

Can we really measure f_{nl} from the galaxy power spectrum?

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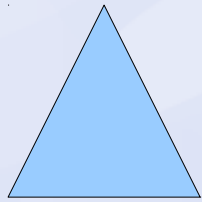
TR33 - Summer Institute
Corfu 2012



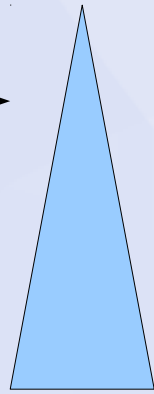
What is Primordial non-Gaussianity?

- ▶ “Primordial”
 - Refers to initial density/potential perturbations
 - Different from NG introduced by non-linear structure formation at late times
- ▶ Describes a wide class of effects with different physical origin
 - Signal determined by specific model of Inflation
 - Standard slow-roll: deviations undetectable
- ▶ Leads to non-zero n-point correlation functions for $n > 2$ (or equivalently: Bispectrum, Trispectrum, ...)
 - A detection would imply new physics!

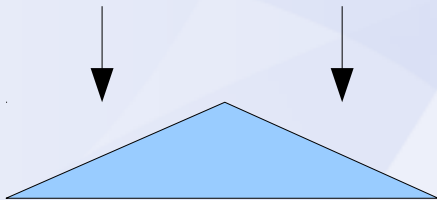
Different types



Equilateral
 $k_1 \approx k_2 \approx k_3$



Local/Squeezed
 $k_1 \approx k_2 \gg k_3$



Folded
 $k_1 \approx k_2 \approx k_3/2$

- Bispectrum “templates” defined by triangle configurations which maximize the signal in the 3-point function
- Most common: local type

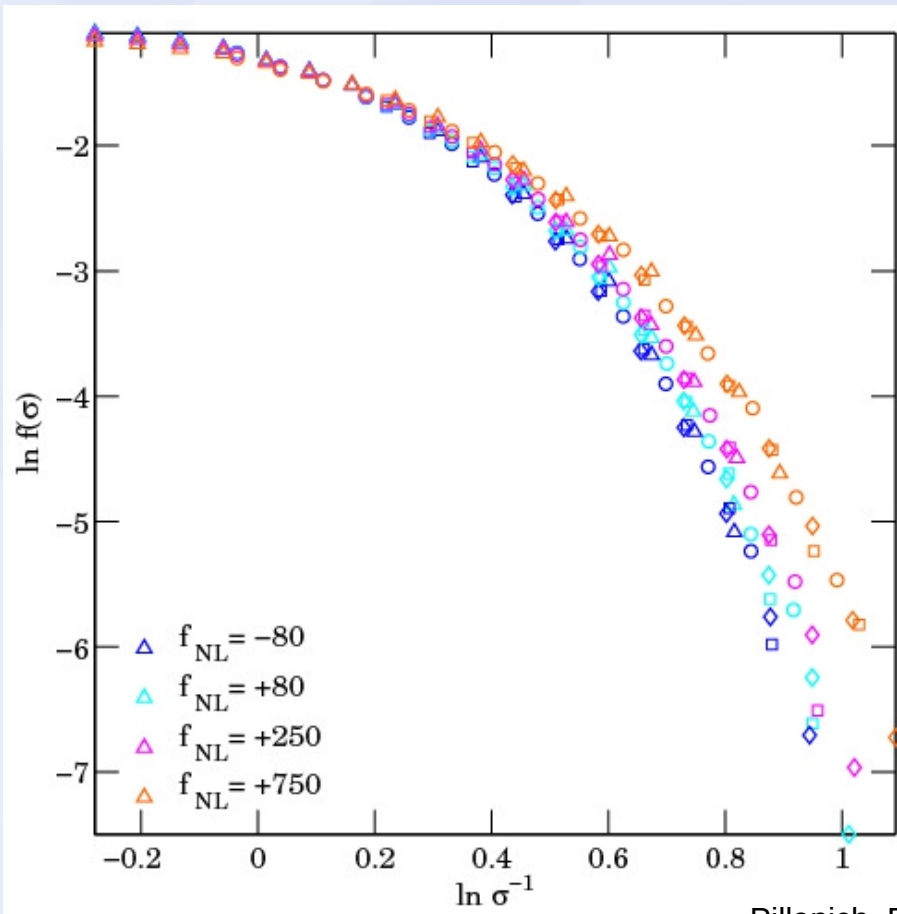
$$\Phi(\mathbf{x}) = \phi(\mathbf{x}) + f_{\text{NL}}^{\text{loc}} \phi^2(\mathbf{x}) + g_{\text{NL}}^{\text{loc}} \phi^3(\mathbf{x})$$

