Impact of the calibration of the halo mass function on galaxy cluster number count cosmology

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ABSTRACT

The halo mass function (HMF) is a critical element in cosmological analyses of galaxy cluster catalogs. We quantify the impact of uncertainties in HMF parameters on cosmological constraints from cluster catalogs similar to those from Planck, those expected from the Euclid, Roman, and Rubin surveys, and from a hypothetical larger future survey. We analyze simulated catalogs in each case, gradually loosening priors on HMF parameters to evaluate the degradation in cosmological constraints. While current uncertainties on HMF parameters do not substantially impact Planck-like surveys, we find that they can significantly degrade the cosmological constraints for a Euclid-like survey. Consequently, the current precision on the HMF is not sufficient for Euclid, Roman, Rubin or possible larger surveys. Future experiments will have to properly account for uncertainties in HMF parameters, and it will be necessary to improve the precision of HMF fits to avoid weakening constraints on cosmological parameters.

Key words. galaxies: clusters: general – large-scale structure of Universe – cosmological parameters

1. Introduction

Galaxy clusters are unique tools for astrophysical and cosmological studies, ranging from galaxy formation to fundamental physics (Voit 2005; Allen et al. 2011; Kravtsov & Borgani 2012). In particular, it is now well established that their number density \( n(M,z) \) is a powerful cosmological probe, notably of the matter density, \( \Omega_m \), the root mean square density fluctuation, \( \sigma_8 \), the dark energy equation-of-state, and the sum of the neutrino masses (Weinberg et al. 2013). For this reason, cluster number counts figure among the four primary dark energy probes.

Surveys across different wavebands have used cluster number counts to derive cosmological constraints (e.g., Vikhlinin et al. 2009; Mantz et al. 2010; Rozo et al. 2010; Hasselfield et al. 2013; Planck Collaboration XXIV 2016; Bocquet et al. 2019; Costanzi et al. 2019, 2021). These analyses are based on the halo mass function (HMF) originating in the Press & Schechter (1974) formalism. The HMF provides the key link between cosmological parameters and the number density of clusters; it essentially bundles the complicated non-linear physics of halo formation into a simple analytical formula involving only the linear power spectrum and other linear-theory quantities. This is essential because it avoids running costly N-body simulations at each point in parameter space when establishing cosmological constraints.

Current cluster count analyses use catalogs of hundreds (in millimeter and X-ray data) to thousands (in optical data) of clusters, and they are not limited by our knowledge of the HMF. However, uncertainties on the HMF could become important as future catalogs grow by factors of 10–100. Euclid will detect tens of thousands of clusters (Sartoris et al. 2016) as will eROSITA (Hofmann et al. 2017), Simons Observatory (Ade et al. 2019), Rubin (Ivezić et al. 2019), and CMB-S4 (Abazajian et al. 2019). In the longer term, Roman (Gehrels et al. 2015), PICO (Hanany et al. 2019), and a Voyage 2050 Backlight mission (Basu et al. 2021) could detect from hundreds of thousands to one million clusters.

As statistical errors on cluster counts dramatically drop, effects that are now considered as second order will begin to noticeably contribute to the uncertainty budget of cosmological analyses (Fumagalli et al. 2021). In this paper, we examine the potential importance of uncertainties related to modeling errors in the HMF. Popular HMF formulas such as Tinker et al. (2008) predict cluster counts in the \( \Lambda \)-cold dark matter (\( \Lambda \)CDM) parameterization to a precision of several percent (\( \sim 5\% \)) (compared to full N-body simulations). The recent Despali et al. (2015) HMF provides an expression with parameters and associated errors fitted on \( \Lambda \)CDM N-body simulations. Figure 1 presents the relative difference of the Tinker and Despali mass functions at \( z = 1 \) as a function of the virial mass. The Tinker mass function is parameterized for different overdensity definitions (\( \Delta = 200, 300, 400, 600, \ldots \)), while Despali uses the virial overdensity \( \Delta_{\text{vir}}(z) \). To compare the two mass functions in Fig. 1, we extrapolated the value of the parameters of the Tinker mass function for \( \Delta_{\text{Tinker}} \equiv \Delta_{\text{vir}}(z = 1) \). The difference is below 5\% for masses less than a few \( 10^{14} M_\odot \), so within the quoted error bars for the Tinker mass function. The red band delineates the uncertainty from the HMF parameters provided by Despali et al. (2015). They are much smaller than the difference between the two mass functions, bringing to light the difference between the precision of a given HMF formula and its potential bias (a question of accuracy). The origin of this difference is beyond the scope of our study.
2. Methods

2.1. Mass function and cluster likelihood

The mass function formalism was originally introduced by Press and Schechter (Press & Schechter 1974). It predicts the comoving number density of halos, \( n(M,z) \), for a given set of cosmological parameters as

\[
\frac{dn}{d\ln M}(M,z) = \frac{\bar{\rho}}{M} \frac{d\ln(\sigma(M,z))}{d\ln M} f(\sigma(M,z))
\]

(1)

where \( \bar{\rho} \) is the mean density of the universe today, \( f(\sigma) \) is the multiplicity function (discussed hereafter) and

\[
\sigma^2(M,z) = \int \frac{1}{2\pi^2} P_{lin}(k,z) W^2(kR(M)) k^2 dk,
\]

(2)

with \( P_{lin}(k,z) \) the linear power spectrum at redshift \( z \), \( R(M) = \left[M/(4\bar{\rho}/3)\right]^{1/3} \) and \( W(kR) = [3/(kR)^2][\sin(kR)/(kR) - \cos(kR)] \) is the spherical top-hat window function. Although the classical Press and Schechter approach provides a good approximation under the assumption of a simple Gaussian process for \( f(\sigma) \), numerous recent studies (based on numerical simulations or solving theoretical issues) have devised more accurate parametrizations of the multiplicity function (e.g., Bond et al. 1991; Lacey & Cole 1994; Sheth & Tormen 1999; Jenkins et al. 2001; Crocce et al. 2010; Corasaniti & Achitouv 2011; Despali et al. 2015).

