Self-Calibrated Cluster Counts as a Probe of Primordial Non-Gaussianity

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We show that the ability to probe primordial non-Gaussianity with cluster counts is drastically improved by adding the excess variance of counts which contains information on the clustering. The conflicting dependences of changing the mass threshold and including primordial non-Gaussianity on the mass function and biasing indicate that the self-calibrated cluster counts well break the degeneracy between primordial non-Gaussianity and the observable-mass relation. Based on the Fisher matrix analysis, we show that the count variance improves constraints on $f_{\text{NL}}$ by more than an order of magnitude. It exhibits little degeneracy with dark energy equation of state. We forecast that upcoming Hyper Suprime-cam cluster surveys and Dark Energy Survey will constrain primordial non-Gaussianity at the level $\sigma(f_{\text{NL}}) \approx 8$, which is competitive with forecasted constraints from next-generation cosmic microwave background experiments.

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The measurement of departures from Gaussianity of the initial perturbations provides a unique opportunity to probe the early universe [1]. While the standard single field slow-roll inflation models predict primordial perturbations very close to Gaussian, some models such as multi-field models and the curvaton model can produce the level of non-Gaussianity high enough to be detected in ongoing or future surveys. Thus specific forms of primordial non-Gaussianity contain valuable information on how the initial density fluctuations are generated.

Observationally, primordial non-Gaussianity has mainly been studied using the temperature fluctuation of the Cosmic Microwave Background (CMB). Recently it has attracted considerable attention given a possible detection of non-Gaussianity by Yadav and Wandelt [2]. However, the detection of non-Gaussianity in the CMB is somewhat controversial in the sense that independent analyses yield slightly different results [3], suggesting the importance of other observational probes independent of the CMB. Another powerful probe of primordial non-Gaussianity is provided by the large-scale structure of the universe. In particular, non-Gaussianity induces a scale-dependent halo bias [4–8], and thus by studying large-scale distributions of astronomical objects one can obtain tight constraints that are competitive with the CMB. Constraints from the large-scale structure are also important given that non-Gaussianity can be scale-dependent such that deviations from Gaussian are larger at smaller scales [9].

Primordial non-Gaussianity is also sensitive to the abundance of massive clusters and its redshift evolution [4, 10]. An advantage of using massive clusters is its reasonable one-to-one correspondence with dark halos, which suggests that halo assembly bias (e.g., [6]) is less important. A challenge here is how to calibrate cluster masses; since the cluster mass is not directly observable, one has to resort to well-calibrated correlations between cluster masses and observable quantities such as luminosities, temperatures, and the numbers of member galaxies in order to infer cluster masses. The observable-mass relations always involve uncertainties, suggesting that the change of cluster abundances by primordial non-Gaussianity may be compensated by modifying the relation between observables and masses. Therefore constraints from cluster counts rely on how well we can calibrate such observable-mass relations.

In this Letter, we point out that clustering information breaks the degeneracy and allows us to determine primordial non-Gaussianity surprisingly well with cluster counts. This is because the clustering bias for massive clusters is quite sensitive to both cluster masses and primordial non-Gaussianity, and more importantly, because the cluster abundance and biasing show conflicting dependences on these. Such self-calibrated cluster count technique has been discussed extensively in the context of accurate dark energy probes [11–14], but its use for primordial non-Gaussianity has not been explored.

Here we quantify non-Gaussianity of the local form using the standard parametrization, $\Phi = \phi + f_{\text{NL}} (\phi^2 - \langle \phi^2 \rangle)$, where $\Phi$ is the curvature perturbation and $\phi$ is an auxiliary random-Gaussian field. The parameter $f_{\text{NL}} > 0$ (< 0) indicates that the initial density field is positively (negatively) skewed. The current level of constraints on primordial non-Gaussianity is $|f_{\text{NL}}| \lesssim O(100)$ [2, 3, 6]. We adopt a non-Gaussian correction factor of the cluster mass function based on the Edgeworth expansion [9]:

$$\frac{dn/dM}{dn_G/dM} = 1 + \frac{\sigma_S^3}{6} (6 \nu^2 - 3 \nu) - \frac{1}{6} \frac{d\sigma S_3}{d\ln \nu} \left( \nu - \frac{1}{\nu} \right),$$

where $\nu = \delta_c/\sigma$, $\delta_c \approx 1.68$ is the critical linear overdensity, $\sigma = \sigma(M, z)$ is the linear fluctuation on the mass scale of $M$ which we compute using the transfer function $T(k)$ presented by Eisenstein and Hu [15] ignoring the baryon wiggle. We adopt models of Warren et al. [16].

