

Scale-dependent bias with primordial non-Gaussianities

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collab. with Taka Matsubara (KMI, Nagoya Univ.), ...

SY and Matsubara, arXiv:1210.2495

and also SY, JCAP 1111(2011)001

Gong and SY, MNRAS 417(2011)L79

Congratulations on 60th birthday, Sasaki-san!

Thanks for encouragement and useful discussion in my PhD days (my “delta N” life)!

Contents

- Local-type primordial non-Gaussianity
 - fNL, gNL, tauNL,...
- Formulation for the scale-dependent bias
- Stochastic bias (related with fNL and tauNL)
- Higher order corrections
- Summary

Simplest parameterization

• Focusing on Local type non-Gaussianity

(Komatsu & Spergel(2001), Byrnes, Sasaki & Wands(2006), ...)

primordial curvature fluctuations

$$\Phi = \underbrace{\Phi_G}_{\text{Gaussian fluc.}} + \underbrace{f_{\text{NL}}}_{\sim 10^{-5}} \left(\Phi_G^2 - \langle \Phi^2 \rangle \right) + \underbrace{g_{\text{NL}}}_{\text{non-linearity parameters}} \Phi_G^3 + \dots$$

Gaussian fluc.

$\sim 10^{-5}$

non-linearity parameters



Non-zero higher order spectra

(higher order correlation functions)

leadingly, ...

• Bispectrum (3-point corr. func.) $\longleftrightarrow \langle \Phi(\mathbf{k}_1)\Phi(\mathbf{k}_2)\Phi(\mathbf{k}_3) \rangle$

$$B_\Phi(k_1, k_2, k_3) = 2f_{\text{NL}} [P_\Phi(k_1)P_\Phi(k_2) + P_\Phi(k_2)P_\Phi(k_3) + P_\Phi(k_3)P_\Phi(k_1)]$$

• Trispectrum (4-point corr. func.) $\longleftrightarrow \langle \Phi(\mathbf{k}_1)\Phi(\mathbf{k}_2)\Phi(\mathbf{k}_3)\Phi(\mathbf{k}_4) \rangle_c$

Importance of trispctrum

•Trispectrum

(Byrnes, Sasaki & Wands(2006), Boubekur & Lyth(2006) ...)

$$\langle \Phi_{\mathbf{k}_1} \Phi_{\mathbf{k}_2} \Phi_{\mathbf{k}_3} \Phi_{\mathbf{k}_4} \rangle_c = (2\pi)^3 \delta^{(3)}(\mathbf{k}_1 + \mathbf{k}_2 + \mathbf{k}_3 + \mathbf{k}_4) T_\Phi(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3, \mathbf{k}_4)$$

$$T_\Phi(k_1, k_2, k_3, k_4) = 6g_{\text{NL}} [P_\Phi(k_1)P_\Phi(k_2) + 2 \text{ perms.}] \\ + \frac{25}{9} \tau_{\text{NL}} \{P_\Phi(k_1) [P_\Phi(k_{13}) + P_\Phi(k_{23})] + 10 \text{ perms.}\}$$

need two terms (different momentum-dependence)

Suyama and Yamaguchi (2008), ...

We can generalize ...

e.g.)

$$\Phi = \phi_G + \psi_G + f_{\text{NL}} (\phi_G^2 - \langle \phi_G^2 \rangle)$$

$$\langle \phi_G \psi_G \rangle = 0$$

$$R \equiv P_\phi / P_\psi$$

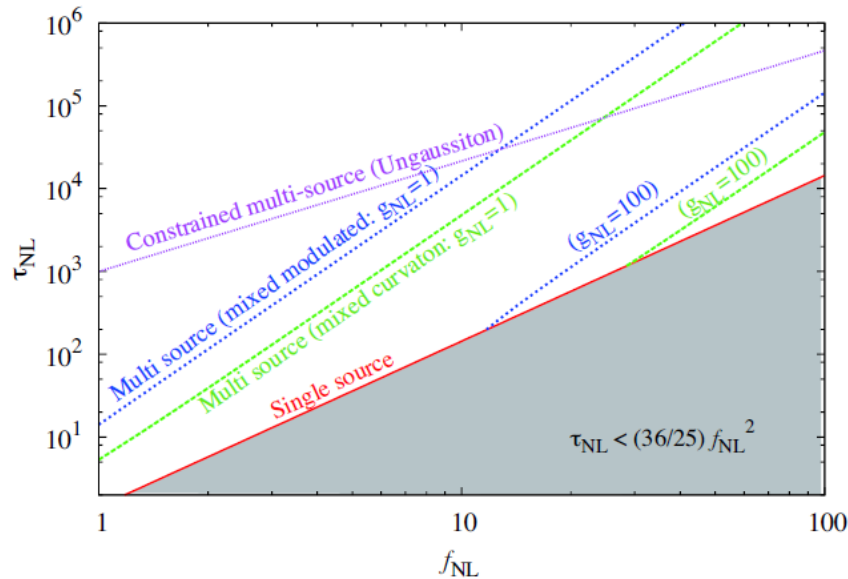


$$\tau_{\text{NL}} = \left(\frac{1+R}{R} \right) \left(\frac{6}{5} f_{\text{NL}} \right)^2$$

$$\tau_{\text{NL}} \geq \left(\frac{6}{5} f_{\text{NL}} \right)^2$$

(Suyama, Takahashi, Yamguchi and SY (2010))