Relying on the Le SBARBBINE simulations, the main interest of the Despali mass function (DMF), compared with other widely used HMF like Tinker (Tinker et al. 2008), is the claim of universality, that is to say the parameters used to describe the multiplicity function \( f(\sigma) \) do not explicitly depend on either the redshift of the clusters or the cosmology. This is important because it means that when performing a cosmological analysis, the universal shape can be used at any redshift and in any cosmological scenario. This universality, however, requires use of \( M_{vir} \), the cluster mass defined with respect to the virial overdensity \( \Delta \) (e.g., Lahav et al. 1991).

The shape of the function is the following, originally from Sheth & Tormen (1999):

\[
\nu f(\nu) = A_0 \left[1 + \frac{1}{\nu'} \left(\frac{\nu'}{2\nu}\right)^{1/2} \right] e^{-\nu'/2},
\]

(3)

where \( \nu' = 4\nu \). The parameter \( \nu \) is related to the root mean square density fluctuation via

\[
\nu = \delta_c(z)/\sigma(M,z),
\]

(4)

where \( \delta_c(z) \) is the critical linear theory overdensity, which can be approximated by Kitayama & Suto (1996)

\[
\delta_c(z) \approx \frac{3}{20} (12\pi)^{2/3} [1 + 0.0123 \log(\Omega_m(z))],
\]

(5)

where log refers to logarithm in base 10.

We are then left with three parameters \( (a, p, A_0) \) describing the high-mass cutoff, the shape at lower masses, and the normalization of the curve, respectively. Once again, if the virial overdensity is used, no change in these parameters is expected with respect to either redshift or cosmology.

We compute the expected number density of objects per unit of mass, redshift, and area on the sky \( \Omega \) as

\[
\partial_c \ln M \partial \Omega N(M,z) = \frac{\bar{\rho}}{M} \frac{d\ln(\sigma(M,z))}{d\ln M} f(\sigma(M,z)) \partial_c \ln V
\]

(6)
where $\partial_{\alpha} V$ is the comoving volume per unit redshift and solid angle at the considered redshift. The quality of the cosmological parameter estimation with cluster number counts primarily depends on the number of clusters in the survey. Since we focus on the effect of the HMF parameterization, we do not consider any observable-mass relation in this work. We also assume that we can directly measure cluster mass and that the selection function can be modeled by a simple mass threshold $M_{\text{thres}}$. We present the survey cases studied (defined by $M_{\text{thres}}, \Omega, \text{redshift range } [z_{\min}, z_{\max}]$) in Sect. 2.3.

For each survey-type, we estimate the cosmological parameters

$$\Theta = \{\Omega_m, \sigma_8, w_0, \ldots\}$$

and simultaneously constrain the Despali HMF parameters,

$$\Xi = \{a, p, A_0\}.$$  

We only consider shot noise, so cosmic variance is neglected. Instead of using the Cash (1979) statistics in bins of mass and redshift, we compute the unbinned Poisson likelihood of the two dimensional point process:

$$\ln L(\Theta, \Xi) = -\int_{M_{\text{thres}}}^{\infty} \int_{z_{\min}}^{z_{\max}} \partial_{\ln M} N(M, z) \int \ln Mdz + \sum_{i=1}^{N_{\text{clusters}}} \ln (\partial_{\ln M} N(M_i, z_i))$$  

where the sum runs over individual clusters in the catalog and $N_{\text{clusters}}$ is the total number of clusters.

### 2.2. Fisher matrix

The aim of our Fisher analysis is to quantify the precision on the HMF parameters, $\Xi$, required to limit the impact of their uncertainty on the cosmological constraints. Using the Markov chain Monte Carlo (MCMC) provided by Despali et al. (2015), we fit a 2-D Gaussian on the HMF parameter distribution, for each pair of parameters. We show an example in Fig. 2 for the parameters $a$ and $A_0$. This enables us to reconstruct the full covariance matrix $C_{\text{HMF}}$ for the HMF parameters, which we provide in Table 1.

From the likelihood described in the previous section, one can compute the associated Fisher matrix for the clusters:

$$F_{\text{clusters},i,j} = -\left(\frac{\partial^2 \ln L}{\partial p_i \partial p_j}\right)$$  

where $p_i \in \Theta \cup \Xi$ and the angle brackets indicate the average over the data ensemble corresponding to the fiducial parameter values. The covariance matrix of the parameters $p_i$ is the inverse of the Fisher matrix.

We build the total Fisher information matrix as the sum of three components:

- The information obtained from cluster number counts, $F_{\text{clusters}}$, given in Eq. (8).
- The cosmological priors from Planck cosmic microwave background measurements, $F_{\text{ cmb}}$ (Planck Collaboration VI 2020).
- The HMF priors, $F_{\text{ HMF}}$.

The full Fisher matrix is then given by

$$F_{\text{tot}} = F_{\text{clusters}} + F_{\text{ cmb}} + F_{\text{ HMF}}.$$  

where $F_{\text{ cmb}}$ (resp. $F_{\text{ HMF}}$) is non zero only in the $\Theta$ (resp. $\Xi$) quadrant. The non zero quadrant of $F_{\text{ HMF}}$ is the inverse of $C_{\text{HMF}}$.