$$B_{\Phi}^{\text{loc}}(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3) = 2f_{\text{NL}}^{\text{loc}} [P_{\phi}(k_1)P_{\phi}(k_2) + (\text{cyc.})]$$

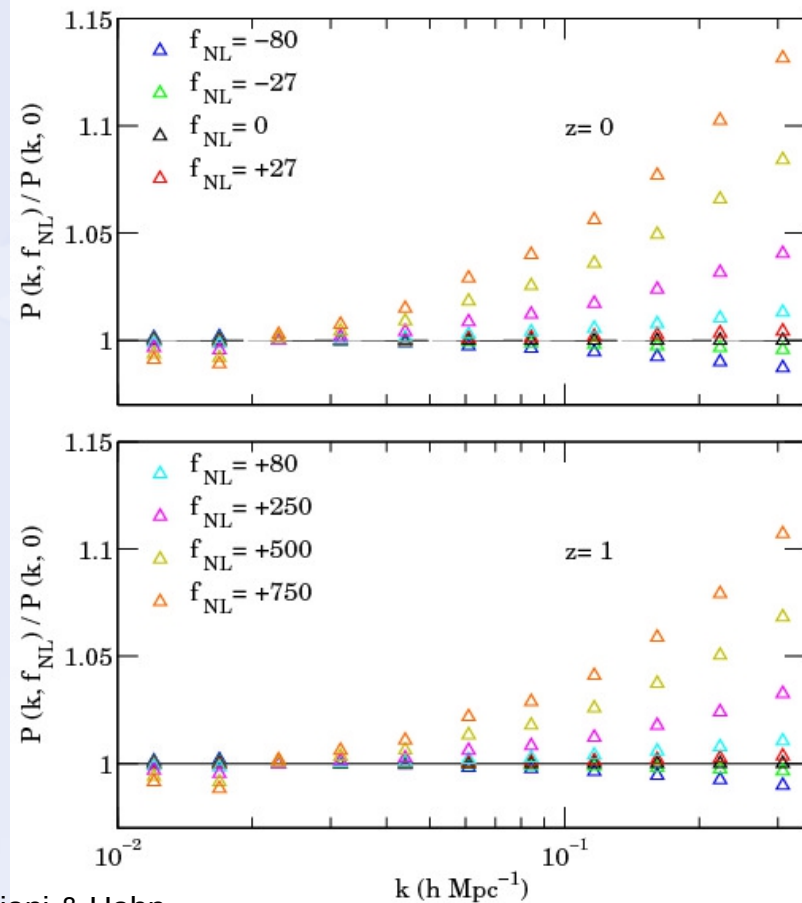
- ✓ Many signals are maximal for local model
- ✗ Other types have to be approximated or calculated numerically

Effects on large-scale structure

Halo mass function



Matter power spectrum



Scale-dependent bias

(e.g. Dalal et al. 2008; Smith, Ferraro & LoVerde 2012)

$$\frac{P_{\text{hm}}(k)}{P_{\text{m}}(k)} = b_1(f_{\text{NL}}, g_{\text{NL}}) + \frac{\beta_f f_{\text{NL}} + \beta_g g_{\text{NL}}}{\alpha(k)}.$$