f_{NL} vs τ_{NL}

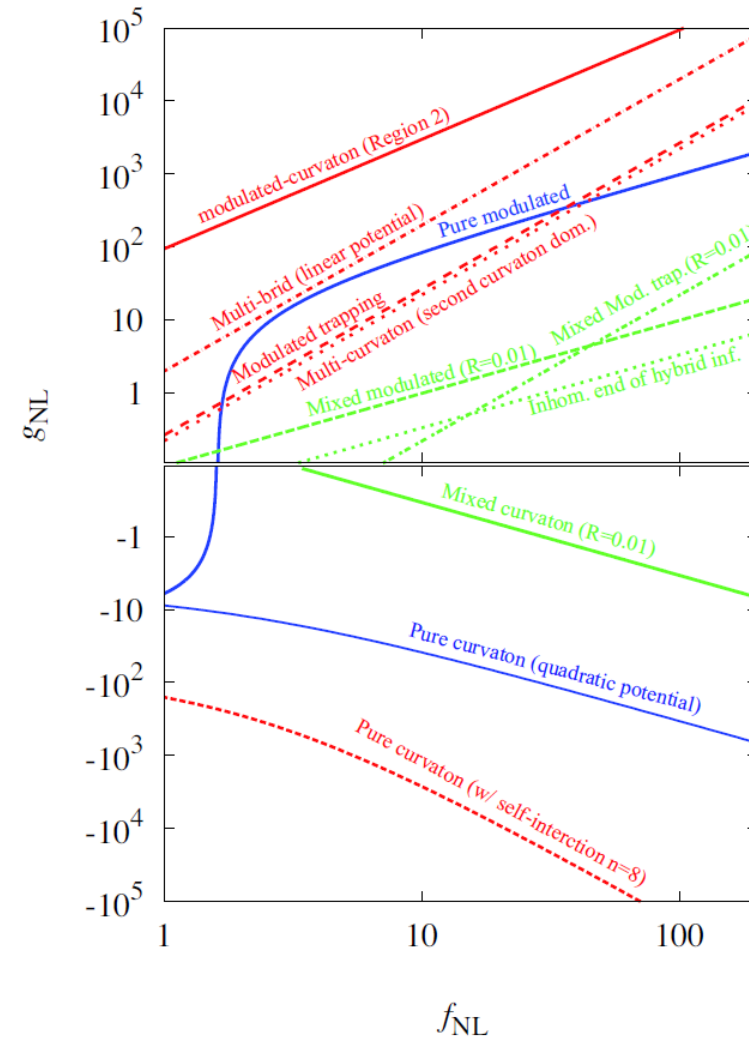


Different lines represent different models



distinguishing models !!!

f_{NL} vs g_{NL}



How accuracy can we measure these parameters with using cosmological observations?

Observational constraints

- From CMB observations (CMB trispectrum)

Smidt et al.(2010), Fergusson et al.(2010)

$$-7.4 \times 10^5 < g_{\text{NL}} < 8.2 \times 10^5$$

$$-0.6 \times 10^4 < \tau_{\text{NL}} < 3.3 \times 10^4$$

95%CL. with WMAP data

- Large scale structure?

How can we see these non-Gaussian effects ?

Structure formation in non-Gaussian universe

- scale-dependent bias
with primordial non-Gaussianity -

Simple picture

- Halo bias **bias = P_h / P_m**

- **matter density fluctuations** (matter power spectrum; $P_m(k)$)



- **Halo number density fluctuations** (halo power spectrum; $P_h(k)$)
(peaky objects)

Gaussian + linear → **linear (constant) bias**

Kaiser (1984), Bardeen, Bond, Kaiser and Szalay (1986), Sheth, Mo and Thormen(2001), ...



non-Gaussian case ? → **scale-dependent bias**

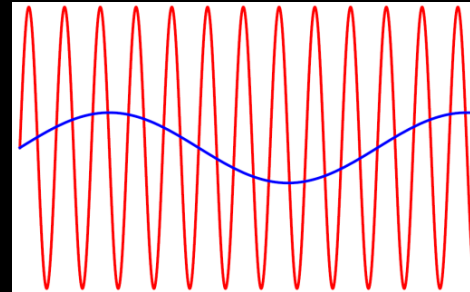
Dalal et al.(2008), Slosar et al.(2008), ...

Simple picture

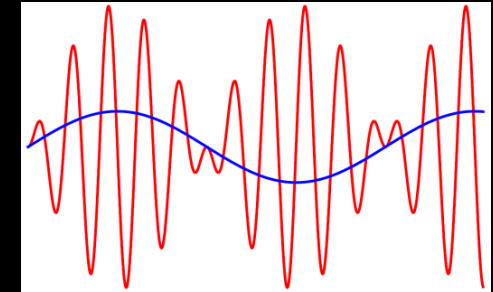
$$\Phi = \Phi_G + f_{\text{NL}} \Phi_G^2$$

decomposition;

$$\Phi_G = \Phi_{\text{long}} + \Phi_{\text{short}}$$



Gaussian

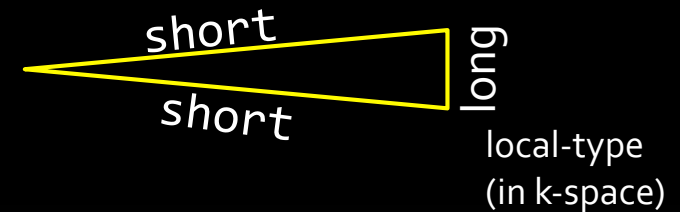


non-Gaussian

due to the non-Gaussianity
 → long-short mode coupling !

Blue; long mode (background)
 Red; short mode (peak)

- collapsed objects ↔ short mode fluctuations



$$\delta_{\text{short}} = \delta_G (1 + 2f_{\text{NL}} \Phi_{\text{long}}(x))$$

$$\sigma_{\text{short}}^2 \equiv \langle \delta_{\text{short}}^2 \rangle ; \text{variance}$$

$$\Phi_{\text{long}} \propto \delta_{\text{long}} / k^2$$

$$\Phi \rightarrow \delta$$

Poisson eq. $\Delta \Phi = 4\pi G \rho \delta$

→ modulation of the variance of the short mode !!