In this work, we examine the impact of $F_{\text{ HMF}}$ on cosmological parameter determination. In order to do so, we introduce the parameter $\alpha$, and we modify the priors on the HMF in Eq. (9) as follows:

$$F_{\text{ HMF}} \leftarrow \frac{1}{\alpha} F_{\text{ HMF}}.$$  

The covariance matrix being the inverse of the Fisher information, increasing $\alpha$ degrades the constraints on the HMF. On the other hand, $\alpha \rightarrow 0$ describes the situation where the HMF parameters are perfectly known. For $p \in \Xi$, this yields

$$\sigma_p^2 \leftarrow \alpha \sigma_p^2,$$

where $\sigma_p$ is the error on one parameter of the HMF. Figure 3 illustrates how two values of $\alpha$ impact the confidence ellipses for the HMF parameters.

Finally, we obtain

$$F_{\text{tot}}(\alpha) = F_{\text{clusters}}(\alpha = 1) + F_{\text{ cmb}} + \left(\frac{1}{\alpha} - 1\right) F_{\text{ HMF}}.$$  

Here, the last term comes from the fact that the cosmological parameters are degenerate with the HMF for cluster cosmology. To get $F_{\text{clusters}}$, we need to set a value for $\alpha$ (here $\alpha = 1$) and remove it afterwards. In the following sections, we change the value of $\alpha$ to measure its impact on $\sigma_p$ for any $p \in \Theta$, the set of the cosmological parameters, and to identify the value necessary to avoid degrading the cosmological parameter constraints by more than 5%, and the total FoM by more than 10%.

### Table 1. Covariance matrix, $C_{\text{HMF}}$, reconstructed for parameters $(a, p, A_0)$, all dimensionless.

| $a$     | $p$     | $A_0$   |
|---------|---------|---------|
| $5.56 \times 10^{-7}$ | $-1.01 \times 10^{-6}$ | $9.00 \times 10^{-8}$ |
| $-1.01 \times 10^{-6}$ | $3.19 \times 10^{-6}$ | $-1.93 \times 10^{-7}$ |
| $9.00 \times 10^{-8}$ | $-1.93 \times 10^{-7}$ | $4.37 \times 10^{-8}$ |
We do not consider the optical depth to reionization, \( \tau \), as a parameter, since it has no impact on cluster counts, the normalization of the primordial power spectrum being set with \( \sigma_8 \).

2.3. Cosmological models and mock catalogs

We generate catalogs for three different cosmological scenarios. Each model assumes a flat cosmology:

- A base \( \Lambda \)CDM cosmology (5 cosmological parameters\(^1\), \( \Omega_m, \sigma_8, H_0, \Omega_\nu, n_s \));
- A \( \omega_0 \)w\(_0\)CDM cosmology, where we assume a dynamical dark energy model (7 parameters, the parameters from the base flat \( \Lambda \)CDM and \( \omega_0 \)w\(_0\)CDM described in Section 2.2. The Fisher matrix is then deduced as the inverse of the covariance.
- A \( v \)CDM cosmology, in which we consider the physics of massive neutrinos. In this case, the parameters are the same as the ones presented in the base \( \Lambda \)CDM, but we now compute the power spectrum, as described in Sect. 3.3 (6 parameters, the parameters from the base flat \( \Lambda \)CDM and \( \sum m_\nu \)).

The expected cluster counts associated with these models are computed with the python package \textsc{HMF}{} from Murray et al. (2013), which we have modified for the two latter cosmological scenarios in a way described below. The values adopted for the cosmological parameters for each scenario can be found in Table 2. They are the mean values of the chains provided by Planck Collaboration VI (2020); PklimHM-TE-EE-lowE-lowlE-lowlE-lowlE-lowlE-bao-Riess18-Panthe for \( \omega_0 \)w\(_0\)CDM. For the case of the \( v \)CDM scenario, we adopt the lower limit provided by Abe et al. (2008), \( \sum m_\nu < 0.06 \text{eV} \), because the best fit value provided by \textsc{Planck} is lower.

The current lack of a precise understanding of the selection function of future optical surveys leads us to consider simple mass and redshift thresholds for the completeness of the catalog:

\[
\chi(M, z) = \begin{cases} 
1 & \text{if } M \in [M_{\text{thres}}, +\infty] \text{ and } z \in [0, 2] \\
0 & \text{otherwise}
\end{cases}
\]

The total expected number of objects in the catalog is then

\[
N_{\text{tot}} = \int_{M_{\text{thres}}}^{\infty} \int_{z=0}^{2} \frac{dz}{d\ln M} N(M, z) d\ln M dz. 
\]

We set the mass threshold so that \( N_{\text{tot}} \) is close to the number of clusters expected with a survey similar to \textit{Euclid}, that is to say \( \sim 60 \times 10^3 \) clusters over \( 15000 \text{deg}^2 \) (Sartoris et al. 2016). For the \( \Lambda \)CDM, this mass threshold is \( \log (M_{\text{thres}}/M_\odot) = 14.18 \). We then reduced and increased the area of the survey to obtain \( N_{\text{tot}} \) of \( \sim 500 \) and \( \sim 1.5 \times 10^4 \) clusters (\textit{Planck}-like and Future projects).

We generate five cluster catalogs. The first three (\( \text{C}_{\text{Euclid}}, \text{C}_{\text{Planck}}, \text{C}_{\text{Future}} \)) fix the sum of the neutrino masses to its minimal value of \( \sum m_\nu = 0.06 \) and are used for analyses in the \( \Lambda \)CDM and \( \omega_0 \)w\(_0\)CDM scenarios. Since constraining the masses of the neutrinos requires a special treatment for the power spectrum (see Sect. 3.3 for details), we generate two additional catalogs for the \( \nu \)CDM scenario corresponding to \textit{Euclid}-like and Future surveys. Since we are interested in forecasting the precision on the measurement of the neutrino mass scale, we do not generate a \textit{Planck}-like catalog for the \( \nu \)CDM scenario. The mocks are obtained from a Poisson draw of the DMF. The characteristics of each catalog are given in Table 3.