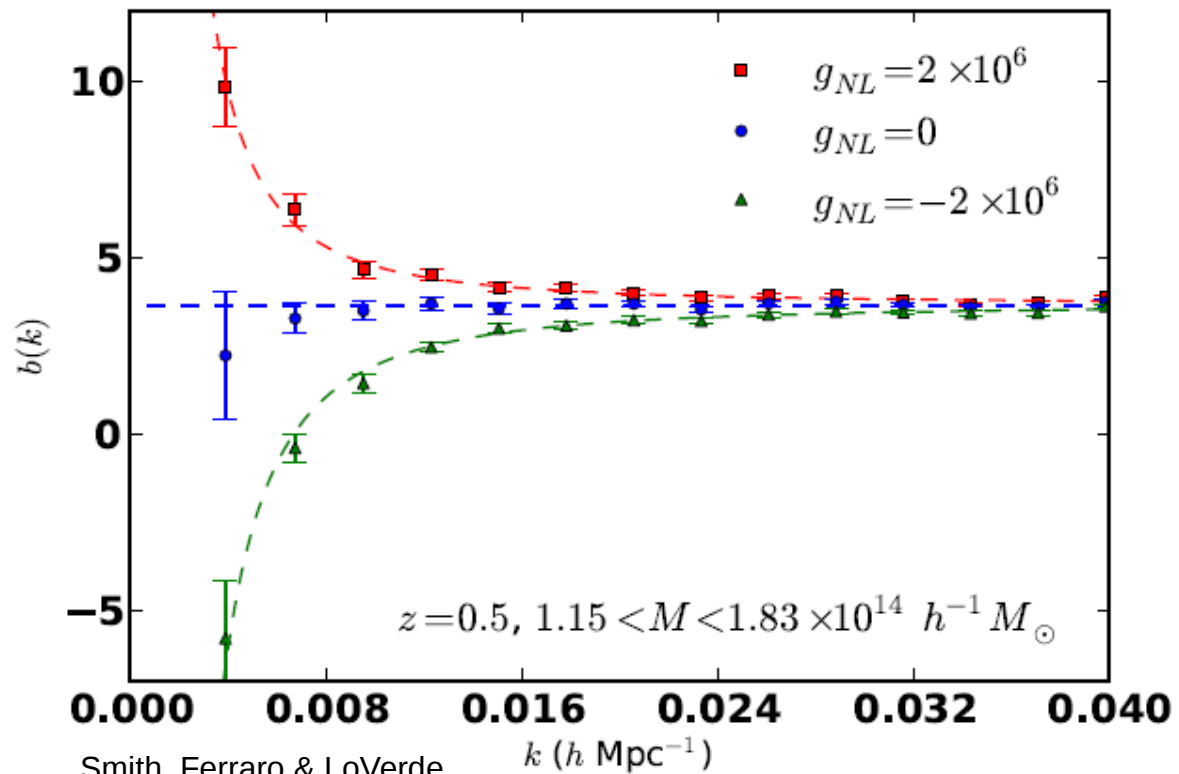
$$\beta_f = 2\nu^2 - 2$$

$$\alpha(k, z) = \frac{2k^2 D(z) T(k)}{3\Omega_{\text{m}} H_0^2}$$

$$\nu = [\delta_c(b_1 - 1) + 1]^{1/2} \quad (\text{where } \delta_c = 1.42)$$

$$\beta_g = \kappa_3 \left[-0.7 + 1.4(\nu - 1)^2 + 0.6(\nu - 1)^3 \right] - \frac{d\kappa_3}{d \log \sigma^{-1}} \left(\frac{\nu - \nu^{-1}}{2} \right)$$