Through the local-type non-Gaussianity, the formation process of the objects can be modified → scale-dependence of the bias

Toward precision cosmology

need more precise theoretical prediction

- not so high-peak objects
- long- and short- decomposition is right?
- GR effects? (Roy Maartens-san's talk)



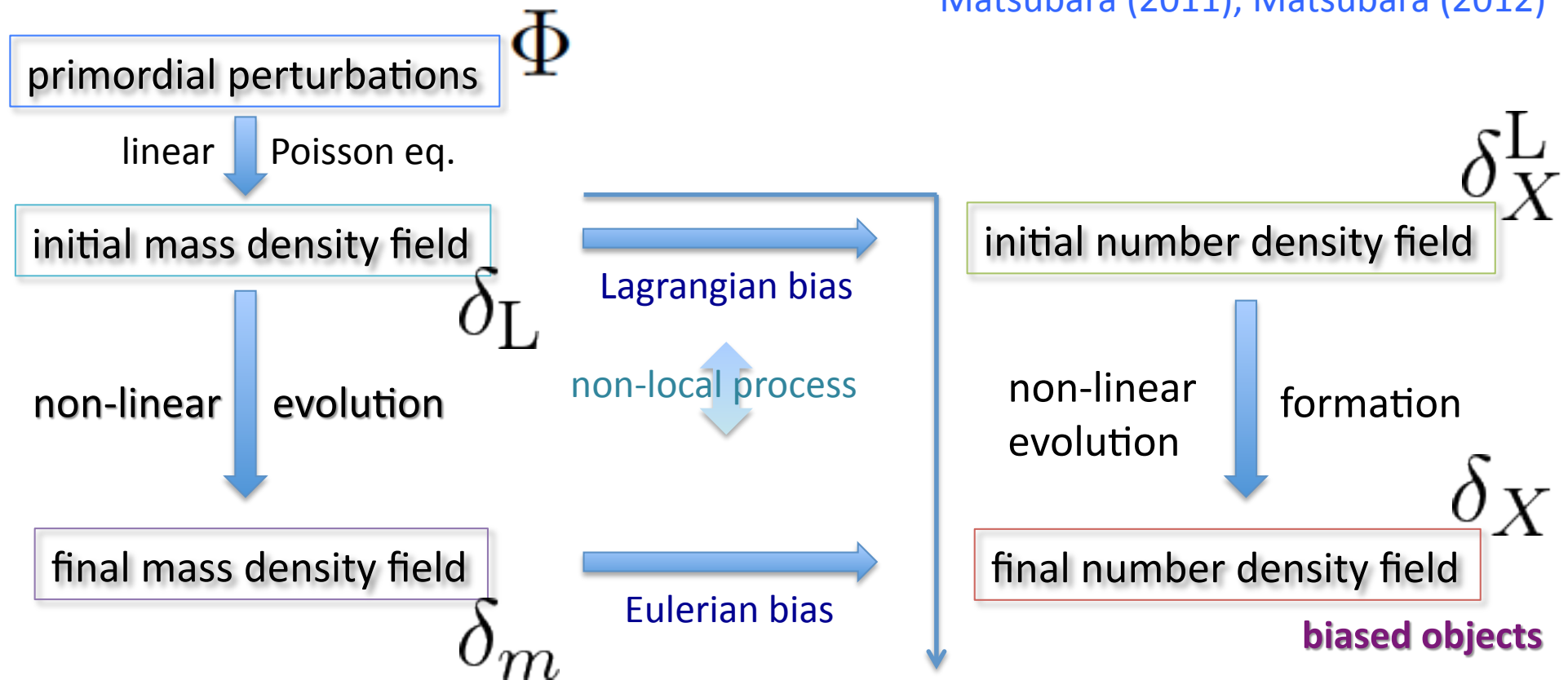
- using N-body simulation
- deriving accurate analytic formula

e.g., Desjacques + (2011) → mass dependence, scale-independent corrections

Scoccimarro + (2012)

Integrated Perturbation Theory (iPT)

Matsubara (2011), Matsubara (2012)



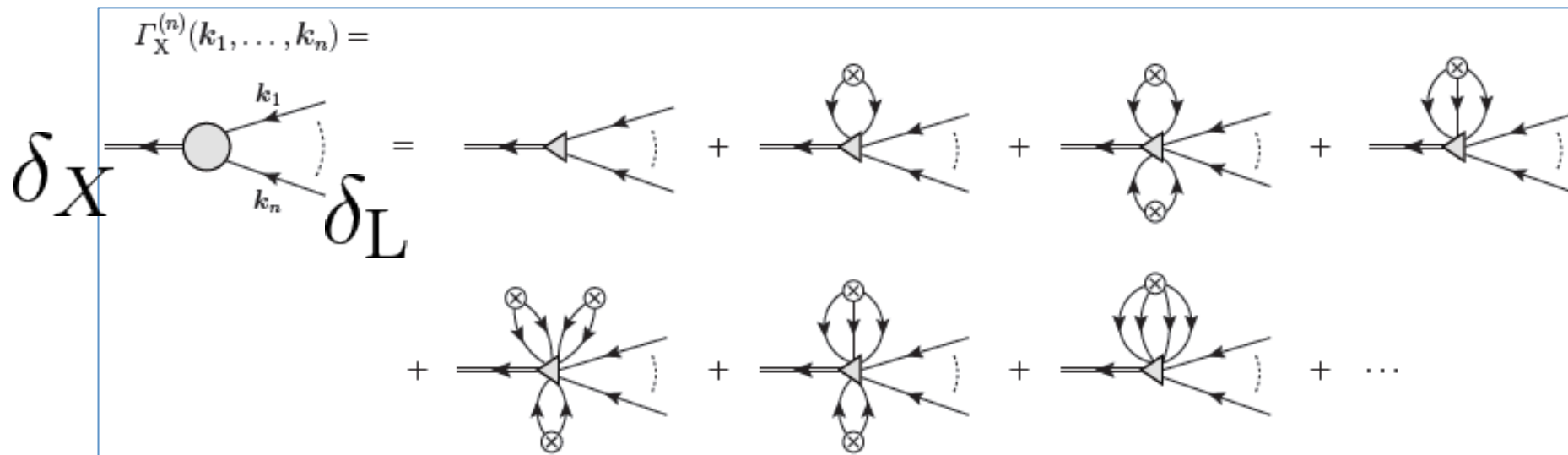
Non-linear perturbation theory integrated with non-local bias, redshift-distortions, and primordial non-Gaussianity