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for the mass function in the Gaussian case, $dn_G/dM$. The skewness $S_3$ is related to $f_{NL}$ as [8]

$$
\sigma S_3 = \frac{f_{NL}}{\sigma^3} \int_0^\infty \frac{dk_1}{k_1} a(k_1) W(M, k_1) \Delta^2_\phi(k_1) \\
\times \int_0^\infty \frac{dk_2}{k_2} a(k_2) W(M, k_2) \Delta^2_\phi(k_2) \\
\times \int_{-1}^1 d\mu a(\mu) W(M, k) \left[ 1 + \frac{P_\phi(k)}{P_\phi(k_1)} + \frac{P_\phi(k)}{P_\phi(k_2)} \right]
$$

where $\Delta^2_\phi = k^3P_\phi(k)/2\pi^2$ is the power spectrum of the curvature perturbation, $\alpha = [2D(z)T(k)/3\Omega_M](ck/H_0)^2$, $D(z)$ is the linear growth rate normalized to $(1+z)^{-1}$ in the matter-dominant era, and $k^2 = k_1^2+k_2^2+2\mu k_1 k_2$. For the window function $W(M,k)$ we adopt the real space top-hat filter. In practice we use the following fitting formula for $\sigma S_3$:

$$
\sigma S_3 \approx (8.66 \times 10^{-5}) f_{NL} \frac{\Omega_M D(0)}{\mathcal{D}_0^2} \Gamma^{-1.4} \sigma_8 \\
\times m_{10}^{0.0272 - 0.11(n_s - 0.96) - 0.0008 \ln \frac{m_{10}}{10} + 1},
$$

with $m_{10} = [M/(10^{10}h^{-1}M_\odot)]/\Gamma^3(\Omega_M h^2)^{-1}$ and $\Gamma = \Omega_M h \exp[-\Omega_0(1+\sqrt{2}\pi/\Omega_M)]$ is so-called the shape parameter [17]. This fitting formula should be accurate at a few percent level in the mass scale range $10^7 h^{-1}M_\odot \lesssim M \lesssim 10^{15} h^{-1}M_\odot$.

The non-Gaussian correction of the halo bias is computed as [8]

$$
\Delta b(M, z, k) = \frac{2f_{NL}\delta_c}{\alpha} (b_G - 1) - \frac{\nu}{\delta_c} \frac{d}{dv} \left( \frac{dn/dM}{dn_G/dM} \right)
$$

The halo bias in the Gaussian case, $b_G$, is assumed to be the form presented by Sheth and Tormen [18].

Fig. 1 illustrates the reason why the clustering information is so important. As shown in the Figure, including positive $f_{NL}$ increases the number of clusters above some mass threshold $M_{th}$. This increment can be compensated by raising $M_{th}$. However, these two models with the same numbers of clusters result in quite different halo biases because both raising $M_{th}$ and $f_{NL}$ increase the biasing. Thus by including clustering information we can strongly break the degeneracy between $M_{th}$ and $f_{NL}$, and can obtain tight constraints on $f_{NL}$.

We now forecast constraints on $f_{NL}$ from future cluster surveys. We include the clustering information using a count-in-cell analysis. Specifically we approximate the Fisher matrix as [12, 13]

$$
F_{\alpha\beta} = m^T \alpha C^{-1} m_\beta + \frac{1}{2} \text{Tr} [C^{-1} S_{\alpha} C^{-1} S_{\beta}] + \frac{\delta_{\alpha\beta}}{\sigma^2_\alpha}
$$

where $\sigma_\alpha$ represents the prior information on each parameter, and the covariance matrix is given by $C \equiv S + \text{diag}(M)$.