Scale-dependent bias

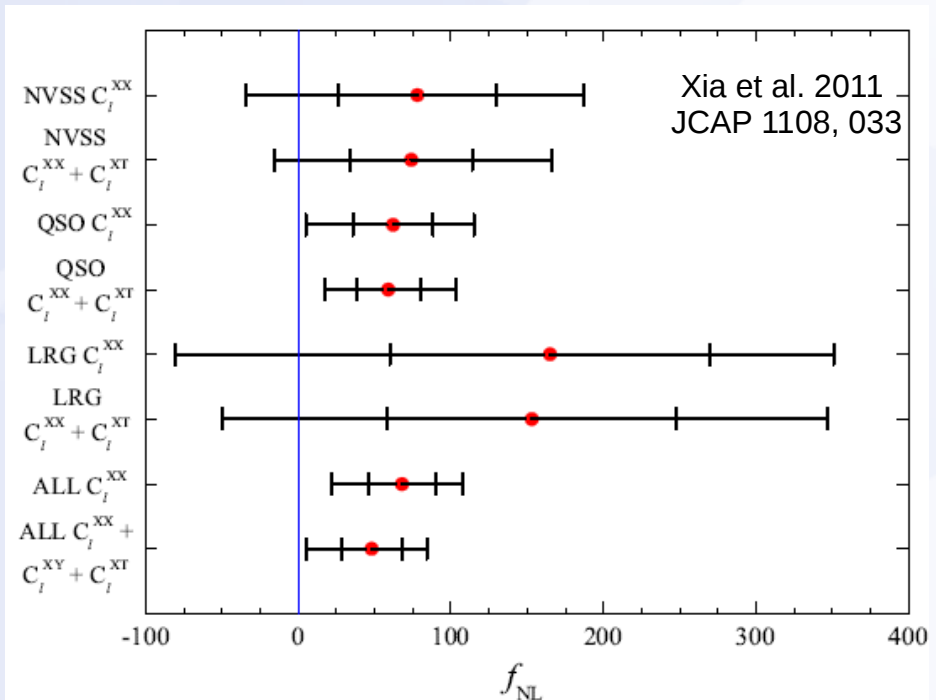


Smith, Ferraro & LoVerde
2012, JCAP 03, 032

A detection of $f_{\text{NL}} > 0$?

(Xia et al. 2011)

- Analysis of angular power spectra of radio sources (NVSS), SDSS QSOs, MegaZ-LRGs (SDSS II) & cross-correlation with CMB temperature map
- No indication for non-local type of PNG
- No indication for (local) $g_{\text{NL}} \neq 0$
- Some indication of local-type PNG with $f_{\text{NL}} \approx 40 \pm 20$



Critical assumption: The underlying model has either $f_{\text{NL}} \neq 0$ or $g_{\text{NL}} \neq 0$.

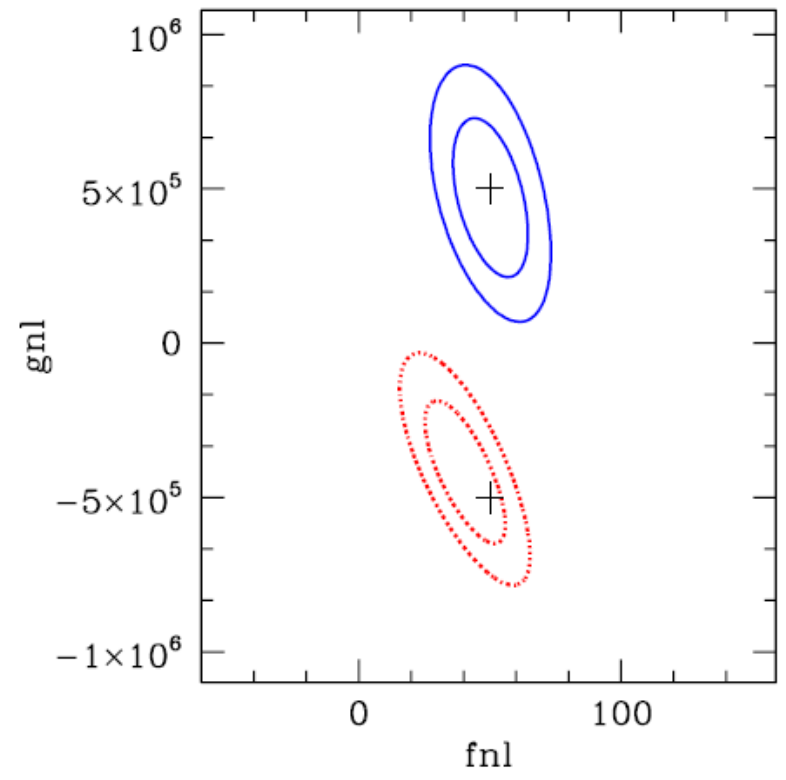
Simulations

- ▶ WMAP 5 cosmology, comoving 1.2 Gpc/h box
- ▶ $\{f_{\text{nl}}, g_{\text{nl}}\} = \{50, \pm 5e5\}$ (within current limits from CMB), 4 realizations each
- ▶ Mass range: $10^{13} - 10^{15} M_{\odot}/h$ (at redshift 0)
- ▶ “Current sample”: $z = \{0.5, 1, 1.5\}$ and $b_1 = 1.8 - 2.9$
- ▶ “Future sample”: $z = \{0, 0.5, 1, 1.5, 2\}$ and $b_1 = 1.8 - 5.9$