3. Results

We provide here the results for the three cosmological scenarios (\( \Lambda \)CDM, \( \omega_0 \)w\(_0\)CDM, \( v \)CDM) and the three catalogs (\textit{Planck}-like, \textit{Euclid}-like and Future) whose characteristics are given in the previous section. For each scenario and each catalog, we run the same analysis, and we give the precision required on the halo mass function.

Parameters which are poorly constrained with cluster counts such as \( \Omega_\nu, n_s, H_0 \) make the computation of the coefficients of the cluster Fisher matrix \( F_{\text{clusters}} \) difficult. Indeed, since the log-likelihood function is almost constant, we are more subject to numerical instabilities when we compute the second order derivatives. However, our analysis requires that we are able to capture sufficiently the correlations between the different parameters. We thus decided to run first an MCMC analysis and fit a Gaussian on the contours as we did for the HMF parameters in Sect. 2.2. The Fisher matrix is then deduced as the inverse of the covariance matrix. The MCMC analysis is performed with the emcee python package from Foreman-Mackey et al. (2013). Large flat priors are applied for the cosmological parameters of interest. They are summarized in Table 4. Convergence is checked with the integrated autocorrelation time, as suggested by Goodman & Weare (2010).

3.1. \( \Lambda \)CDM cosmology

We present the results for the clusters-only case, \( F_{\text{clusters}}(\alpha = 1) + \left( \frac{1}{\alpha} - 1 \right) F_{\text{HMF}}, \) in Sect. 3.1.1 and the clusters + \textit{Planck} CMB-case, \( F_{\text{clusters}}(\alpha = 1) + F_{\text{cmb}} + \left( \frac{1}{\alpha} - 1 \right) F_{\text{HMF}}, \) in Sect. 3.1.2.

3.1.1. Cosmological constrains from clusters only

Since galaxy cluster number counts are particularly sensitive to \( \Omega_m \) and \( \sigma_8 \), we naturally focus on these parameters for our analysis. In Fig. 4, we show the cosmological constraints for the
Euclid-like catalog with $\alpha = 1$, as well as our Gaussian approximation. As expected, $\Omega_m$ is poorly constrained as the clusters primarily probe the total matter content of the universe.

We now vary the parameter $\alpha$ encapsulating the uncertainty on the HMF parameters to see if we reach the required precision on the cosmological parameters. Recall that we require 5% precision on individual cosmological parameters (here, $\Omega_m$ and $\sigma_8$), and 10% precision on the total FoM (here, in the ($\Omega_m, \sigma_8$)-plane).

Figure 5 shows that the precision on the parameter $\Omega_m$ is not strongly affected by the variation of the parameters of the mass function. This is expected since Fig. 4 shows that this parameter is poorly correlated with the HMF parameters.

On the other hand, $\sigma_8$ is correlated with $a$, $p$ and $A_0$. Figure 6 shows that, for this parameter and a Future cluster catalog, we need better constraints (20% improvement) on the halo mass function parameters than those currently available.

The last issue concerns the FoM. Figure 7 demonstrates that, for clusters only, for a catalog of the size of Euclid and for a Future catalog, the current precision is sufficient to match our objective of 10%.

3.1.2. Cosmological constrains from clusters with Planck 2018 priors

Since clusters only poorly constrain $\Omega_m$, $H_0$ and $n_s$, we combine them with external constraints. The most natural complementary probe is the Planck primary CMB (Planck Collaboration VI 2020). Consequently, we include priors from Planck as $F_{\text{cmb}}$ (see Sect. 2.2) in our total Fisher matrix, and we run our analysis a second time. The priors are the same as in the previous section (see Table 4). For clarity, we do not add figures similar to those presented in the previous section, but instead summarize our results in Table 5. Corner plots for cosmological and HMF parameters are given in Appendix A.

Table 2. Fiducial cosmological parameters adopted to generate the mock catalogs in the three cosmological scenarios.

| $\Omega_m$ | $\sigma_8$ | $H_0$ (km s$^{-1}$ Mpc$^{-1}$) | $\Omega_b$ | $n_s$ | $w_0$ | $w_a$ |
|-----------|-----------|-------------------------------|-----------|-------|-------|------|
| $\Lambda$CDM (Planck2018) & $\nu$CDM |
| 0.3153 | 0.8111 | 67.36 | 0.04931 | 0.9649 | −1 | 0 |
| 0.2976 | 0.836 | 69.39 | 0.04645 | 0.9646 | −0.99 | −0.29 |

Notes. We adopt $T_{\text{CMB}} = 2.725$ K, $N_{\text{eff}} = 3.04$ and $\Sigma m_e = 0.06$ eV in the three cases. The parameters in black are the mean values of the chains provided by Planck Collaboration VI (2020). In red, the dark energy equation of state parameters, which are fixed for a $\Lambda$CDM cosmology.

Table 3. Characteristics of the mock catalogs.

| $\Omega(\text{deg}^2)$ | $N_{\text{clusters}}$ | $N_{\text{clusters}}^\Lambda$CDM | $N_{\text{clusters}}^\nu$CDM |
|------------------------|-----------------------|-------------------------------|-------------------------------|
| $C_{\text{Future}}$ | 34 503 | 153 891 | 191 408 |
| $C_{\text{Euclid}}$ | 15 000 | 66 252 | 82 694 |
| $C_{\text{Planck}}$ | 114 | 513 | 65 994 |

Notes. The first column is the survey area. We note that varying neutrino mass in the analysis changes the power spectrum. Thus, we generate a special set of two catalogs when letting the neutrino mass free. The two following columns are the number of clusters present in the catalog for a $\Lambda$CDM and a $w_0 w_a$CDM cosmology, and the last one is the same for $\nu$CDM (The cosmological parameters correspond to Table 2). For all the catalogs, $[z_{\text{min}}, z_{\text{max}}] = [0, 2]$ and $M_{\text{lim}}(\log(M/M_\odot)) = 14.18$.