without high peak limit and peak-background split picture

Integrated Perturbation Theory (iPT)

- Introducing multi-point propagators

$$\left\langle \frac{\delta^n \delta_X(\mathbf{k})}{\delta \delta_L(\mathbf{k}_1) \cdots \delta \delta_L(\mathbf{k}_n)} \right\rangle = (2\pi)^{3-3n} \delta^{(3)}(\mathbf{k}_1 + \cdots + \mathbf{k}_n - \mathbf{k}) \Gamma_X^{(n)}(\mathbf{k}_1, \cdots, \mathbf{k}_n),$$



➔ response of the final number density field to the initial mass density field

Power spectrum of the biased objects

Up to the leading order contributions from fNL, gNL and tauNL

$$P_X(k) = \left[\Gamma_X^{(1)}(\mathbf{k}) \right]^2 P_L(k) \quad \text{---} \circ \text{---} \otimes \text{---} \circ \text{---} \quad \text{Gaussian case}$$

$$+ \Gamma_X^{(1)}(\mathbf{k}) \int \frac{d^3 p}{(2\pi)^3} \Gamma_X^{(2)}(\mathbf{p}, \mathbf{k} - \mathbf{p}) B_L(\mathbf{k}, -\mathbf{p}, -\mathbf{k} + \mathbf{p}) \quad \text{---} \circ \text{---} \otimes \text{---} \circ \text{---}$$

\leftrightarrow fNL

$$+ \frac{1}{3} \Gamma_X^{(1)}(\mathbf{k}) \int \frac{d^3 p_1 d^3 p_2}{(2\pi)^6} \Gamma_X^{(3)}(\mathbf{p}_1, \mathbf{p}_2, \mathbf{k} - \mathbf{p}_1 - \mathbf{p}_2) T_L(\mathbf{k}, -\mathbf{p}_1, -\mathbf{p}_2, -\mathbf{k} + \mathbf{p}_1 + \mathbf{p}_2)$$

\leftrightarrow gNL

$$+ \frac{1}{4} \int \frac{d^3 p_1 d^3 p_2}{(2\pi)^6} \Gamma_X^{(2)}(\mathbf{p}_1, \mathbf{k} - \mathbf{p}_1) \Gamma_X^{(2)}(-\mathbf{p}_2, -\mathbf{k} + \mathbf{p}_2) T_L(\mathbf{p}_1, \mathbf{k} - \mathbf{p}_1, -\mathbf{p}_2, -\mathbf{k} + \mathbf{p}_2)$$

\leftrightarrow tauNL

We can also consider the higher order effects order by order.

$$\Gamma_X^{(n)}(\mathbf{k}_1, \dots, \mathbf{k}_n)$$

- Multi-point propagators depend on
 - non-linear evolution of the matter density field (initial field \leftrightarrow final field)
 - non-linear bias function (matter density field \leftrightarrow number density field)
- Introducing renormalized bias functions

$$c_n^L(\mathbf{k}_1, \mathbf{k}_2, \dots, \mathbf{k}_n) = (2\pi)^{3n} \int \frac{d^3p}{(2\pi)^3} \left\langle \frac{\delta^n \delta_X^L(\mathbf{p})}{\delta\delta_L(\mathbf{k}_1) \dots \delta\delta_L(\mathbf{k}_n)} \right\rangle$$

\leftarrow can be written in terms of the standard Lagrangian local bias; b_n^L
(obtained from the mass function)

$$\Gamma_X^{(1)}(k) = 1 + c_1^L(k)$$

\rightarrow $\Gamma_X^{(2)}(k_1, k_2) = \underline{F_2(k_1, k_2)} + \left(1 + \frac{\mathbf{k}_1 \cdot \mathbf{k}_2}{k_1^2}\right) c_1^L(k_2) + \left(1 + \frac{\mathbf{k}_1 \cdot \mathbf{k}_2}{k_2^2}\right) c_1^L(k_1) + c_2^L(k_1, k_2)$

non-linear evolution of the matter density field;

$$F_n(\mathbf{k}_1, \mathbf{k}_2, \dots, \mathbf{k}_n) \rightarrow 0$$

$$\mathbf{k}_1 + \mathbf{k}_2 + \dots + \mathbf{k}_n \rightarrow 0 \quad (\text{on large scales})$$