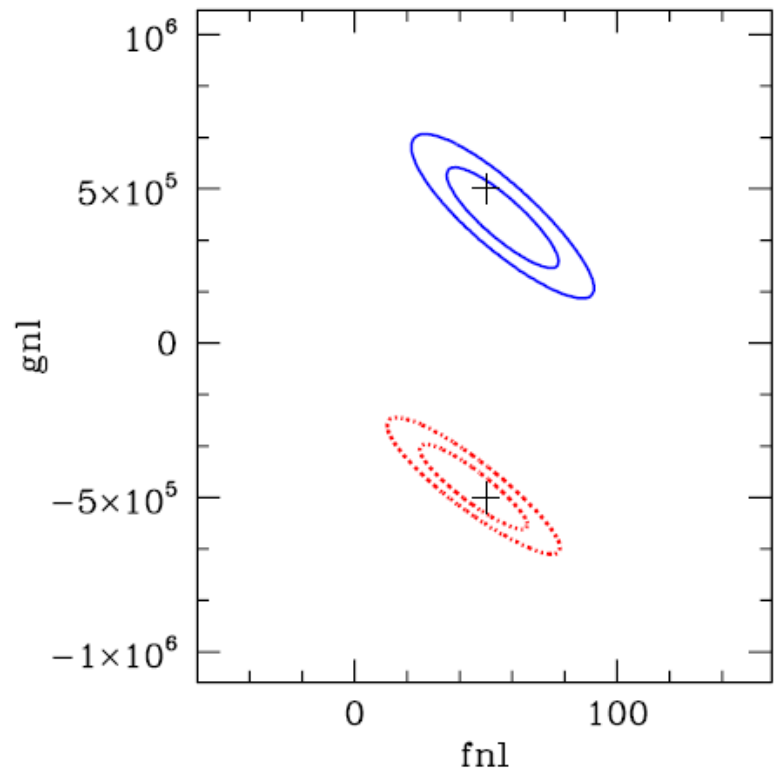
$$\frac{P_{\text{hm}}(k)}{P_{\text{m}}(k)} = b_1(f_{\text{NL}}, g_{\text{NL}}) + \frac{\beta_f f_{\text{NL}} + \beta_g g_{\text{NL}}}{\alpha(k)}.$$

- 1) Calculate P_{hm} and P_{mm} (average over realizations)
- 2) Get $b(k)$ from ratio
- 3) Find best $f_{\text{nl}}, g_{\text{nl}}$ (likelihood) after marginalization over b_1

Degeneracy

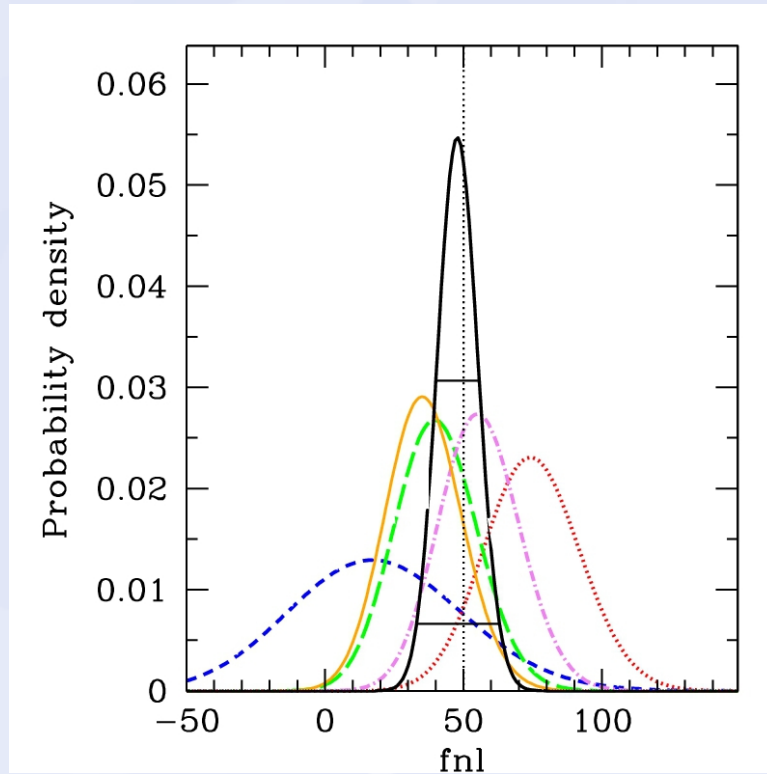


“Current sample”
(3 bins, up to $z=1.5$)