Table 4. Priors applied on the flat $\Lambda$CDM cosmology.

| Parameter | Prior | Range |
|-----------|-------|-------|
| $\Omega_m$ | Flat | [0.1, 0.7] |
| $\sigma_8$ | Flat | [0.3, 0.99] |
| $H_0$ | Flat | [40, 250] |
| $\Omega_b$ | Flat | [0.03, 0.06] |
| $n_s$ | Flat | [0.5, 1.7] |
| $a$ | Gaussian from $\alpha \times C_{\text{HMF}}$ | |
| $p$ | Gaussian from $\alpha \times C_{\text{HMF}}$ | |
| $A_0$ | Gaussian from $\alpha \times C_{\text{HMF}}$ | |

For $\Omega_m$ the impact is more important because the constraints are tighter, but the current level of precision on the HMF is nevertheless good enough to avoid degrading the constraints beyond our threshold of 5%. However, this time $\sigma_8$ and the FoM are strongly affected, and improvements in HFM precision are needed to avoid degrading these constraints. All the results are provided in Table 5. We highlighted in red when the precision on the HMF needs to be improved. The precision on the HMF parameters needs to be improved for Euclid-like and Future catalogs (up to ~70% for the FoM for Future catalogs) if one wants to fully exploit their cosmological potential.

3.2. $w_0 w_a$CDM

The most common way of modeling the accelerated expansion of the universe is to consider a fluid with negative pressure, which is characterized by its equation-of-state $p = w \rho$. A cosmological constant corresponds to $w = -1$ and $\rho = c^2$. Even though the cosmological constant is consistent with current observations, many different dark energy models predict a dynamical equation-of-state. One of the most popular parametrizations was proposed by Chevallier & Polarski (2001) and Linder (2003),

$$w(a) = w_0 + w_a(1 - a),$$

which, as a two parameter expansion, adds a layer of complexity.

In this model, the HMF parameters have a different effect. First, leaving $w_0$ and $w_a$ free increases the size of the confidence contours. In addition, the shape of the 68 and 95% regions changes (see Figs. B.2 and B.3). Corner plots for cosmological and HMF parameters are provided in Appendix B.

For these reasons, the impact of the HMF parameters is smaller than in the $\Lambda$CDM case. For all three catalogs, we never reached a loss of relative precision comparable to the $\Lambda$CDM case, not only for $\Omega_m$ and $\sigma_8$, but also for the dark energy equation-of-state parameters $w_0$ and $w_a$. This should not obscure the fact that these results are highly dependant on the HMF uncertainties. Moreover, future experiments (CMB, BAO, etc)
Fig. 4. In light blue, MCMC 68% and 95% contours for the cosmological parameters and the halo mass function parameters for a Euclid-like catalog. We fixed $\alpha = 1$. The dark blue contours are obtained when adding Planck priors. The dashed lines are the Gaussian fit from which the covariance matrix and subsequently the Fisher matrix $F_{\text{clusters}}(\alpha = 1)$ is computed.
will constrain cosmology better than the Planck priors used here. These two aspects are discussed further in Sect. 5.

### 3.3. vCDM

For our final study we vary the sum of the neutrino masses. The presence of massive neutrinos induces a scale dependence of the growth factor, and slightly changes the model. Following Costanzi et al. (2013), we model the power spectrum as

$$P_{\text{cdm}}(k,z) = P_m(k,z) \left( \frac{\Omega_{\text{cdm}} T_{\text{cdm}}(k,z) + \Omega_\nu T_\nu(k,z)}{T_m(k,z) (\Omega_{\text{cdm}} + \Omega_\nu)} \right)^2,$$

and we also change the collapse threshold (Hagstotz et al. 2019)

$$\delta_c^* = \frac{\sigma_c(z)}{\sigma_{\text{cdm}}(z)} \delta_c.$$

With this formalism, we forecast cosmological constraints from the two mock catalogs generated to study varying neutrino mass, as described in Sect. 2.3. As we have already noted, the HMF uncertainty is not significant compared to the statistical uncertainties for a Planck-like catalog.

For the primary CMB prior $P_{\text{cmb}}$, we take the Planck chains PklinHM + TT + TE + EE + lowE + lensing + BAO with neutrino mass as a free parameter. Since the best fit value for the total neutrino mass provided by these chains is lower than the minimal value required by neutrino oscillation experiments (Abel et al. 2008), we center the Planck Fisher contours on the fiducial value of 0.06 eV, the minimal mass value.

Figure 8 shows the results for the largest catalog. We expect to achieve a 2σ measurement of the neutrino mass scale. Full corner plots for both cosmological and HMF parameters are provided in Appendix C. We see that clusters significantly improve the constraints on neutrino mass when added to the current Planck priors, in particular by constraining the lower mass range.

#### Table 5. Maximum value allowed for the $\alpha$ parameter in flat $\Lambda$CDM.

|                     | Clusters only | Clusters+Planck2018 |
|---------------------|---------------|---------------------|
| **Planck-like**     |               |                     |
| $\Omega_m$          | $\gg 1$       | $\gg 1$             |
| $\sigma_8$          | $\gg 1$       | $\gg 1$             |
| FoM                 | 1.6           | 0.67                |
| **Euclid-like**     |               |                     |
| $\Omega_m$          | $\gg 1$       | $\gg 1$             |
| $\sigma_8$          | 1.2           | 0.8                 |
| FoM                 | 1.6           | 0.67                |
| **Future**          |               |                     |
| (153 891 objects)   |               |                     |
| $\Omega_m$          | $\gg 1$       | 3.6                 |
| $\sigma_8$          | 0.8           | 0.32                |
| FoM                 | 1             | 0.31                |

**Notes.** Critical cases requiring improved precision on HMF parameters correspond to a value lower than 1 and are highlighted in bold.
Fig. 8. Cosmological parameters obtained for a catalog of 151 427 objects (Future). In orange, the Planck (2018) contours, and in green, the clusters combined with Planck priors. We expect, for the largest catalog, a 2σ measurement of the neutrino mass scale. The full corner plot with all parameters is given in Appendix C.

