COSMOLOGICAL CONSTRAINTS FROM SUNYAEV–ZELDOVICH CLUSTER COUNTS: AN APPROACH TO ACCOUNT FOR MISSING REDSHIFTS

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ABSTRACT

The accumulation of redshifts provides a significant observational bottleneck when using galaxy cluster surveys to constrain cosmological parameters. We propose a simple method to allow the use of samples where there is a fraction of the redshifts that are not known. The simplest assumption is that the missing redshifts are randomly extracted from the catalog, but the method also allows one to take into account known selection effects in the accumulation of redshifts. We quantify the reduction in statistical precision of cosmological parameter constraints as a function of the fraction of missing redshifts for simulated surveys, and also investigate the impact of making an incorrect assumption for the distribution of missing redshifts.

Key words: cosmological parameters – cosmology: observations – galaxies: clusters: general – large-scale structure of universe – methods: data analysis

Online-only material: color figures

1. INTRODUCTION

Galaxy clusters have been used to make inferences about cosmology using a number of different approaches (Allen et al. 2011; Voit 2005). One of these is to use the number of clusters as a function of redshift to measure cosmological parameters, most notably the matter density relative to critical, \( \Omega_m \), and the amplitude of the power spectrum of density fluctuations on small scales, \( \sigma_8 \). Cluster samples can be selected using the optical richness (Rozo et al. 2010), X-ray luminosity (Vikhlinin et al. 2009) and the flux of the Sunyaev–Zeldovich (SZ) effect (Haiman et al. 2001; Holder et al. 2001; Battye & Weller 2003).

In this paper we will concentrate on cluster surveys selected using the SZ effect but much of what we say can be applied to other observables.

The SZ effect (Sunyaev & Zeldovich 1970) is the inverse Compton scattering of cosmic microwave background (CMB) photons as they pass through the hot intracluster gas along the line-of-sight between the observer and the last scattering surface of the CMB; see Birkinshaw (1999) for a review of properties of the SZ effect. The key feature in the context of constraining cosmological parameters is that the SZ flux is directly proportional to the integrated gas pressure and this is expected to be (Springel et al. 2001; da Silva et al. 2004; Motl et al. 2005; Nagai 2006; Kay et al. 2012) and observed to be (Marrone et al. 2012; Planck Collaboration et al. 2013a; Sifón et al. 2013) an excellent proxy for the total mass of the cluster.

Large samples of clusters selected via the SZ effect are presently being accumulated using the Atacama Cosmology Telescope (ACT; Hasselfield et al. 2013), the South Pole Telescope (SPT; Reichardt et al. 2013), and the Planck satellite (Planck Collaboration et al. 2013b). These have been used to constrain cosmological parameters and notably in Planck Collaboration et al. (2013c) a discrepancy between the measured values of \( \Omega_m \) and \( \sigma_8 \) from clusters and the primary CMB (Planck Collaboration et al. 2013d) has been identified. In order to make further progress, two things are necessary. First, the understanding of the relationship between SZ flux and the total mass of a cluster needs to be better understood. In addition, much larger samples of clusters need to be used to reduce the statistical uncertainties and this requires much more extensive knowledge of cluster redshifts. At present this creates a bottleneck in the process since each cluster has to be targeted individually with follow-up observations. Large-scale surveys, such as PANSTARRS,1 Dark Energy Survey (DES),2 LSST,3 and Euclid4 might ameliorate this over significant regions of the sky, but even then it is likely that there will not be complete redshift coverage.

The objective of this paper is to develop a simple method for dealing with samples of clusters for which there is a fraction of objects (up to 20%, say) with missing redshifts. The simplest possible idea would be to assume that the observed redshift distribution of clusters is that of the overall sample and multiply up the observed numbers so as to get the total number of clusters correct. However, this is too strong an assumption. The procedure we propose is to allow clusters whose redshifts are not known to be chosen from a statistical ensemble depending on cosmology. Clearly this will lead to some increase in the statistical errors bars, but this can be traded off by the reduction in the sample size that would be necessary to obtain a pure sample. We will discuss this in the context of simulated surveys that are designed to be similar to those being performed by ACT, SPT, and Planck. In addition, it may be that there is some selection effect associated with the follow-up campaign, resulting in a systematic difference in the counts for known and unknown samples. If the selection effect is known then it can easily be corrected within the proposed method. However, this may not necessarily be the case and we discuss the impact of making an incorrect assumption.