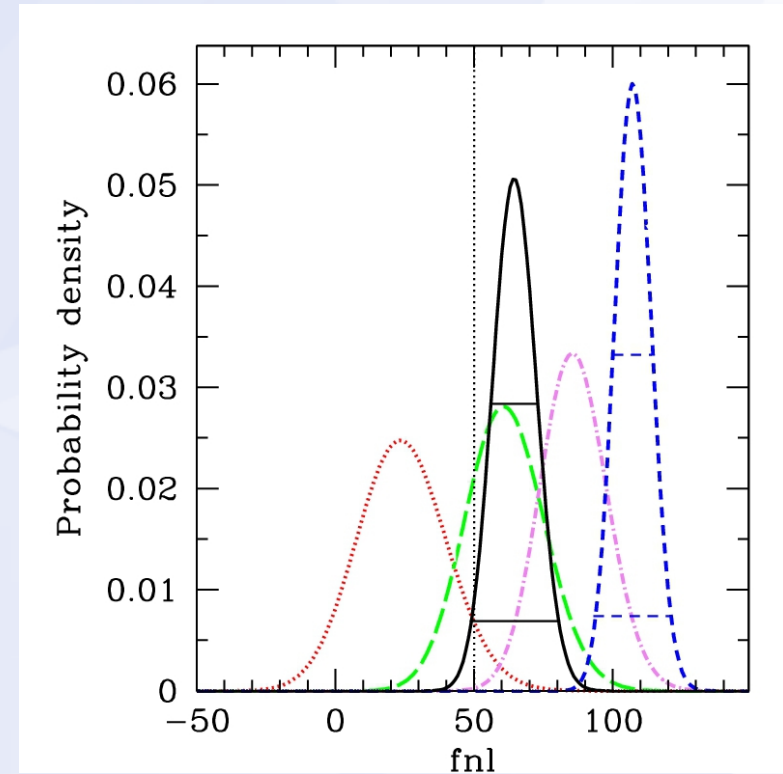


“Future sample”
(5 bins, up to $z=2$)

One-parameter model (fnl)

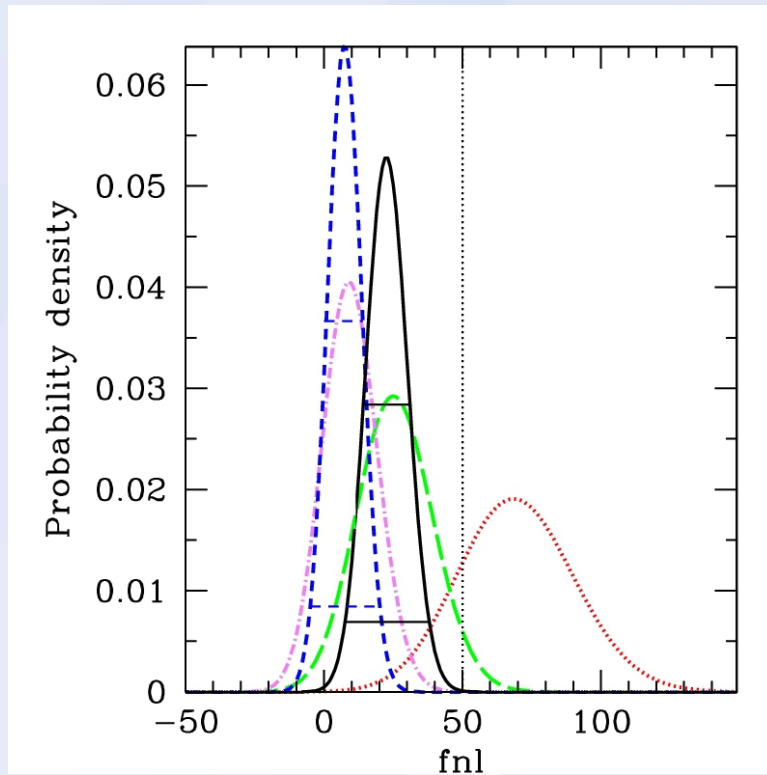


Simulation: $f_{nl} = 50$, $g_{nl} = +5e5$
Model: $g_{nl} = +5e5$, f_{nl} fitted, "future sample"

