

Non-Gaussianity from cubic order primordial perturbations: signatures in the CMB

Pravabati Chingangbam

Korea Institute for Advanced Study, South Korea.

Plan of talk

- Introduction
 - ▶ CMB temperature fluctuations as a random field
 - ▶ Initial conditions from inflation
- Simulating CMB maps with non-Gaussian primordial perturbations.
- Measuring non-Gaussianity
 - ▶ Genus
- Summary and forthcoming work

CMB observations today

- ▶ In any direction in the sky the CMB photons behave like black-body radiation with a temperature T_0 .
- ▶ WMAP discretizes the sky sphere into about a million pixels and measures the temperature fluctuation in each pixel,

$$\frac{\Delta T(\hat{n})}{T_0} \equiv \frac{T(\hat{n}) - T_0}{T_0}.$$

- ▶ $\Delta T(\hat{n})$ gives **ONE** realization of a 2 dimensional random field with variance of order 10^{-5} .
- ▶ In multipoles:

$$\frac{\Delta T(\hat{n})}{T_0} = \sum_{\ell m} a_{\ell m} Y_{\ell m}(\hat{n}), \quad a_{\ell m} = \int d\Omega \frac{\Delta T(\hat{n})}{T_0} Y_{\ell m}^*(\hat{n}).$$

ΔT as a random field

- ▶ Mean:

$$\left\langle \frac{\Delta T(\hat{n})}{T_0} \right\rangle = 0, \quad \langle a_{\ell m} \rangle = 0.$$

- ▶ Variance:

$$C(\theta) = \left\langle \frac{\Delta T(\hat{n}_1)}{T_0} \frac{\Delta T(\hat{n}_2)}{T_0} \right\rangle_{\hat{n}_1 \cdot \hat{n}_2 = \cos \theta}, \quad C_\ell = \frac{1}{2\ell + 1} \sum_{m=-\ell}^{\ell} |a_{\ell m}|^2.$$

If ΔT is Gaussian:

$$C(\theta) = \frac{1}{4\pi} \sum_{\ell} (2\ell + 1) C_\ell P_\ell(\theta).$$

- ▶ Mean and variance completely characterize the field.
- ▶ If ΔT is non-Gaussian we need to know all higher moments to find out its distribution function.

ΔT from theory

$a_{\ell m}$ can be calculated as:

$$a_{\ell m} = 4\pi(-i)^\ell \int \frac{d^3k}{(2\pi)^3} \Phi(\vec{k}, \eta_i) \Delta_\ell(k, \eta_0) Y_{\ell m}^*(\hat{k}),$$

- ▶ $\Phi(\vec{k}, \eta_i) \equiv$ Fourier transform of primordial metric perturbations $\Phi(\vec{x}, \eta_i)$ at sufficiently early time η_i .
- ▶ $\Delta_\ell(k, \eta_0)$: radiation transfer function.
Determined by the mechanism of creation and the physics around decoupling epoch and subsequent history.

Computing $\Delta_\ell(k, \eta_0)$

- ▶ Let f = distribution function for photons.
- ▶ At sufficiently early time **mean free path of photons** \ll **Hubble length**, and we can assume there is equilibrium in local regions, with T fluctuating from region to region about some average value T_0 .
- ▶ Then we can write

$$f(\vec{x}, \vec{p}, \eta) = \frac{1}{e^{p/T(\vec{x}, \hat{p}, \eta)} - 1}, \quad f^{(0)}(p, \eta) \equiv \frac{1}{e^{p/T} - 1},$$

with

$$T(\vec{x}, \hat{p}, \eta) = T_0(1 + \Delta(\vec{x}, \hat{p}, \eta))$$

- ▶ f can be thought of as perturbed about f^0 as:

$$f(\vec{x}, \vec{p}, \eta) = f^{(0)}(p, \eta) + \delta f(\vec{x}, \vec{p}, \eta),$$

assuming only **linear perturbations**.