2. METHOD

We write the predicted number of galaxy clusters as a function of redshift for a given survey as

\[
\frac{dN}{dz} = \int d\Omega \int \frac{dV}{dz} \frac{d\Omega}{dV} \int_{M_{\text{lim}}(z)}^{\infty} dM \frac{dn}{dM}
\]

1 http://pan-starrs.ifa.hawaii.edu
2 http://www.darkenergysurvey.org
3 http://www.lsst.org
4 http://www.euclid-ec.org
where $d\Omega$ is the solid angle element, $dV/d\Omega$ is the volume element, and $dn/dM$ is the mass function. The limiting mass $M_{\text{lim}}(z)$ describes the selection function of the survey. Using the SZ signal as a mass proxy, this can be related to a limit in flux density $S_{\text{lim}}(z)$. A more realistic SZ selection function (e.g., Planck Collaboration et al. 2013c) is expressed in terms of a limit in detection significance as a function of the size of the cluster and the position in the sky. However, such a detailed description is beyond the scope of this paper.

By defining $N_b$ suitable redshift bins, of width $\Delta z$, we can write the predicted number of clusters in the $i$th bin as

$$N_{\text{th}}(i) = \int_{z_i - \Delta z/2}^{z_i + \Delta z/2} dz \frac{dN}{dz},$$

(2)

which depends on cosmology.

The likelihood we use is

$$\ln P = \sum_{i=1}^{N_b} \left[ N_{\text{cat}}(i) \ln(N_{\text{th}}(i)) - N_{\text{th}}(i) - \ln(N_{\text{cat}}(i)!) \right],$$

(3)

which gives the probability of finding $N_{\text{cat}}(i)$ clusters in each of $N_b$ bins given an expected number of $N_{\text{th}}(i)$, assuming they follow a Poisson distribution (Cash 1979). When appropriate, other likelihood functions can be used to account for additional effects, such as spatial correlations between galaxy clusters (e.g., Hu & Cohn 2006 and references therein).

Computing the likelihood, being that of Equation (3) or a more general one, requires that all redshifts in the catalog are known. Let us now assume that this is not the case, i.e., the catalog contains $N_{\text{tot}}$ clusters, of which $N_{\text{red}}$ have redshifts and $N_{\text{miss}}$ do not. The minimal strategy (adopted by Planck Collaboration et al. 2013c) is to multiply the distribution of known redshifts by $N_{\text{tot}}/N_{\text{red}}$ and use it as $N_{\text{cat}}$. This assumes that the distribution of missing redshifts is exactly the same as the known ones, which would not take into account any possible statistical uncertainty. Another concern is that this strategy underestimates the error bars, which are proportional to the number of clusters in the catalog. Both issues were negligible in Planck Collaboration et al. (2013c), as there was only one missing redshift out of 189 ($\sim0.5\%$); however, here we want to deal with larger fractions of missing redshifts.

An improvement over this minimal strategy would be to use the sub-sample with known $z$ to estimate the redshift distribution of the full sample by means of resampling techniques, such as bootstrap methods (see, e.g., Efron 1982). This strategy would capture statistical uncertainties; however, propagating the uncertainty on the resulting estimate of the full distribution to the cosmological parameters would still be an outstanding issue. The approach we advocate avoids this problem by modeling the corrections directly into the likelihood function. We modify the likelihood of Equation (3) to be

$$\ln P = \ln \left( \frac{1}{N} \sum_{j=1}^{N} P_j \right),$$

$$\ln P_j = \sum_{i=1}^{N_b} \left[ N_{\text{cat}}^j(i) \ln(N_{\text{th}}(i)) - N_{\text{th}}(i) - \ln(N_{\text{cat}}^j(i)!) \right],$$

(4)

where $N_{\text{cat}}^j(i)$ is the $j$th version of the catalog counts obtained by complementing the $N_{\text{red}}$ known redshifts with a random realization of the $N_{\text{miss}}$ unknown redshifts. We stress that, although in this work we are using the likelihood in Equation (3), it would be straightforward to modify any likelihood function in the same way. In practice, at a given point in the cosmological parameters space we average the probability corresponding to different catalog counts, where the missing redshifts are randomly generated. If we consider a large enough number of realizations $N$, the likelihood becomes close to deterministic.