Scale-dependent bias with fNL, gNL and tauNL

- Bias parameter

$$P_X(k) \equiv b_X(k)P_L(k)$$

on large scales ($k \rightarrow 0$), $b_X(k) \approx b_1(M)^2 + \Delta b(k)$

Scale-dependent part;

$$\begin{aligned} \Delta b(k) \approx & \underline{4f_{\text{NL}}} \frac{b_1(M)}{\mathcal{M}(k)} \left[b_2^{\text{L}}(M) + 2 \frac{1 + \delta_c(b_1(M) - 1)}{\delta_c^2} \right] \sigma_M^2 \\ & + \left(\underline{g_{\text{NL}}} + \frac{25}{27} \tau_{\text{NL}} \right) \frac{b_1(M)}{\mathcal{M}(k)} \\ & \times \left[b_3^{\text{L}}(M) + \frac{2 + 2\delta_c(b_1(M) - 1) + \delta_c^2 b_2^{\text{L}}(M)}{\delta_c^3} \left(4 + \frac{d \ln S_3(M)}{\ln \sigma_M} \right) \right] \sigma_M^4 S_3(M) \\ & + \frac{25}{9} \underline{\tau_{\text{NL}}} \frac{1}{\mathcal{M}(k)^2} \left[b_2^{\text{L}}(M) + 2 \frac{1 + \delta_c(b_1(M) - 1)}{\delta_c^2} \right]^2 \sigma_M^4, \end{aligned}$$

$$S_3(M) = \frac{\langle \delta_M^3 \rangle}{\langle \delta_M^2 \rangle^2} \quad \delta_c = 1.68 \quad b_1(M) = 1 + b_1^{\text{L}}(M)$$

Scale-dependent bias with fNL, gNL and tauNL

- Bias parameter

$$P_X(k) \equiv b_X(k) P_L(k)$$

on large scales ($k \rightarrow 0$), $b_X(k) \approx b_1(M)^2 + \Delta b(k)$

Scale-dependent part;

$$\Delta b(k) \approx 4f_{\text{NL}} \frac{b_1(M)}{\mathcal{M}(k)} \left[b_2^L(M) + 2 \frac{1 + \delta_c(b_1(M) - 1)}{\delta_c^2} \right] \sigma_M^2 \longrightarrow 1/k^2 \text{ - dependence}$$

$$\delta_L = \mathcal{M}(k) \Phi(k), \quad \mathcal{M}(k) \propto k^2$$

from Poisson eq.

Large enhancement on large scales ($k \rightarrow 0$)

Scale-dependent bias with fNL, gNL and tauNL

- Bias parameter

$$P_X(k) \equiv b_X(k) P_L(k)$$

on large scales ($k \rightarrow 0$), $b_X(k) \approx b_1(M)^2 + \Delta b(k)$

Scale-dependent part;

$$\Delta b(k) \approx$$

$$+ \left(g_{\text{NL}} + \frac{25}{27} \tau_{\text{NL}} \right) \frac{b_1(M)}{\mathcal{M}(k)} \longrightarrow 1/k^2 \text{ - dependence}$$

$$\times \left[b_3^{\text{L}}(M) + \frac{2 + 2\delta_c(b_1(M) - 1) + \delta_c^2 b_2^{\text{L}}(M)}{\delta_c^3} \left(4 + \frac{d \ln S_3(M)}{\ln \sigma_M} \right) \right] \sigma_M^4 S_3(M)$$

Scale-dependent bias with fNL, gNL and tauNL

- Bias parameter

$$P_X(k) \equiv b_X(k) P_L(k)$$

on large scales ($k \rightarrow 0$), $b_X(k) \approx b_1(M)^2 + \Delta b(k)$

Scale-dependent part;

$$\Delta b(k) \approx$$

$$+ \frac{25}{9} \tau_{\text{NL}} \frac{1}{\mathcal{M}(k)^2} \left[b_2^L(M) + 2 \frac{1 + \delta_c (b_1(M) - 1)}{\delta_c^2} \right]^2 \sigma_M^4,$$

→ 1/k⁴ - dependence

Higher order local bias parameters-dependence

- Bias parameter

$$P_X(k) \equiv b_X(k)P_L(k)$$

on large scales ($k \rightarrow 0$), $b_X(k) \approx b_1(M)^2 + \Delta b(k)$

Scale-dependent part;

$$\Delta b(k) \approx 4f_{\text{NL}} \frac{b_1(M)}{\mathcal{M}(k)} \left[b_2^{\text{L}}(M) + 2 \frac{1 + \delta_c(b_1(M) - 1)}{\delta_c^2} \right] \sigma_M^2 \longrightarrow k^{-2}\text{-dependence}$$

$$+ \left(g_{\text{NL}} + \frac{25}{27} \tau_{\text{NL}} \right) \frac{b_1(M)}{\mathcal{M}(k)}$$

$$\times \left[b_3^{\text{L}}(M) + \frac{2 + 2}{\mathcal{M}(k)} \right]$$

$$+ \frac{25}{9} \tau_{\text{NL}} \frac{1}{\mathcal{M}(k)^2} \left[b_2^{\text{L}}(M) \right]$$

In general, we need to know higher order Lagrangian local bias parameters !!
(c.f. for Press-Schechter mass function, b_n can be written in b_{n-1} ;

well-known result; $\Delta b = 4f_{\text{NL}} \frac{(b_1^{\text{L}} + 1)b_1^{\text{L}}}{\mathcal{M}(k)}$

$$S_3(M) = \frac{\langle \delta_M^3 \rangle}{\langle \delta_M^2 \rangle^2} \quad \delta_c = 1.68 \quad b_1$$

Stochastic bias

S-Y inequality in the structure formation

$$\tau_{\text{NL}} \geq \left(\frac{6}{5} f_{\text{NL}} \right)^2$$

- fNL vs tauNL ; inequality $\tau_{\text{NL}} \geq \left(\frac{6}{5} f_{\text{NL}}\right)^2$