Computing $\Delta_\ell(k, \eta_0)$

$\delta f(\vec{x}, \vec{p}, \eta)$ can be related to $\Delta(\vec{x}, \hat{p}, \eta)$ as

$$\delta f(\vec{x}, \vec{p}, \eta) = -p \frac{\partial f^0}{\partial p} \Delta(\vec{x}, \hat{p}, \eta).$$

- ▶ Time evolution of f can be obtained via Boltzmann equation :

$$\begin{aligned} \frac{df}{dt} &= C[f] \\ C[f] &= \text{collision term.} \end{aligned}$$

- ▶ This translates into a first order equation for $\Delta(\vec{x}, \hat{p}, \eta)$ with all the interactions going in as the source terms.
- ▶ Then we do the following:
 - ▶ Fourier transform the \vec{x} dependence.
 - ▶ Define $\mu \equiv \hat{k} \cdot \hat{p}$ and transform to multipoles ℓ .
 - ▶ Integrate in time.
- ▶ The final result is $\Delta_\ell(k, \eta_0)$. Computed by publicly available codes : CMBFAST, CAMB.

Initial conditions $\Phi(\vec{k}, \eta_i)$: inflation

- ▶ **Inflation** tells us $\Phi(\vec{k}, \eta_i)$ is a random field since it comes from vacuum fluctuations of inflaton.
- ▶ Variance given by inflationary power spectrum:

$$P_{\Phi}(k) \sim \langle \Phi_k \Phi_k \rangle = \frac{A_0}{k^3} \left(\frac{k}{k_0} \right)^{n_s - 1} .$$

- ▶ Hence by studying the properties of $\Delta T(\hat{n})$ we are 'directly' probing properties of the inflaton field.

How can ΔT be non-Gaussian ?

a_{lm} can be non-Gaussian due to :

1. non-linear transfer function $\Delta_{\ell}(k, \eta_0)$.
 - ▶ Expected to be very small since linear perturbation theory has proved to be a very good approximation.
2. Or non-Gaussian $\Phi(k)$:
 - ▶ Inflationary perturbation theory when treated to non-linear order predict deviation of $\Phi(k)$ from Gaussianity. True of **ALL** models. Can express deviation as:

$$\begin{aligned}\Phi &= \Phi^G + \Delta\Phi, \\ \Delta\Phi &\ll \Phi^G.\end{aligned}$$

- ▶ The **amount** and the **functional form of deviation** is model dependent. Detailed knowledge can discriminate between different models.

Prediction of $\Delta\Phi$ from inflation in Fourier space

Second order correction to Φ^G is of order $\sim (\Phi^G)^2$. [Salopek & Bond (1990)]

3-point function :

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

- ▶ Different models predict different magnitude and shape of $F(k_1, k_2, k_3)$.

4-point function :

$$\langle \Phi(\vec{k}_1)\Phi(\vec{k}_2)\Phi(\vec{k}_3)\Phi(\vec{k}_4) \rangle = (2\pi)^3 \delta^3(\vec{k}_1 + \vec{k}_2 + \vec{k}_3 + \vec{k}_4) H(k_1, k_2, k_3, k_4).$$

- ▶ Different models predict different magnitude and shape of $H(k_1, k_2, k_3, k_4)$.

Prediction of $\Delta\Phi$ from inflation in configuration space

- ▶ Schematically

$$\Phi(\vec{x}) = \Phi^G(\vec{x}) + \int d^3y d^3z K(\vec{y}, \vec{z}) \Phi^G(\vec{x} - \vec{y}) \Phi^G(\vec{x} - \vec{z}) + \dots$$

- ▶ Consider the simplified **ansatz**

$$\Phi(\vec{x}) = \Phi^G(\vec{x}) + f_{NL} \left((\Phi^G(\vec{x}))^2 - \langle (\Phi^G)^2 \rangle \right) + \dots$$