As a distribution for drawing the $N_{\text{miss}}$ missing redshifts we can use the (normalized) theoretical counts, $N_{\text{th}}(z)$, corresponding to the cosmological parameters that are currently being sampled. We note that this does not depend on the fiducial cosmology. In Section 4.3 we introduce a modification to this simple recipe that can be used to account for systematic effects.

We can interpret the average over the probability in Equation (4) as a marginalization over different versions of the catalog. Such marginalization correctly inflates the error bars for the cosmological parameters according to the uncertainty in the redshift distribution of clusters. With respect to simply rescaling the distribution of known redshifts, we get more realistic error bars and reduce possible biases, because we are making a milder assumption on the distribution of missing redshifts.

In this work we do not address the purity of the sample of $N_{\text{miss}}$ clusters. Because false detections are typically identified with the follow-up observations, the sample of $N_{\text{miss}}$ clusters could be less than 100% pure. This could be accounted for statistically, by correcting $N_{\text{miss}}$ for the predicted number of false detection. The purity of the sample could either be estimated with simulations or derived from that of the sample for which follow-up observations have been made. Uncertainties on the purity could be accounted for by varying $N_{\text{miss}}$ according to its predicted distribution during the estimation of cosmological parameters.

### 3. Simulations

In order to test the method, we adopted the fiducial cosmology defined by the parameters of Table 1. We verified that this choice does not impact the results of the validation. In the following, we will denote with $N_{\text{th}}^i(i)$ the predicted counts for the fiducial cosmology. We computed the constraints only for the $\Omega_m$ and $\sigma_8$ parameters while keeping the other parameters fixed at the fiducial values. This could make our results slightly stronger than those we would expect in reality; nonetheless, the trends we observe should be representative. The estimations have been performed for 0, 5%, 10%, and 20% missing redshifts.

We considered two simulated surveys, as detailed below, using either a limit in flux density or a limit in mass as a selection function. To convert from the SZ flux, $Y_{500}$, to cluster mass, $M_{500}$, integrated within the radius where the mean enclosed density is 500 times the critical density, we adopted a relation of the form

$$Y_{500} = Y_e \left( \frac{M_{500}}{3 \times 10^{14} h^{-1} M_\odot} \right)^\alpha \left( \frac{E(z)}{500 \text{ Mpc}} \right)^{2/3} \left( \frac{d_A}{500 \text{ Mpc}} \right)^{-2},$$

(5)

| $\Omega_m$ | $\Omega_b$ | $\Omega_{\Lambda}$ | $\sigma_8$ | $n_s$ | $H_0$ (km s$^{-1}$ Mpc$^{-1}$) |
|-----------|-----------|------------------|-----------|-------|-------------------------------|
| 0.3       | 0.04      | 0.7              | 0.8       | 1     | 70                           |

![Table 1: Fiducial Cosmological Parameters of the Simulation](image)
where \( d_A(z) \) is the angular diameter distance to redshift \( z \), 
\[ E^2(z) = \Omega_m (1+z)^3 + \Omega_{\Lambda}, \quad Y_e = 8.9 \times 10^{-4} \text{ and } \alpha = 1.79. \]
This is very similar to the one used in Planck Collaboration et al. (2013c). We used the mass function by Tinker et al. (2008).

**Survey 1.** The first survey attempts to mimic the catalog expected for Planck at signal-to-noise ratio (S/N) > 5. It has a large sky coverage (65% of the sky, 26814.4 deg\(^2\)) and a constant flux limit of \( Y_{500, \text{lim}} = 3.5 \times 10^{-3} \). The corresponding mass limit \( M_{500, \text{lim}} \) can be obtained from Equation (5) and it depends both on redshift and cosmology. The predicted counts for the fiducial cosmology \( N^0_{\text{red}}(z) \) are shown as the solid line in Figure 1. The total number of clusters between \( z = 0 \) and \( z = 5 \) is 516.3. As a term of comparison, we also considered a higher S/N selection \( (Y_{500} = 6.5 \times 10^{-3}) \) with no missing redshifts. This yields 201.4 clusters and can be thought of as being similar to the S/N = 7 cosmological sample used in Planck Collaboration et al. (2013c).