Introducing a **stochasticity parameter**;

$$r(k) \equiv \frac{P_m(k)P_X(k)}{P_{mX}(k)^2}$$

$P_{mX}(k)$; matter density field – biased objects
cross power spectrum

$P_m(k)$; matter density field
power spectrum

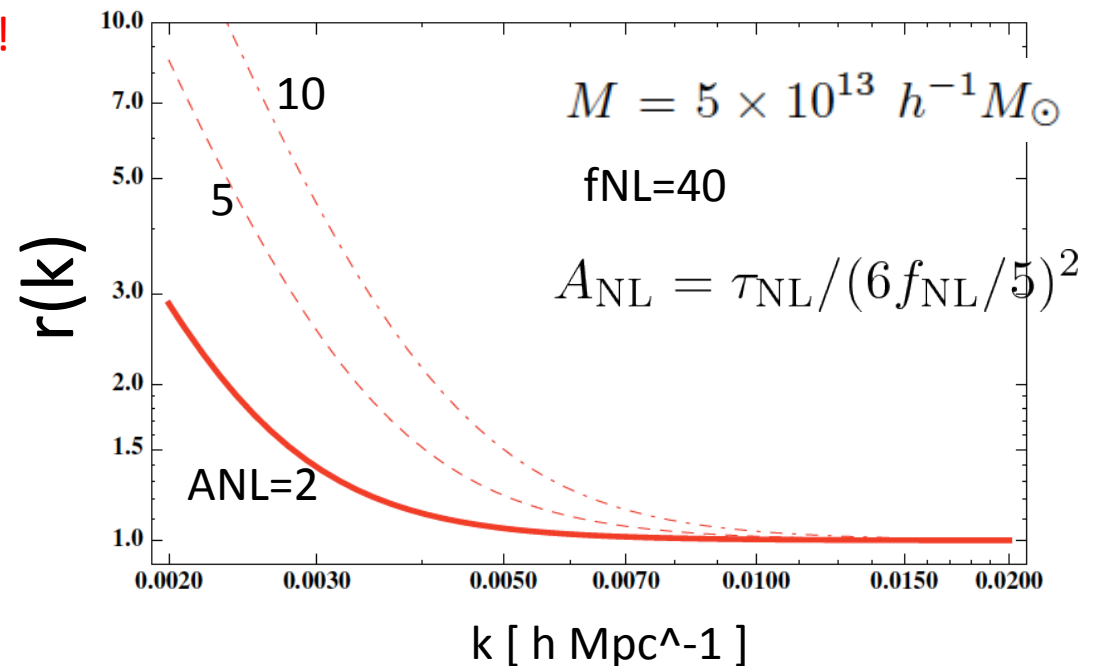
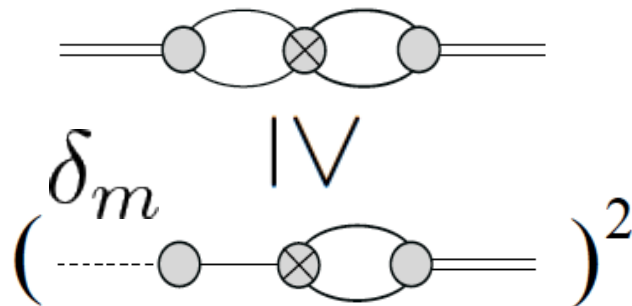
On large scales,

$$r(k) \simeq 1 + \left(\frac{25}{9} \tau_{\text{NL}} - 4f_{\text{NL}}^2\right) \frac{1}{b_1(k)^2 \mathcal{M}(k)^2} \left[\int \frac{d^3p}{(2\pi)^3} c_2^L(\mathbf{p}, -\mathbf{p}) P_L(p) \right]^2$$

Directly dependent on the inequality !

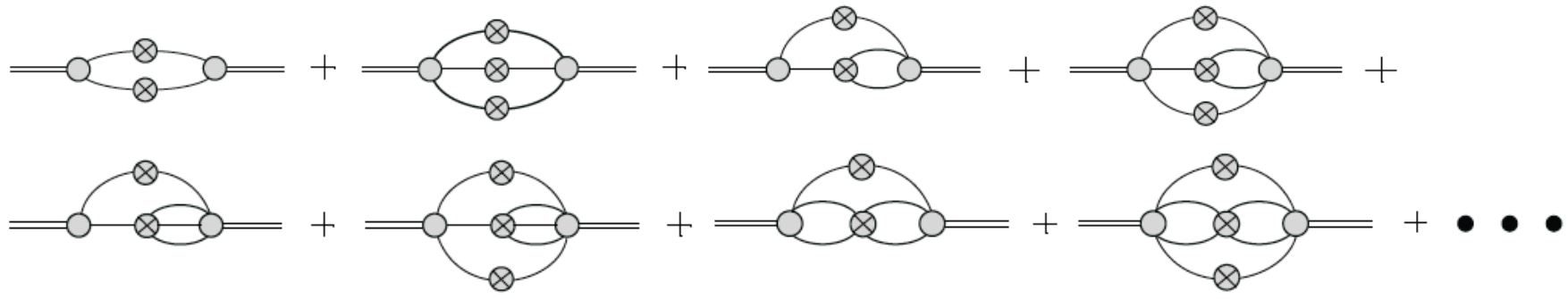
$r(k)-1 = 0, \text{ or } < 0, \text{ or } > 0 ??$