- ▶ Characterized by non-linearity parameter f_{NL} .
- ▶ *Local* since the non-linear contributions depend only on same spatial point.
- ▶ f_{NL} is very well studied theoretically and observationally. The tightest constraint from WMAP observation so far

$$-4 < f_{NL} < 80 \quad (95\%CL).$$

Smith and Zaldariagga (2009)

Beyond f_{NL}

We can write $\Phi(\vec{x})$ to third order as:

$$\Phi(\vec{x}) = \Phi^G(\vec{x}) + f_{NL} \left((\Phi^G(\vec{x}))^2 - \langle (\Phi^G)^2 \rangle \right) + g_{NL} (\Phi^G(\vec{x}))^3 + \dots$$

- ▶ Becomes relevant if g_{NL} can be relatively large.
- ▶ Several recent works have shown that in **curvaton models** or **multibrid models** or **ekpyrotic scenario** it can happen that f_{NL} is small or even zero while g_{NL} can be large $\sim \mathcal{O}(10^5)$.

Allen, Grinstein & Wise (1987)

Sasaki, Valiviita & Wands (2006), Byrnes, Sasaki & Wands (2006),
Enqvist & Takahashi (2008), Huang (2008), PC & Huang (2009);
Sasaki (2008), Huang (2009);

Renaux-Petel (2009).

- ▶ f_{NL} can be zero due to symmetry such as $\Phi \rightarrow -\Phi$ or special cancellations of terms.

Simulating non-Gaussian maps

Why is it important :

- ▶ it is just solving the time evolution of the temperature fluctuations and hence understanding what theory is predicting.
- ▶ more importantly, the simulations can be used as testbeds to study what we should be looking for in the observational data.
- ▶ Numerically highly non-trivial, need fast method.

P.C and Changbom Park, astro-ph/0908.1696 [astro-ph.CO]

Simulating non-Gaussian maps

Liguori, Mattarese & Moscardini (2003)

Rewrite $a_{\ell m}$ as real space integral

$$a_{\ell m} = \int dr r^2 \Phi_{\ell m}(r) \Delta_{\ell}(r)$$

where

$$\Delta_{\ell}(r) \equiv \frac{2}{\pi} \int dk k^2 \Delta_{\ell}(k) j_{\ell}(kr),$$

$$\Phi_{\ell m}(r) \equiv \frac{(-i)^{\ell}}{2\pi^2} \int dk k^2 \Phi_{\ell m}(k) j_{\ell}(kr),$$

$$\Phi_{\ell m}(k) = 4\pi(i)^{\ell} \int dr r^2 \Phi_{\ell m}(r) j_{\ell}(kr),$$

$\Phi_{\ell m}(r)$ split into Gaussian and non-Gaussian parts:

$$\Phi_{\ell m}(r) \equiv \Phi_{\ell m}^G(r) + f_{NL} \Phi_{\ell m}^{NG}(r) + g_{NL} \Phi_{\ell m}^{NNG}(r)$$

Simulating non-Gaussian maps

First generate $\Phi_{\ell m}^G(r)$ in (ℓ, m, r) space.

To compute $\Phi_{\ell m}^{NG}(r)$:

- ▶ Harmonic transform to get $\Phi^G(\vec{r}) = \sum_{\ell m} \Phi_{\ell m}^G(r) Y_{\ell m}(\hat{r})$.
- ▶ Square at each \vec{r} , subtract variance to get $\Phi^{NG}(\vec{r})$.
- ▶ Inverse harmonic transform to get $\Phi_{\ell m}^{NG}(r) = \int d\Omega \Phi_{\ell m}^{NG}(r) Y_{\ell m}^*(\hat{r})$.

To get $\Phi_{\ell m}^{NNG}(r)$:

- ▶ Harmonic transform to get $\Phi^G(\vec{r}) = \sum_{\ell m} \Phi_{\ell m}^G(r) Y_{\ell m}(\hat{r})$.
- ▶ Cube at each \vec{r} to get $\Phi^{NG}(\vec{r})$.
- ▶ Inverse harmonic transform to get $\Phi_{\ell m}^{NNG}(r) = \int d\Omega \Phi_{\ell m}^{NNG}(r) Y_{\ell m}^*(\hat{r})$.