**Survey 2.** The second survey mimics the SPT–SZ complete survey, which will contain around 500 clusters (Reichardt et al. 2013), and is similar also to that performed by ACT. It has a large sky coverage (65% of the sky, 26814.4 deg\(^2\)) and a constant flux limit of \( Y_{500, \text{lim}} = 3.5 \times 10^{-3} \). The corresponding mass limit \( M_{500, \text{lim}} \) can be obtained from Equation (5) and it depends both on redshift and cosmology. The predicted counts for the fiducial cosmology \( N^0_{\text{red}}(z) \) are shown as the solid line in Figure 1. The total number of clusters between \( z = 0 \) and \( z = 5 \) is 516.3. As a term of comparison, we also considered a higher S/N selection \( (Y_{500} = 6.5 \times 10^{-3}) \) with no missing redshifts. This yields 201.4 clusters and can be thought of as being similar to the S/N = 7 cosmological sample used in Planck Collaboration et al. (2013c).

For each survey, we computed the total catalog counts, \( N_{\text{tot}}(z) \), directly as the fiducial counts \( N^0_{\text{red}}(z) \). As pointed out by Sahlén et al. (2009), using such an “average catalog,” as opposed to a particular realization, allows an unbiased recovery of the input cosmological parameters. This would highlight eventual biases due to our procedure for dealing with the missing redshifts. The counts obtained in this way are used as actual catalog counts, including Poisson error bars.

Starting from \( N_{\text{tot}}(z) \), we model the distribution of known redshifts, \( N_{\text{red}}(z) \) as
\[ N_{\text{red}}(z) = W(z) N_{\text{tot}}(z). \]

The function \( W(z) \) is a generic function of redshift normalized such that
\[ \int_0^{z_{\text{max}}} W(z) N_{\text{tot}}(z) dz = N_{\text{red}} \]
and represents a selection effect for the known redshifts. In general, \( W(z) \) depends on the SZ survey and on the way the redshifts are collected.

The true distribution of missing redshifts, \( N_{\text{miss}}(z) \), is clearly
\[ N_{\text{miss}}(z) = N_{\text{tot}}(z) - N_{\text{red}}(z) = [1 - W(z)] N_{\text{tot}}(z). \]
However, as this distribution is not known, we use \( N_{\text{red}}(z) \) to generate realizations of it, \( N_{\text{red}}^j(z) \), and compute \( N_{\text{cat}}^j(z) \) for the likelihood of Equations (4) and (5) as \( N_{\text{cat}}^j(z) = N_{\text{miss}}^j(z) + N_{\text{red}}(z) \).

### 4. RESULTS

In this section we compare the cosmological constraints predicted for \( \Omega_m \) and \( \sigma_8 \) for Survey 1 and Survey 2 as a function of the number of missing redshifts. We first consider the case where the missing redshifts are excluded randomly from the catalogs (Section 4.1), and then consider the impact of selection effects and ways to account for them (Sections 4.2 and 4.3).

#### 4.1. No Selection Effects

If there is no selection effect for the known redshifts, \( W(z) = \text{const} \) and \( N_{\text{red}}(z) \) is simply \( N_{\text{tot}}(z) \) rescaled by \( N_{\text{red}}/N_{\text{tot}} \). As a consequence, the assumption made by our method regarding the distribution of the missing redshifts is verified. The cosmological constraints are shown in Figure 2. The increase in the one-dimensional (1D) error bars with respect to the \( N_{\text{miss}} = 0 \) case is moderate (up to 30%–35% for \( N_{\text{miss}} = 0.2 N_{\text{tot}} \)). Biases on \( \Omega_m \) and \( \sigma_8 \) are much smaller than \( 1\sigma \) for low \( N_{\text{miss}} \), and reach \( 0.2\sigma \) for \( N_{\text{miss}} = 20\% \).

For Survey 1, the results with up to 20% missing redshifts are significantly better than the high S/N case with no missing redshifts, shown as a dashed area in the left panel of Figure 2. The latter yields \( \Omega_m = 0.301 \pm 0.019 \) and \( \sigma_8 = 0.798 \pm 0.023 \); this is somewhat stronger than the results presented in Planck Collaboration et al. (2013c) because in our case we fix all parameters except \( \Omega_m \) and \( \sigma_8 \).