![FIG. 1: The illustration of self-calibrating cluster counts to probe primordial non-Gaussianity $f_{NL}$. Here the redshift is $z = 1$. In cluster surveys, we basically obtain the number of clusters above some mass threshold $M_{th}$. The plot indicates the increase of the mass function $dn/dM$ due to positive $f_{NL}$ can be compensated by increasing the $M_{th}$. However, while the two models predict the same number of clusters, the corresponding halo bias $b(M)$ (here we adopted $k = 0.02h\text{Mpc}^{-1}$) are quite different because both raising $M_{th}$ and $f_{NL}$ increase $b(M)$ (see arrows). Although in this plot the scatter in the observable-mass relation is ignored for simplicity, it will be included in the Fisher matrix analysis.](image-url)
information from the variance of the counts which contain clustering (biasing) information. Using the Fisher matrix, one can estimate a marginalized error on each parameter as $\sigma(\alpha) = \sqrt{(F^{-1})_{\alpha \alpha}}$.

We calculate the Fisher matrix in 10-dimensional parameter space; 6 standard cosmological parameters including dark energy equation of state (the matter density $\Omega_{\text{m}} h^2$, the baryon density $\Omega_{\text{b}} h^2$, the power spectrum tilt $n_s$, the normalization of the power spectrum $\delta_c$ [19], the dark energy density $\Omega_{\text{DE}}$, and dark energy equation of state $w$), 1 parameter representing primordial non-Gaussianity ($f_{\text{NL}}$), and 3 parameters from the observable-mass relation, $\sigma_{\ln M}$ and $\ln M_{\text{bias},0} + \gamma \ln(1+z)$. The Five-Year Wilkinson Microwave Anisotropy Probe (WMAP5) result for $\Lambda$CDM [20], $(\Omega_{\text{m}} h^2, \Omega_{\text{b}} h^2, n_s, \delta_c, \Omega_{\text{DE}}, w)=(0.133, 0.0227, 0.963, 4.61 \times 10^{-5}, 0.742, -1)$, is adopted as our fiducial cosmological model. We add conservative priors to the first 4 parameters, $\sigma_p(\Omega_{\text{m}} h^2) = 0.006$, $\sigma_p(\Omega_{\text{b}} h^2) = 0.0006$, $\sigma_p(n_s) = 0.015$, and $\sigma_p(\delta_c) = 10^{-6}$; these are the level of accuracies which has already been achieved by WMAP5. In addition, our fiducial model has $f_{\text{NL}} = 0$, $\sigma_{\ln M} = 0.25$, ln $M_{\text{bias},0} = 0$, and $\gamma = 0$.

For illustrative purposes, we consider the following three upcoming surveys: Hyper Suprime-Cam on Subaru telescope (HSC; since the design of the HSC cluster survey is still tentative, we consider both 1000 deg$^2$ and 2000 deg$^2$), Dark Energy Survey (DES; 5000 deg$^2$) [21], and Large Synoptic Survey Telescope (LSST; 20000 deg$^2$) [22]. We adopt a simplified assumption that these optical surveys will find clusters with $M_{\text{obs}} > 10^{15.7} h^{-1} M_{\odot}$ out to $z_{\text{max}} = 1.4, 1.0, \text{and } 1.7$, respectively. For the count-in-cell analysis, we use the cell size of 20 deg$^2$ and the redshift interval $\Delta z = 0.1$. Three mass bins with spacing of $\Delta \log M_{\text{obs}} = 0.5$ are also adopted.

In Fig. 2, we show marginalized constraints on $f_{\text{NL}}$ and the correlations with the parameters $\sigma_{\ln M}$, expected for the 2000 deg$^2$ HSC cluster survey. As expected, constraints are drastically improved by combining the number counts with the variance which includes the clustering information. This is partly because constraints from number counts and variance show different degeneracy directions, suggesting that both the number counts and clustering are essential for accurate determinations of $f_{\text{NL}}$.

Table I summarizes forecasted constraints on various cosmological parameters. For all the upcoming surveys, the count variance drastically enhances the ability to probe primordial non-Gaussianity, by more than an order of magnitude improvement in $f_{\text{NL}}$ compared with the number counts alone. Predicted marginalized errors of $\sigma(f_{\text{NL}}) \sim 8$ for HSC and DES and $\sim 2$ for LSST are competitive with constraints from next-generation CMB experiments (e.g., [23]) and galaxy power spectrum measurements (e.g., [4, 24]). The variance helps to regulate the observable-mass relation, improving an accuracy of $\ln M_{\text{bias},0}$ and $\sigma_{\ln M}$ measurements by a factor of two or more. Measurements of dark energy equation of state are improved as well, which is consistent with earlier work. We find that $w$ and $f_{\text{NL}}$ are not correlated very much, indicating that we can well determine these two parameters simultaneously using self-calibrated cluster counts. Fig. 3 shows contours of $\sigma(f_{NL})$ as a function of the survey area and the maximum redshift. The expected constraints on $f_{\text{NL}}$ is a steep function of the maximum redshift even at $z > 1$, which indicate the importance of the deep surveys to detect clusters out to $z > 1$.