Inputs for the simulations

Physical parameters of Λ CDM:

- ▶ Used WMAP 5-year best fit parameter values.
- ▶ Taken $n_s = 1$.
- ▶ Set $f_{NL} = 0$.

Resolution:

- ▶ Used $\ell_{max} = 1100$.
- ▶ Resolves points of angular separation $\theta \sim 9.8 \text{ arcmin}$
- ▶ Used Healpix $N_{side} = 512$, corresponds to dividing sphere into about 3×10^6 pixels.

Normalization

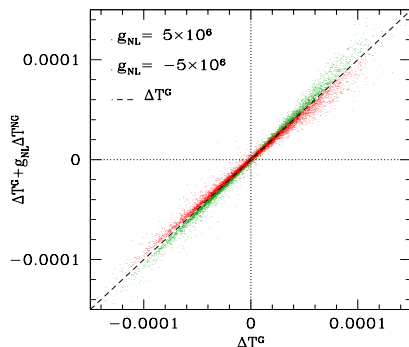
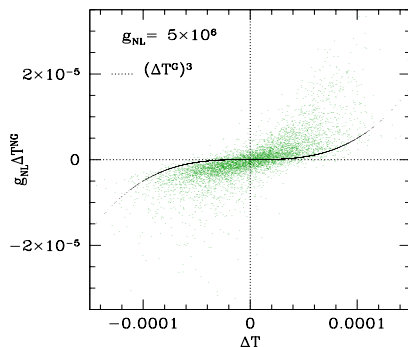
- ▶ Gaussian maps normalized by CMBFAST.
- ▶ Non-Gaussian maps normalized by matching with Gaussian maps at $\ell = 220$.

Temperature maps

Full temperature fluctuations:

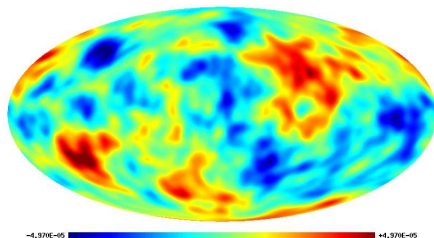
$$\Delta T(\hat{n}) = \Delta T^G + f_{NL}\Delta T^{NG} + g_{NL}\Delta T^{NNG}$$

Distribution of ΔT^{NNG} about ΔT^G :

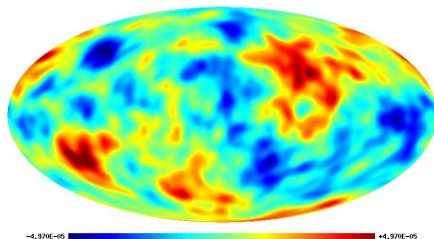


Maps: Positive g_{NL}

Gaussian map with smoothing FWHM= 7° :

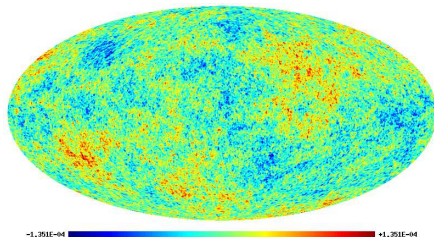


Non-Gaussian map with $g_{NL} = 5 \times 10^6$, same smoothing :

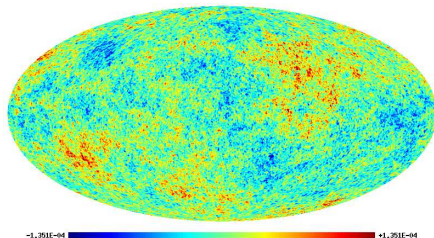


Maps: Positive g_{NL}

Gaussian map with smoothing FWHM= 30':