#### 4.2. Mild Versus Strong Selection Effects

We now consider the impact of a nontrivial selection \( W(z) \neq \text{const} \) in order to quantify the biases in the cosmological parameters when such an effect is neglected. We considered two opposite situations: a mild selection \( (W(z) \) varying slowly with redshift) and a strong selection \( (W(z) \) varying rapidly with redshift). Both are intended to be indicative rather than representative. In both cases we adopted decreasing functions of the redshift because low-\( z \) clusters are easier to detect and observe.

For the mild selection we used a linear function of the redshift, while as a strong selection we used a step function \( (W(z) \) varying slowly with redshift) and a strong selection \( (W(z) \) varying rapidly with redshift). Both are intended to be indicative rather than representative. In both cases we adopted decreasing functions of the redshift because low-\( z \) clusters are easier to detect and observe.

In Figures 4 and 5 we show the results on the cosmological parameters for Survey 1, those corresponding to Survey 2 being
very similar. Figure 4 shows the increase of the 68% confidence level (CL) error bar on \( \Omega_m \) and \( \sigma_8 \) and the absolute value of the bias on the best-fit parameters as a function of the number of missing redshifts. The size of the error bars can either increase or decrease moderately with respect to the case with no selection, depending on the number of missing redshifts and the choice of \( W(z) \). Selection effects mainly impact the best-fit value of the cosmological parameters, which will be biased. The bias is moderate (<1 \( \sigma \)) for the linear selection function, and much more pronounced (up to 4\( \sigma \)) for the step selection function.
Figure 5 shows the contours for $\Omega_m$ and $\sigma_8$ for Survey 1. Interestingly, the contours shift along the degenerate line. This suggests that the combination of $\Omega_m$ and $\sigma_8$ perpendicular to the degenerate line is almost insensitive to the selection effects. We looked for the well measured combination of the form $P = (\Omega_m/0.27)^\gamma \sigma_8$ corresponding to the minor axis of the ellipse. We computed the exponent $\gamma$ by performing a principal component analysis of the results with no missing redshifts. We obtained $\gamma = 0.45$ for Survey 1 and $\gamma = 0.20$ for Survey 2. The results in terms of this derived parameter $P$ are indeed much more robust against the systematic effects (the bias being at worse 1.4$\sigma$ instead of 4$\sigma$). This suggests that $P$ is a measure of the total number of clusters and the other principal component depends on the redshift distribution.

4.3. Modeling Selection Effects

The results of Section 4.2 support the conclusion that strong selection effects must be accounted for to avoid significant biases on $\Omega_m$ and $\sigma_8$. If $W(z)$ is known, this can be achieved by drawing the missing redshifts from a probability distribution given by $[1 - W(z)] N_{\text{th}}(z)$ instead of $N_{\text{th}}(z)$. We checked that, once modeled correctly into the analysis, the step selection function yields unbiased results, comparable with those shown in Figure 2.

When dealing with real surveys, the redshift accumulation strategy could provide a prior knowledge of $W(z)$. Alternatively, the selection function could be estimated from the data, for example, by measuring all the redshifts for a “deep” region of the survey and comparing the distribution $N_{\text{red}}(z)$ for this region with $N_{\text{th}}(z)$ for the rest of the survey. Assessing the impact of errors on $W(z)$ on the cosmological parameters is beyond the scope of this paper since it would be very survey specific. However, we suggest that a reasonable estimate of $W(z)$ would only leave mild residual selection effects, which would then have a limited impact on the results.

5. CONCLUSIONS

We have presented a statistical method for dealing with incomplete redshift coverage of SZ cluster catalogs during the estimation of cosmological parameters. We tested it on simulated data which mimic those of SPT/ACT and Planck. For a number of missing redshifts up to 20% and in the absence of systematic effects in the accumulation of redshifts, the recovery of the cosmological parameters is almost unbiased and there is only a mild increase in the 1D error bars (up to 35%).

Selection effects for the known redshifts, if not accounted for, can cause significant offsets in the cosmological parameters. The contours shift along the degenerate line, allowing a robust computation of the parameter perpendicular to it. We showed how known systematic effects can be modeled into the analysis to avoid biases in the cosmological parameters.

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