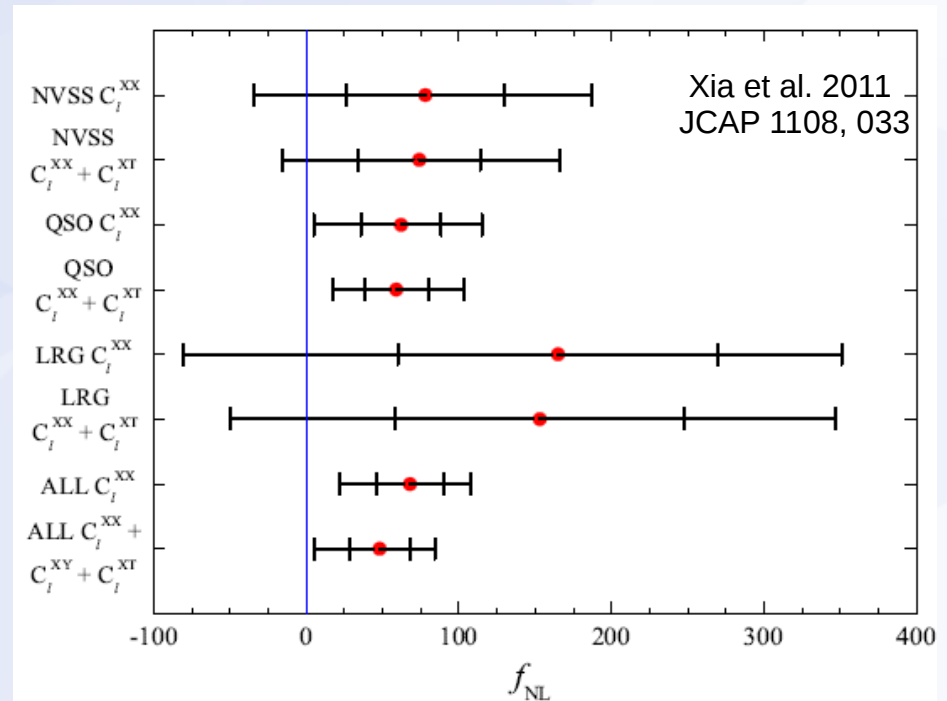


Simulation: $f_{nl} = 50$, $g_{nl} = +5e5$
Model: $g_{nl} = 0$, f_{nl} fitted, both samples

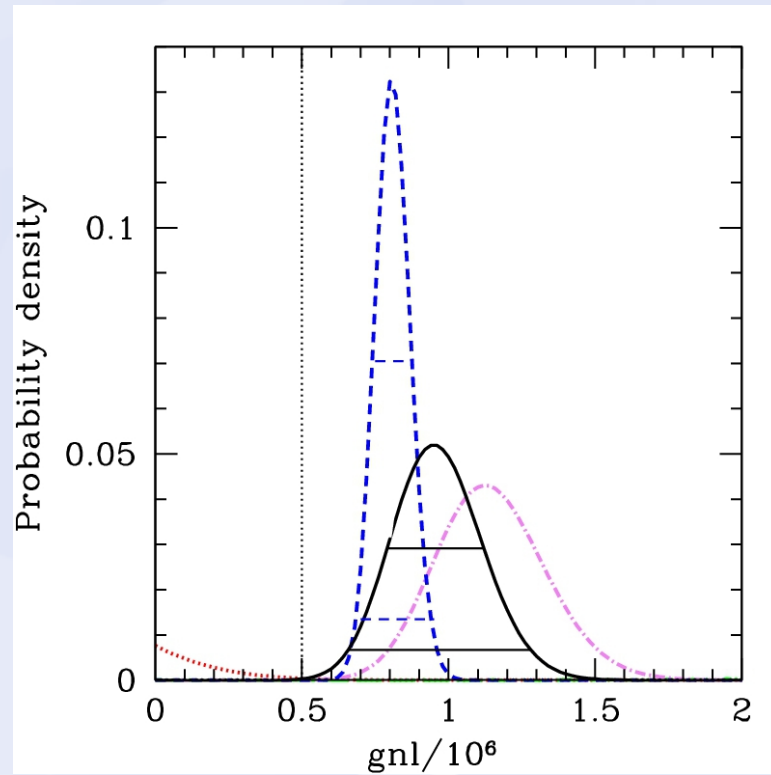
Redshift dependence?