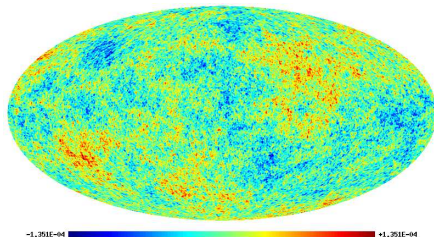


Non-Gaussian map with $g_{NL} = 5 \times 10^6$, same smoothing :

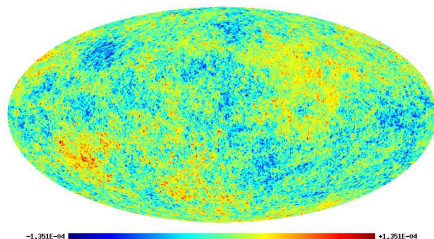


Maps : Negative g_{NL}

Gaussian map with smoothing FWHM= 30':

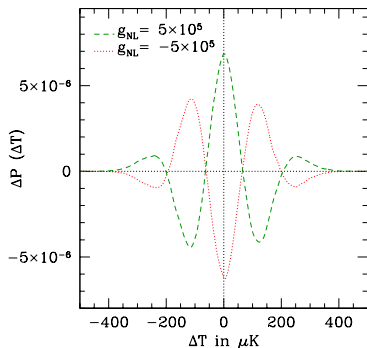
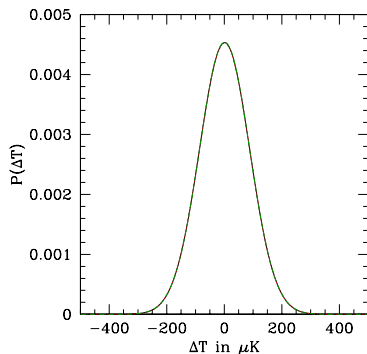


Non-Gaussian map with $g_{NL} = -5 \times 10^6$, same smoothing :



One-point PDF : g_{NL} maps

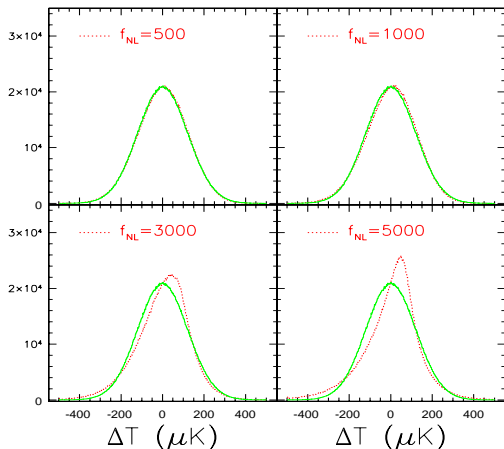
- ▶ g_{NL} affects the kurtosis.



One-point PDF : f_{NL} maps

Liguori, Mattarese & Moscardini (2003)

- ▶ f_{NL} affects the skewness.



Measuring non-Gaussianity

- ▶ Need observables that are sensitive to non-Gaussianity.
- ▶ Different statistical tools complement each other and provide cross checks though their sensitivities may vary.
- ▶ Defined on harmonic space, pixel (real) space, wavelet space, . . .

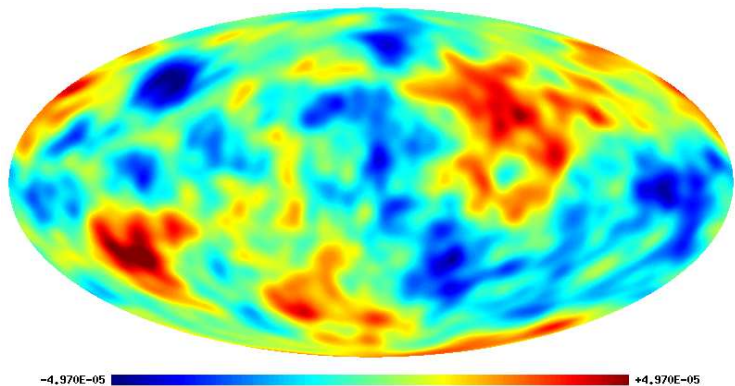
List a few:

1. *Harmonic space observables*: 3-point and 4-point function in multipole space.
2. *Real space observables* : Minkowski functionals, Pixel clustering correlation, etc.
3. *Wavelet, Needlet bispectrum etc.*

Genus: Useful features

- ▶ Real space quantity \Rightarrow real world issues such as foreground, galaxy mask etc., are easy to handle.
- ▶ Contains information of all correlators since they are global quantities.
- ▶ Hence can be more sensitive, compared to measuring individual N-point functions, to other forms of non-Gaussianity than purely f_{NL} or purely g_{NL} .