Diagrammatically,

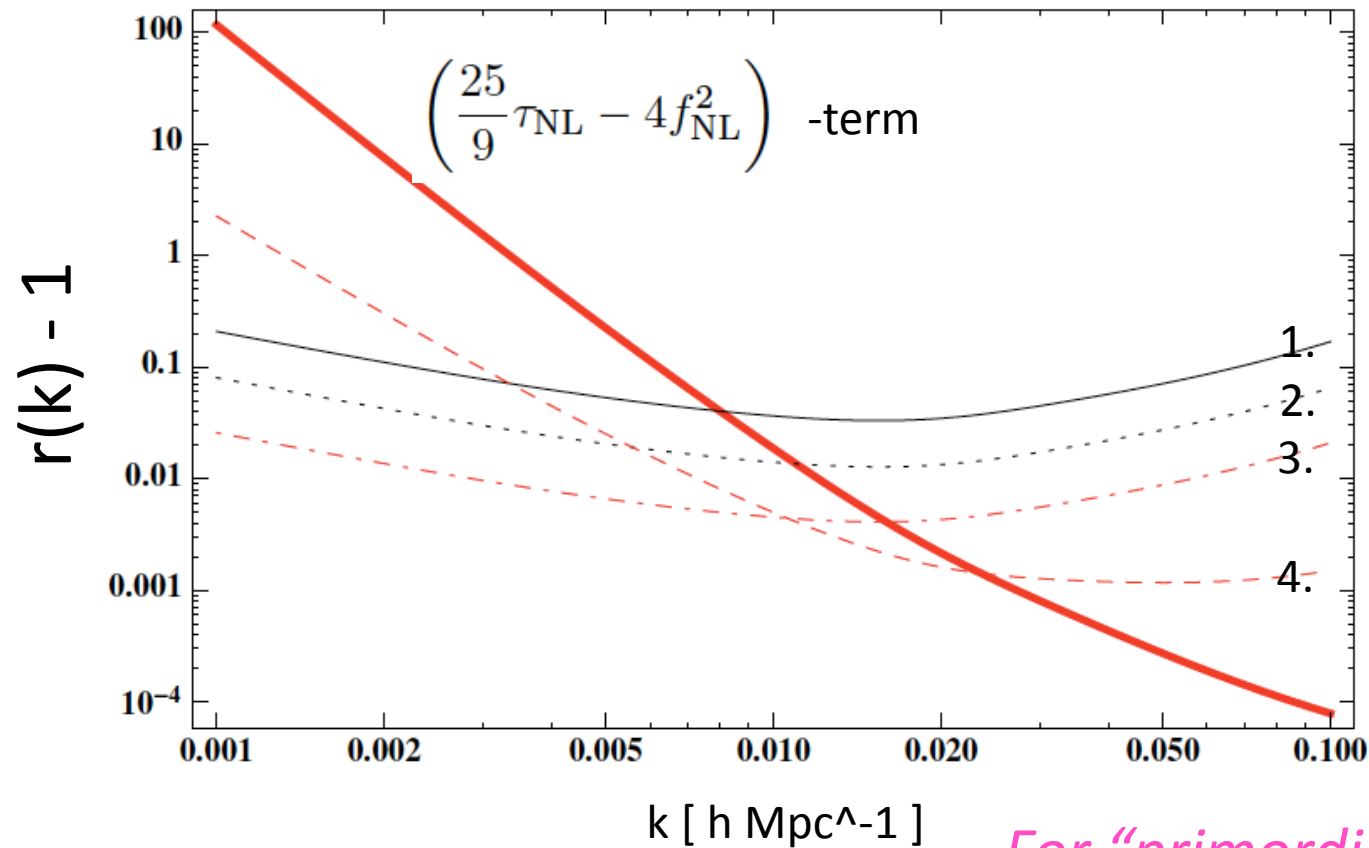


- Higher order contribution (from b2, b3, ...)

$$\begin{aligned}
P_X(k) = & \sum_{n=1}^{\infty} \left[\frac{1}{n!} \int \frac{d^3 p_1 \cdots d^3 p_{n-1}}{(2\pi)^{3n}} \Gamma_X^{(n)}(\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{k} - \mathbf{P}_{n-1})^2 P_L(p_1) P_L(p_2) \cdots P_L(|\mathbf{k} - \mathbf{P}_{n-1}|) \right. \\
& + \frac{1}{(n-1)!} \int \frac{d^3 p_1 \cdots d^3 p_n}{(2\pi)^{(3n)}} \Gamma_X^{(n)}(\mathbf{p}_1, \dots, \mathbf{p}_{n-1}, \mathbf{k} - \mathbf{P}_{n-1}) \\
& \quad \times \Gamma_X^{(n+1)}(-\mathbf{p}_1, \dots, -\mathbf{p}_n, -\mathbf{k} + \mathbf{P}_n) P_L(p_1) \cdots P_L(p_{n-1}) B_L(\mathbf{k} - \mathbf{P}_{n-1}, -\mathbf{p}_n, -\mathbf{k} + \mathbf{P}_n) \\
& + \frac{1}{3(n-1)!} \int \frac{d^3 p_1 \cdots d^3 p_{n+1}}{(2\pi)^{3n+3}} \Gamma_X^{(n)}(\mathbf{p}_1, \dots, \mathbf{p}_{n-1}, \mathbf{k} - \mathbf{P}_{n-1}) \\
& \quad \left. \times \Gamma_X^{(n+2)}(-\mathbf{p}_1, \dots, -\mathbf{p}_{n+1}, -\mathbf{k} + \mathbf{P}_{n+1}) P_L(p_1) \cdots P_L(p_{n-1}) T_L(\mathbf{k} - \mathbf{P}_{n-1}, -\mathbf{p}_n, -\mathbf{p}_{n+1}, -\mathbf{k} + \mathbf{P}_{n+1}) \right] \\
& + \sum_{n=2}^{\infty} \frac{1}{4(n-2)!} \int \frac{d^3 p_1 \cdots d^3 p_{n-1} d^3 q}{(2\pi)^{3n}} \Gamma_X^{(n)}(\mathbf{p}_1, \dots, \mathbf{p}_{n-1}, \mathbf{k} - \mathbf{P}_{n-1}) \\
& \quad \times \Gamma_X^{(n)}(-\mathbf{p}_1, \dots, -\mathbf{p}_{n-2}, -\mathbf{q}, -\mathbf{k} + \mathbf{P}_{n-2} + \mathbf{q}) \\
& \quad \times P_L(p_1) \cdots P_L(p_{n-2}) T_L(\mathbf{p}_{n-1}, \mathbf{k} - \mathbf{P}_{n-1}, -\mathbf{q}, -\mathbf{k} + \mathbf{P}_{n-2} + \mathbf{q}), \tag{33}
\end{aligned}$$