We have shown that adding clustering information from the count variance drastically improves measurements of primordial non-Gaussianity with cluster counts. Although the calibration of cluster masses limits the use of cluster counts as a cosmological probe, the self-calibration technique allows us to determine both the observable-mass relation and $f_{\text{NL}}$ simultaneously. The significant effect of the count variance comes from the conflicting dependences of the mass threshold and $f_{\text{NL}}$ on the cluster mass function and biasing parameter (Fig. 1). Allowing dark energy equation of state to vary does not degrade $f_{\text{NL}}$ measurements very much. Resulting forecasted constraints on $f_{\text{NL}}$, $\sigma(f_{\text{NL}}) \sim 8$ for HSC and DES and $\sim 2$ for LSST, suggest that cluster counts can become a competitive probe compared to the CMB or the large-scale galaxy power spectrum.

We have here made a number of simplified assumptions. For instance, it is important to check how the possible redshift evolution of the observable-mass relation

![FIG. 2: Expected marginalized constraints on the $f_{\text{NL}}$-$\sigma_{\ln M}$ plane from the upcoming 2000 deg$^2$ HSC cluster survey. WMAP5 cosmology is assumed as a fiducial model. Contours indicate the 68% confidence regions from the number counts (dotted), the variance of the counts which contains the clustering information (dash-dotted), and the combination of the number counts and variance (solid).](image-url)
TABLE I: Marginalized constraints on cosmological parameters estimated from the Fisher matrix analysis using the number counts and/or the variance of counts (clustering). WMAP5 cosmology is assumed as a fiducial model. Constraints in four future survey parameters, HSC1 (1000 deg$^2$, $z_{\text{max}} = 1.4$), HSC2 (2000 deg$^2$, $z_{\text{max}} = 1.4$), DES (5000 deg$^2$, $z_{\text{max}} = 1.0$), and LSST (20000 deg$^2$, $z_{\text{max}} = 1.7$), are presented.

<table>
<thead>
<tr>
<th>Survey</th>
<th>Counts $\sigma(\Omega_{DE})$ $\sigma(w) \sigma(f_{NL})$</th>
<th>Variance $\sigma(\Omega_{DE})$ $\sigma(w) \sigma(f_{NL})$</th>
<th>Counts + Variance $\sigma(\Omega_{DE})$ $\sigma(w) \sigma(f_{NL})$</th>
</tr>
</thead>
<tbody>
<tr>
<td>HSC1</td>
<td>0.030 0.151 240.4</td>
<td>0.012 0.189 37.8</td>
<td>0.010 0.103 8.1</td>
</tr>
<tr>
<td>HSC2</td>
<td>0.023 0.108 188.2</td>
<td>0.011 0.142 28.0</td>
<td>0.009 0.074 6.2</td>
</tr>
<tr>
<td>DES</td>
<td>0.032 0.081 210.6</td>
<td>0.011 0.102 35.3</td>
<td>0.009 0.055 7.9</td>
</tr>
<tr>
<td>LSST</td>
<td>0.009 0.037 106.1</td>
<td>0.006 0.051 6.9</td>
<td>0.005 0.024 1.9</td>
</tr>
</tbody>
</table>

FIG. 3: Expected marginalized constraints on $f_{NL}$ as a function of two survey parameters, the survey area and the maximum redshift $z_{\text{max}}$. Contours are drawn per 0.25 dex. Four survey parameters considered in this paper are indicated by filled squares.

affects our results [12, 13]. The impact of other systematics, such as cluster photometric redshifts [14] and the effect of halo assembly bias [25], should be addressed. On the other hand, we used only the count variance of each cell as the clustering information. Since the effect of primordial non-Gaussianity is more significant at larger scales, including the count covariance (or including the full clustering information with the power spectrum [11]) may improve the constraints further. We leave such more comprehensive treatments for future work.

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