Genus, G



► G = number of hotspots - number of cold spots.

Genus, G

- ▶ Threshold:

$$\nu \equiv \frac{\Delta T / T}{\sigma_0}, \quad \sigma_0 = \sqrt{\left\langle \frac{\Delta T}{T} \frac{\Delta T}{T} \right\rangle}.$$

- ▶ Then G is given by

$$G(\nu) = \frac{1}{2\pi} \int_C K ds$$

$C \equiv$ contour connecting pixels with same ν

$K \equiv$ curvature of C

Genus for Gaussian field

For Gaussian random field:

$$G = A \nu e^{-\nu^2}$$

where

$$A = \frac{1}{(2(2\pi)^{3/2})} \frac{\sum \ell(\ell+1)(2\ell+1) C_\ell W_\ell^2}{\sum (2\ell+1) C_\ell W_\ell^2}$$

$$W_\ell = e^{-\ell(\ell+1)\theta_s^2/2} = \text{Gaussian smoothing kernel},$$

$$\theta_s = \text{smoothing angle}.$$

It is:

- ▶ independent of the normalization.
- ▶ depends crucially on the shape of C_ℓ .
- ▶ sensitive to the smoothing scale.

Genus for weakly non-Gaussian field

Upto f_{NL} order :

- ▶ When the field is *weakly non-Gaussian* approximate analytic expressions **may be** obtained.
- ▶ Then the G 's can be expressed as

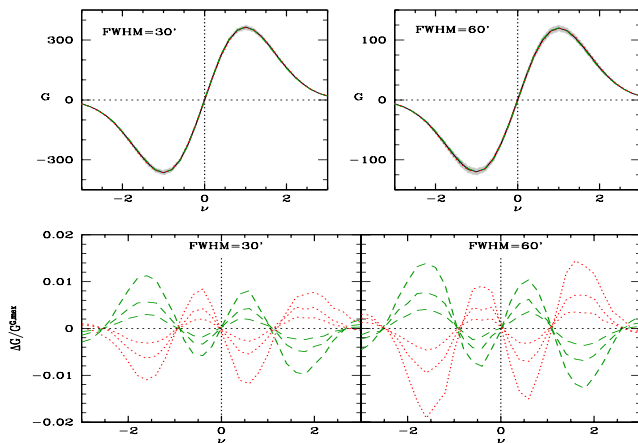
$$G = G^G + \Delta G.$$

- ▶ ΔG is known for f_{NL} type non-Gaussianity.
Matsubara (2003), Hikage et al (2006), Hikage et al (2008).

Upto g_{NL} order :

- ▶ No known approximate analytic expressions upto g_{NL} .
- ▶ Can be computed directly using simulated non-Gaussian maps.