- Higher order in stochasticity parameter



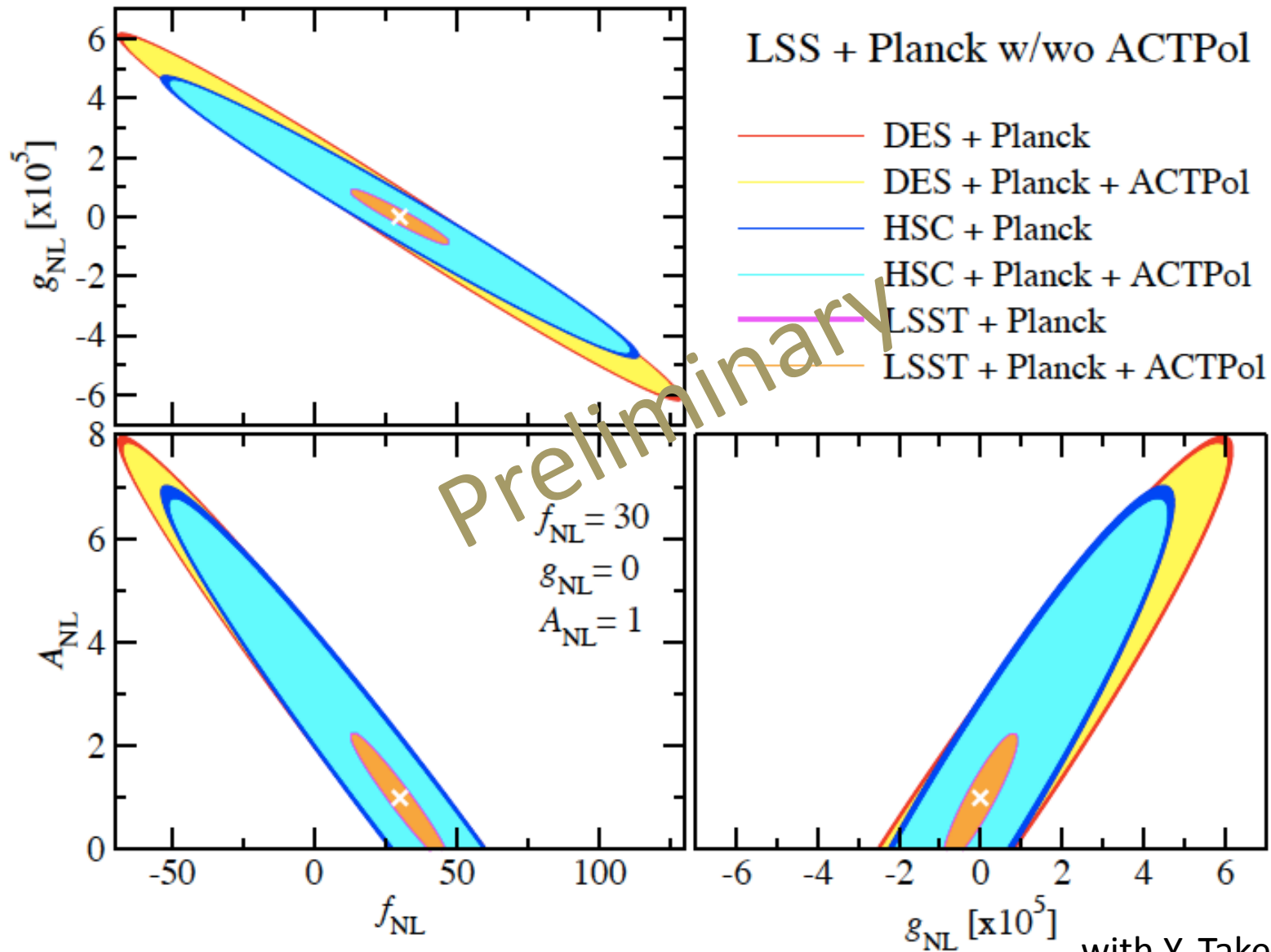
$f_{\text{NL}} = 40$ and $\tau_{\text{NL}} = 5 \times 36 f_{\text{NL}}^2 / 25$.
 fixing $z = 1$ and $M = 5 \times 10^{13} h^{-1} M_{\odot}$

For “primordial stochasticity”,

$$k < \mathcal{O}(10^{-2}) h\text{Mpc}^{-1}$$

$\longleftrightarrow P_X(k) k^3 < 1$ is needed ??

- We are trying to obtain precise forecast...



with Y. Takeuchi in prep.

Summary

- Derive an accurate formula for the bias parameter with primordial non-Gaussianity by using integrated Perturbation Theory
- wide (large scales) and deep (redshift-dependence) surveys are needed.
- Stochasticity of bias \leftrightarrow primordial Suyama-Yamaguchi ineq.
- We find that higher order contributions can be safely neglected in the regime where the dimensionless power spectrum is less than order of unity..
- Galaxy bispectrum?
- Forecast for the constraints on fNL, gNL and tauNL? (HSC, ...)