Fig. 9. Evolution of the uncertainty in the $\Omega_m - \Sigma m_\nu$ plane. In blue, the Euclid-like case, and in green a larger Future catalog.

And if neutrinos are slightly more massive than the minimal value, the confidence limits would improve: The green contour in the $\Sigma m_\nu - \sigma_8$ plane would move to higher values, but since the CMB would still constrain the upper value of the neutrino mass, we would obtain notably tighter constraints. In this sense, the plot is pessimistic. Moreover, the result shown here marginalizes over the Planck uncertainty on the optical depth to reionization, which can be expected to improve in the coming years and hence yield even tighter constraints on neutrino mass.

We perform the same analysis as the previous section, focusing only on the Clusters+Planck2018 case. Results are summarized in Table 6. We observe the same effect as for the $w_0w_z$ CDM scenario: since we allow the chains to explore a larger area of parameter space, it appears that the impact of our knowledge of the HMF is less important than in the case of $\Lambda$CDM. For a Euclid-like catalog, the current precision on the HMF parameter is sufficient, but not for the case of a Future catalog, which will require improvements by ~40% in order not to degrade the constraint on $\sigma_8$ by more than 5%, as illustrated in Fig. 9.

We must emphasize, however, that the best fit values of the HMF parameters ($\alpha$, $\nu$, $A_0$) (and associated uncertainties) are provided for a $\Lambda$CDM scenario. We have assumed these to be valid in the $\nu$CDM case, although we could reasonably expect the errors to be larger for a $\nu$CDM cosmology. We return to this point in the discussion.

4. Summary of results

Our study focuses on cluster counts as a cosmological probe. The wider and deeper surveys of the future will demand every increasing attention to systematics. As the primary theoretical link between cosmological quantities and the observed counts, the HMF is the mainstay of every past and future analysis. The most important aspects of the HMF in this context are:

- Universality, or the concept that the HMF functional parameters are not explicitly dependent on redshift nor cosmology.
- Accuracy, the ability to recover unbiased cosmological parameters.
- Precision in the fit of the HMF functional parameters when fit on numerical simulations.

State-of-the-art simulations are improving our understanding of the first two aspects (Despali et al. 2015). In this paper, we focus on the last point, estimating the required precision on a given set of parameters in order to avoid degrading cosmological constraints. We discuss the first two aspects in Sect. 5.

For our purposes, we set a threshold of 5% for allowed degradation in the precision for any single cosmological parameter, and 10% for the FoM. We find that, for the $\Lambda$CDM scenario, and for a Euclid-like catalog when Planck priors are applied, one of the principal parameter of interest, $\sigma_8$, as well as the FoM violate this threshold with the present precision: the precision on the HMF parameters must be improved by ~30% (Table 5). For a Future catalog, the situation is more critical: an improvement of ~70% is required to avoid degrading the FoM on $(\Omega_m, \sigma_8)$ by more than 10%.

Table 6. Maximum value allowed for $\alpha$ in a $\nu$CDM cosmology.

|                | Clusters+Planck2018 |
|----------------|----------------------|
| $\Omega_m$     | 7.7                  |
| $\sigma_8$     | 1.03                 |
| $\Sigma m_\nu$ | $\ll 1$              |
| $\text{FoM}(\Omega_m - \sigma_8)$ | 1.57            |
| $\text{FoM}(\sigma_8 - \Sigma m_\nu)$ | 1.45            |

Notes. Critical cases requiring improved precision on HMF parameters correspond to a value lower than 1, and are highlighted in bold.

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For $w_0, w_a$, CDM, any degradation in the dark energy equation-of-state parameters remains under a percent. This is simply because allowing these parameters to vary increases the size of the confidence ellipses, thereby reducing the impact of the HMF parameters. A look at the corner plots in Appendix B shows that there is in fact little correlation between $(w_0, w_a)$ and $(\alpha, p, A_0)$ in this case.

Lastly, we examined constraints on the sum of neutrino masses. Once again for a future catalog, the current precision on HMF parameters does not meet our threshold requirement, and one needs to improve the constraints on the HMF parameters by $\sim 30\%$ (Table 6). For a Euclid-like catalog, the current precision appears to be sufficient.

### 5. Discussion and conclusion

The HMF is the foundation of cosmology with cluster counts. It promises to provide a robust, simple, analytical form for the abundance of dark matter halos across cosmological models as a function of linear-theory quantities only. Universality is the key desired property: a functional form whose parameters do not explicitly depend on the cosmological scenario. Cosmological dependence only enters through the linear-theory power spectrum and growth factor. In this way, a universal HMF enables exploration of cosmological parameter space without the need for expensive numerical simulations at every step. In most analyses to date, the HMF parameters are fitted on a set of numerical simulations, and then fixed in the cosmological analysis, for example Planck Collaboration XXIV (2016).

Cluster catalogs will vastly increase in size with planned experiments, such as eROSITA (now operating), Simons Observatory, CMB-S4, Euclid, Roman and Rubin. As statistical power increases, we must quantify both the accuracy and the precision of the HMF fitting form, and evaluate their impact on cosmological constraints. The former tells us to what extent the best-fit HMF functional form reproduces the halo abundance in simulations; the latter refers to the precision of the HMF parameters around their best-fit values. This has not been an issue at the times when cluster catalogs were containing only a few hundreds objects. With Euclid, we expect $10^5$ clusters, so these errors can no longer be ignored (Costanzi et al. 2019).

Our study provides a framework to estimate the impact of the precision of HMF parameters on cosmological parameter estimation. We do this by freeing the parameters of the HMF, applying priors from the fit and quantifying the impact on the cosmological inference. We found that improvements of up to $70\%$ in the precision of HMF parameters is required to prevent degrading cosmological constraints by more than our threshold of $5\%$ on a single parameter and $10\%$ on the FoM.