Genus for g_{NL}

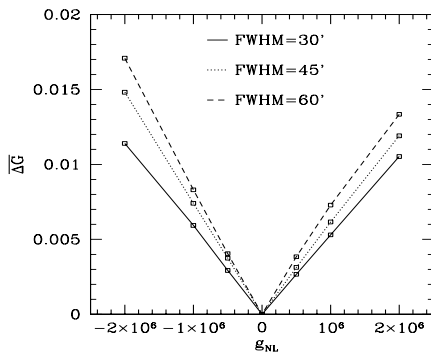


Green represents $g_{NL} > 0$, red $g_{NL} < 0$

Values are $g_{NL} = \pm 5 \times 10^5, \pm 10^6, 2 \times 10^6$

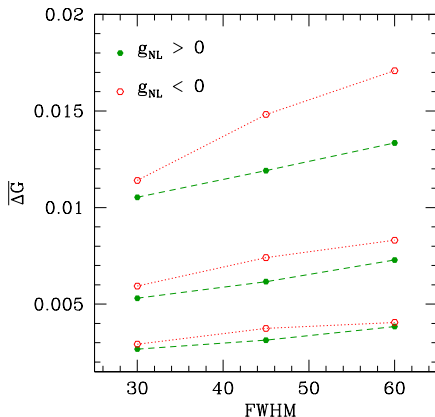
ΔG functional dependence on g_{NL}

- ▶ ΔG depends linearly on g_{NL} :



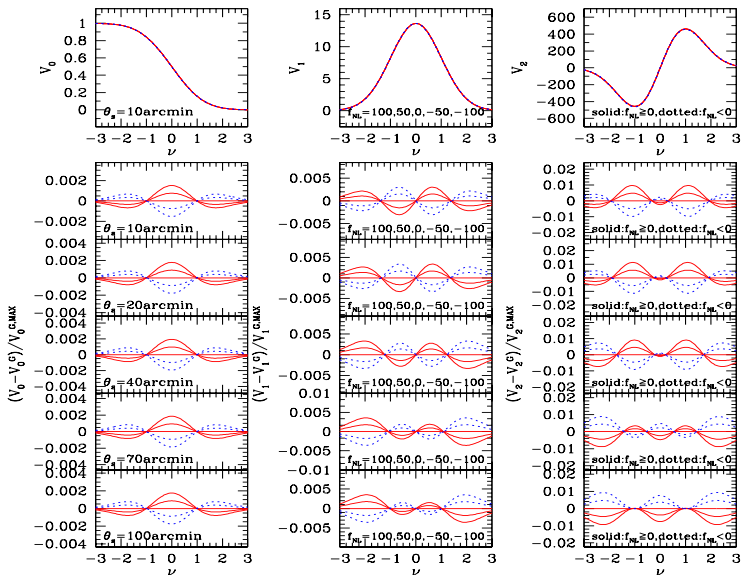
ΔG functional dependence on smoothing scale

- ▶ ΔG increases mildly with smoothing scale, at the scales we have probed :



Genus for f_{NL}

Hikage, Komatsu and Matsubara (2006)



Comparing g_{NL} and f_{NL} non-Gaussianities

Characteristic	f_{NL}	g_{NL}
Number of spots	if $f_{NL} > 0$, increases hot spots and decreases cold spots. Vice versa if $f_{NL} < 0$	$g_{NL} > 0$ increases both hot and cold spots. Vice versa if $g_{NL} < 0$
Shape of ΔG	symmetric	anti-symmetric
Dependence on f_{NL} or g_{NL}	linear	linear
Dependence on smoothing	strongly dependent	mildly dependent

Observables derived from genus

- ▶ $G(\nu)$ at different ν values are strongly (anti-)correlated.
- ▶ Can think of derived observables which will maximize the non-Gaussian deviations and also the difference between f_{NL} and g_{NL} .
- ▶ They can then be used to compare with observations to constrain f_{NL} and g_{NL} .

Observables derived from genus

List four observables:



$$R_{\text{cold}} \equiv \frac{N_{\text{cold}}}{N_{\text{cold}}^{\text{G}}}, \quad R_{\text{hot}} \equiv \frac{N_{\text{hot}}}{N_{\text{hot}}^{\text{G}}}.$$

$$N_{\text{cold}} \equiv \int_{-\nu_2}^{-\nu_1} d\nu G(\nu), \quad N_{\text{hot}} \equiv \int_{\nu_1}^{\nu_2} d\nu G(\nu)$$

$$N_{\text{cold}}^{\text{G}} \equiv \int_{-\nu_2}^{-\nu_1} d\nu G^{\text{fit}}(\nu), \quad N_{\text{hot}}^{\text{G}} \equiv \int_{\nu_1}^{\nu_2} d\nu G^{\text{fit}}(\nu).$$

