wealth of primordial information about the Universe
- The scale-PKHF and scale-VLYDF are complementary to the traditional power spectrum
- Combining them two with the power spectrum improve constraints on all parameters compared to the bispectrum and power spectrum combination
- Our methodology is well-suited for future surveys
- Further research is required
  - theoretical modeling of the scale-PKHF and scale-VLYDF
  - comparing them with other advanced statistics
  - dealing with redshift space distortions
  - investigating the effects of tracer bias



arXiv:2408.13876

# Capturing primordial non-Gaussian signatures in the late Universe by multi-scale extrema of the cosmic log-density field

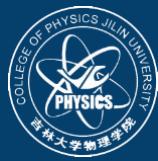
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(Dated: August 27, 2024)

We construct two new summary statistics, the scale-dependent peak height function (scale-PKHF) and the scale-dependent valley depth function (scale-VLYDF), and forecast their constraining power on PNG amplitudes  $\{f_{\text{NL}}^{\text{local}}, f_{\text{NL}}^{\text{equil}}, f_{\text{NL}}^{\text{ortho}}\}$  and standard cosmological parameters based on ten thousands of density fields drawn from QUIJOTE and QUIJOTE-PNG simulations at  $z = 0$ . With the Fisher analysis, we find that the scale-PKHF and scale-VLYDF are capable of capturing a wealth of primordial information about the Universe. Specifically, the constraint on the scalar spectral index  $n_s$  obtained from the scale-VLYDF (scale-PKHF) is 12.4 (8.6) times tighter than that from the power spectrum, and 3.9 (2.7) times tighter than that from the bispectrum. The combination of the two statistics yields constraints on  $\{f_{\text{NL}}^{\text{local}}, f_{\text{NL}}^{\text{equil}}\}$  similar to those from the bispectrum and power spectrum combination, but provides a 1.4-fold improvement in the constraint on  $f_{\text{NL}}^{\text{ortho}}$ . After including the power spectrum, its constraining power well exceeds that of the bispectrum and power spectrum combination by factors of 1.1–2.9 for all parameters.



Welcome any comments!