Our results are reliable only for the mass function from Despali et al. (2015). The results could change for other HMFs depending on the correlations between HMF parameters and cosmological quantities. Thus, our main result is that a similar discussion should be made when presenting error bars extracted from simulations. It will not be possible in future analyses to simply fix the best-fit parameters of the HMF when running a cosmological analysis. One will have to adjust at the same time cosmology, and halo mass function parameters, with priors from simulation for these latter.

In this work, we have chosen not to include a fiducial scatter or bias on the mass estimates. Cluster masses are not directly measured, but inferred through proxies originating from various physical effects, and we thus expect at least a dispersion on the recovered masses. The probability distribution function of the proxies given the true masses (at a given redshift and a given cosmology) will be one of the keys required to measure cosmological parameters with cluster counts together with the selection function.

However, since the HMF is usually fitted directly on N-body simulations, we can decorrelate the study of the probability distribution function of the mass estimates, and the impact of the HMF, assuming that the masses are perfectly measured. Consequently, our results should be understood as the primary requirement that one should have: if one is considering an ideal HMF with no scatter on the masses, we do not want to degrade the cosmological parameters. The underlying problem is the impact of the systematics for a realistic study. In Artis et al. (in prep), we show that in the presence of a realistic scatter designed for Euclid, as well as a realistic selection function, the HMF errors still represent about $10\%$ of the total error budget. This is the same order of magnitude as the errors generated by the statistical uncertainty due to the number of clusters in the final catalog. In other words, when the full analysis is considered, the HMF remains a major concern.

No perfect functional form exists for the HMF, and in fact Fig. 1 shows that the error bar can not provide an overlap between two of the most widely used HMFs. This is the issue of accuracy. It is therefore essential to also verify that the adopted HMF provides unbiased cosmological constraints (see Salvati et al. 2020). Differences in HMFs can arise from numerous sources, including the effect of baryons (Castro et al. 2021), which will be a major issue in the upcoming years.

A related issue is whether different cosmological scenarios provide different uncertainties on HMF parameters (apart from the best-fit values). Indeed, the HMF formalism is based on the idea of universality, that is the claim that the HMF parameterization does not depend on the cosmology. Despali et al. (2015) demonstrated that this is nearly the case over a broad range of cosmological scenarios. But this only means that the best-fit is the same for the HMF parameters in different scenarios. Nothing is said about the uncertainties. There is no reason why the errors should not be affected when the cosmology changes, and this should be taken into account.

To solve this issue, one path could be to develop extensive simulated datasets for a broad range of cosmological scenarios. For each cosmological case in the parameter space (including massive neutrinos and the dark energy equation of state parameters), one would have to fit the HMF and recover the associated error bars. A final step would be to deduce the total size of the HMF parameters’ errors, as obtained when we marginalize on the cosmological parameters. This leads to an alternative idea: instead of trying to find a functional form for the HMF, one could build an emulator directly for the simulations. Recently, Bocquet et al. (2020) obtained interesting results with a precision competitive with classical fitting functions. In any case, the errors have to be included when one marginalizes on the cosmological parameters. The volume of the parameter space covered by the cosmological parameters should at least encapsulate the results of state of the art cosmological analyses like Planck Collaboration VI (2020) and Riess et al. (2019).

To summarize, we have to be sure that, in a given mass and redshift interval, the error bars provided encapsulate all the uncertainties related to simulation and halo finder (see Knebe et al. 2011 for a review).

In the near future, we will have discovered almost all dark matter halos hosting massive clusters of galaxies. But as statistics increases, effort must be made to not only provide a good description of what is seen in the simulations, but also to provide...
a framework capable of yielding unbiased cosmological constraints. We have focused on this latter point, in the special case of halo mass function parameters. Our conclusion is that it is important to construct a likelihood where the following are adjusted simultaneously:

- Cosmology (Ωm, σ8, w0, wa, ∑mν, ...),
- Selection function with priors from detection algorithm (see e.g., Euclid Collaboration 2019),
- Halo mass function with priors from simulation,
- Observable/mass relation with priors from simulation (see e.g., Bellagamba et al. 2019),
- Large scale bias of dark matter halos b(M, z) with priors from simulation (see e.g., Tinker et al. 2010).

This goal requires both theoretical and numerical improvements and must be reached by next-generation cluster surveys.

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References

Abazajian, K., Addison, G., Adshead, P., et al. 2019, CMB-S4 Science Case, Reference Design, and Project Plan

Abe, S., Ebihara, T., Enomoto, S., et al. 2008, Phys. Rev. Lett., 100, 221803

Ade, P., Aguirre, J., Ahmed, Z., et al. 2019, JCAP 2019, 056

Allen, S. W., Evrard, A. E., & Mantz, A. B. 2011, ARA&A, 49, 409

Basu, K., Remazeilles, M., Melin, J.-R., et al. 2021, Exp. Astron., in press, [arXiv:1909.01592]

Bellagamba, F., Sereno, M., Roncarelli, M., et al. 2019, MNRAS, 484, 1598

Bocquet, S., Dietrich, J. P., Schrabback, T., et al. 2019, ApJ, 878, 55

Bocquet, S., Heitmann, K., Habib, S., et al. 2020, ApJ, 901, 5

Bond, J. R., Cole, S., Efstathiou, G., & Kaiser, N. 1991, ApJ, 379, 440

Cash, W. 1979, ApJ, 228, 939

Castro, T., Borgani, S., Dolag, K., et al. 2021, MNRAS, 500, 2316

Chevallier, M., & Polarski, D. 2001, Int. J. Mod. Phys. D, 10, 213

Corasaniti, P. S., & Achitouv, I. 2011, Phys. Rev. D, 84, 023009

Costanzi, M., Villaescusa-Navarro, F., Viel, M., et al. 2013, JCAP, 2013, 012

Costanzi, M., Collaboration, D., Rozo, E., et al. 2019, MNRAS, 488, 4779

Crocce, M., Fosalba, P., Castander, F. J., & Gaztañaga, E. 2010, MNRAS, 403, 553

Despali, G., Giocoli, C., Angulo, R. E., et al. 2015, MNRAS, 456, 2486

Euclid Collaboration (Adam, R., et al.) 2019, A&A, 627, A23

Foreman-Mackey, D., Hogg, D. W., Lang, D., & Goodman, J. 2013, PASP, 125, 306

Fumagalli, A., Saro, A., Borgani, S., et al. 2021, A&A, submitted [arXiv:2102.08914]