Gaussian : $R_{\text{cold}} = 1, R_{\text{hot}} = 1$

Choose $\nu_1 = 1, \nu_2 = 2.5$,

If $g_{\text{NL}} > 0$, $R_{\text{cold}} < 1, R_{\text{hot}} < 1$

If $g_{\text{NL}} < 0$, $R_{\text{cold}} > 1, R_{\text{hot}} > 1$

Observables derived from genus



$$R_{\text{spots}} = \frac{N_{\text{cold}} + N_{\text{hot}}}{N_{\text{cold}}^{\text{G}} + N_{\text{hot}}^{\text{G}}}$$

Gaussian : $R_{\text{spots}} = 1$.

$$\text{If } g_{\text{NL}} > 0, R_{\text{spots}} < 1$$

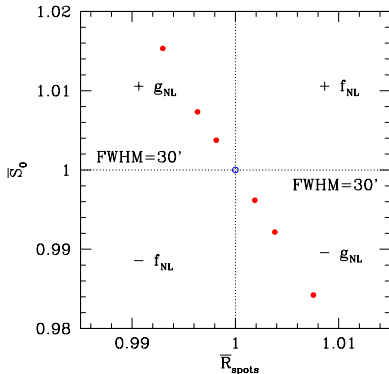
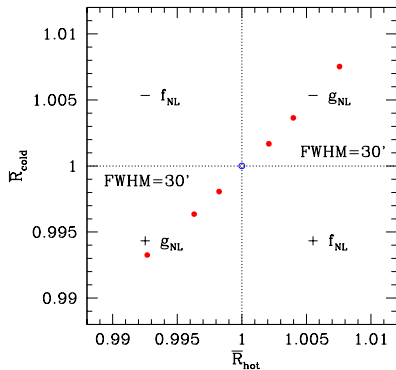
$$\text{If } g_{\text{NL}} < 0, R_{\text{spots}} > 1$$

- ▶ S_0 : ratio of slope of non-Gaussian genus curve to fitted Gaussian at $\nu = 0$.

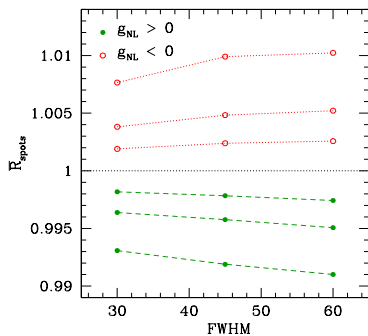
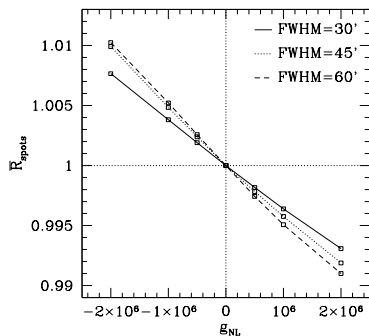
$$\text{If } g_{\text{NL}} > 0, S_0 > 1$$

$$\text{If } g_{\text{NL}} < 0, S_0 < 1$$

R_{cold} versus R_{hot} and R_{spots} versus S_0



Functional dependence of R_{spots} on g_{NL} and smoothing scale



Summary

- ▶ Simulated non-Gaussian CMB maps arising from primordial perturbations upto cubic order.
- ▶ Studied the statistical nature of the non-Gaussian effects on the CMB.
- ▶ Measured genus using the simulations and studied how they vary as a function of g_{NL} and smoothing scale.
- ▶ We showed f_{NL} and g_{NL} have very different signatures in the genus and other derived observables and can be easily distinguished.
- ▶ No observational contaminants such as galaxy mask, point sources, noise, etc were added in this work. Need to include them to compare with real data.
- ▶ We are now working on constraining g_{NL} by comparing the simulations with WMAP 5 year data.