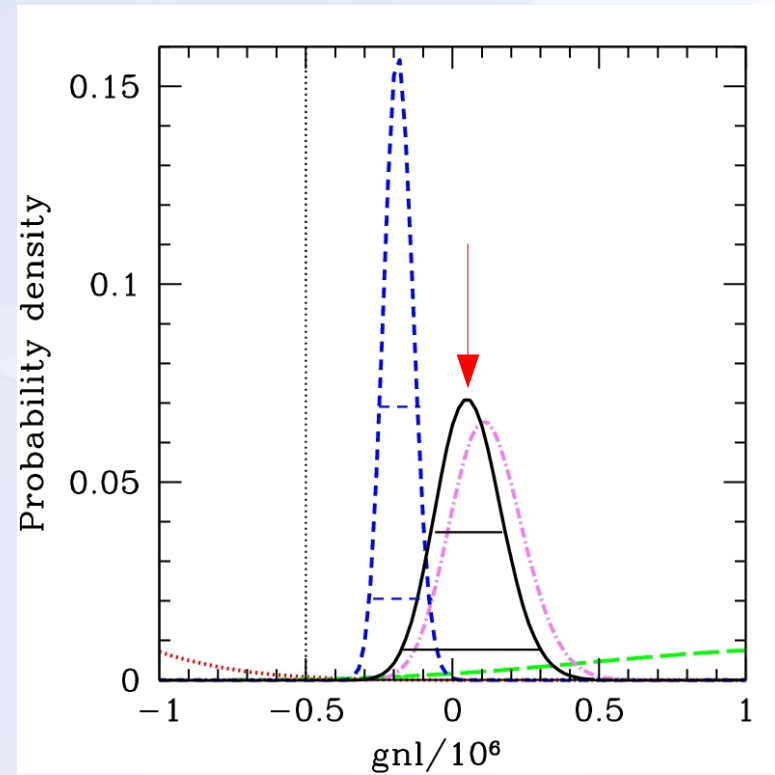
Simulation: $f_{NL} = 50$, $g_{NL} = -5e5$
 Model: **$g_{NL} = 0$, f_{NL} fitted**, both samples



One-parameter model (gnl)



Simulation: $fnl = 50$, $gnl = +5e5$
Model: **$fnl = 0$** , **gnl fitted**, both samples



Simulation with $gnl = -5e5$

Model selection

- ▶ Bayes factor (computed from the likelihood)

$$B_{ab} = \frac{P(D|M_a)}{P(D|M_b)} = \frac{\int \mathcal{L}_a(D|M_a, \theta_a) \pi_a(\theta_a) d\theta_a}{\int \mathcal{L}_b(D|M_b, \theta_b) \pi_b(\theta_b) d\theta_b}$$

- ✓ Also works for nested models (i.e. fixing one parameter)
- ▶ Jeffreys scale (Jeffreys 1961)
 - $B_{ab} < 1$ favors model 'b'
 - $B_{ab} > 30$ gives “very strong” evidence to model 'a'

➔ We find that the 2-parameter model is always favored with $B_{ab} > 30$.

Summary

- ▶ Scale-dependent bias effects from f_{nl} and g_{nl} are degenerate
 - Just adding mass/redshift bins does not help
 - Higher-order statistics can break degeneracy
- ▶ Purely quadratic model ($g_{nl} = 0$)
 - Leads to $f_{nl} = f_{nl}(z)$, which depends on the sign of g_{nl}
 - “Best fit” estimates differ significantly from input value
- ▶ Purely cubic model ($f_{nl} = 0$)
 - Effects can cancel out when actual signs are opposite
 - “Best fit” estimates differ significantly from input value
- ▶ Model selection techniques
 - Can distinguish between different models
 - Should be applied to observations