Gehrels, N., & Spergel, D., WFIRST SDT Project 2015, in Journal of Physics Conference Series, J. Phys. Conf. Ser., 610, 012007

Goodman, J., & Weare, J. 2010. Commun. Appl. Math. Comput. Sci., 5, 65

Hagstøt, S., Costanzi, M., Baldi, M., & Weller, J. 2019, MNRAS, 486, 3927

Hannany, S., Alvarez, M., Artis, E., et al. 2019, Bull. Am. Astron. Soc., 51, 194

Hasselfield, M., Hilton, M., Marriage, T. A., et al. 2013, JCAP, 2013, 008

Hofmann, F., Sanders, J. S., Clerc, N., et al. 2017, A&A, 606, A118

Ivezic, Z., Kahn, S. M., Tyson, J. A., et al. 2019, ApJ, 873, 111

Jenkins, A., Frenk, C. S., White, S. D. M., et al. 2001, MNRAS, 321, 372

Kitayama, T., & Suto, Y. 1996, MNRAS, 280, 638

Knebe, A., Knollmann, S. R., Muldrew, S. L., et al. 2011, MNRAS, 415, 2293

Kravtsov, A. V., & Borgani, S. 2012, ARA&A, 50, 353

Lacey, C., & Cole, S. 1994, MNRAS, 271, 676

Lahav, O., Lilje, P. B., Primack, J. R., & Rees, M. J. 1991, MNRAS, 251, 128

Linder, E. V. 2003, Phys. Rev. Lett., 90, 091301

Mantz, A., Allen, S. W., Rapetti, D., & Ebeling, H. 2010, MNRAS, 406, 1759

Murray, S., Power, C., & Robotham, A. 2013, HMFcalc: An Online Tool for Calculating Dark Matter Halo Mass Functions

Planck Collaboration XXIV. 2016, A&A, 594, A24

Planck Collaboration VI. 2020, A&A, 641, A6

Press, W. H., & Schechter, P. 1974, ApJ, 187, 425

Riess, A. G., Casertano, S., Yuan, W., Macri, L. M., & Scolnic, D. 2019, ApJ, 876, 85

Rozo, E., Wechsler, R. H., Rykoff, E. S., et al. 2010, ApJ, 708, 645

Salvati, L., Douspis, M., & Aghanim, N. 2020, A&A, 643, A20

Sartoris, B., Biviano, A., Fedeli, C., et al. 2016, MNRAS, 459, 1764

Sheth, R. K., & Tormen, G. 1999, MNRAS, 308, 119

Tinker, J., Kravtsov, A. V., Klypin, A., et al. 2008, ApJ, 688, 709

Tinker, J. L., Robertson, B. E., Kravtsov, A. V., et al. 2010, ApJ, 724, 878

Vikhlinin, A., Kravtsov, A. V., Burenin, R. A., et al. 2009, ApJ, 692, 1060

Vot, G. M. 2005, Rev. Mod. Phys., 77, 207

Weinberg, D. H., Mortonson, M. J., Eisenstein, D. J., et al. 2013, Phys. Rep., 530, 870
Appendix A: Cosmological corner plots $\Lambda$CDM

![Cosmological corner plots $\Lambda$CDM](image)

Fig. A.1. Cosmological results (clusters only in light shading, Planck2018 with the black dotted line and clusters+Planck2018 in dark shading) for a Planck-like catalog of 513 objects in a $\Lambda$CDM cosmology.
Fig. A.2. Cosmological results (clusters only in light shading, Planck2018 with the black dotted line and clusters+Planck2018 in dark shading) for a Euclid-like catalog of 66,252 objects in a ΛCDM cosmology.
Fig. A.3. Cosmological results (clusters only in light shading, Planck2018 with the black dotted line and clusters+Planck2018 in dark shading) for a Future catalog of 153 891 objects in a ΛCDM cosmology.
Appendix B: Cosmological corner plots $w_0 w_a CDM$

Fig. B.1. Cosmological results (clusters only in light shading, Planck2018 with the black dotted line and clusters+Planck2018 in dark shading) for a Planck-like catalog of 513 objects in a $w_0 w_a CDM$ cosmology.
Fig. B.2. Cosmological results (clusters only in light shading, Planck2018 with the black dotted line and clusters+Planck2018 in dark shading) for a Euclid-like catalog of 66 252 objects in a $w_0w_a$CDM cosmology.
Fig. B.3. Cosmological results (clusters only in light shading, Planck2018 with the black dotted line and clusters+Planck2018 in dark shading) for a Future catalog of 153,891 objects in a $w_0 \omega_0$CDM cosmology.
Appendix C: Corner plots $\nu$CDM

Fig. C.1. Cosmological results (clusters only in light shading, Planck2018 with the black dotted line, clusters+Planck2018 in dark shading) for a Euclid-like catalog of 65,994 objects in a $\nu$CDM cosmology.
Fig. C.2. Cosmological results (clusters only in light shading, Planck2018 with the black dotted line and clusters+Planck2018 in dark shading) for a catalog of 151 427 objects in a rCDM cosmology.