



The *Swift*/BAT AGN Spectroscopic Survey. IX. The Clustering Environments of an Unbiased Sample of Local AGNs

M. C. Powell¹ , N. Cappelluti^{1,2} , C. M. Urry¹ , M. Koss³ , A. Finoguenov^{4,5} , C. Ricci^{6,7,8,9} , B. Trakhtenbrot^{10,17} , V. Allevato^{11,12} , M. Ajello¹³ , K. Oh^{14,15,18} , K. Schawinski¹⁴ , and N. Secrest¹⁶

¹ Yale Center for Astronomy and Astrophysics, and Physics Department, Yale University, P.O. Box 2018120, New Haven, CT 06520-8120, USA

² Physics Department, University of Miami, Coral Gables, FL 33155, USA

³ Eureka Scientific, 2452 Delmer Street, Suite 100, Oakland, CA 94602-3017, USA

⁴ Department of Physics, University of Helsinki, FI-00014 Helsinki, Finland

⁵ Max-Planck-Institut für extraterrestrische Physik, Giessenbachstrasse 1, D-85748 Garching, Germany

⁶ Núcleo de Astronomía de la Facultad de Ingeniería, Universidad Diego Portales, Av. Ejército Libertador 441, Santiago, Chile.

⁷ Instituto de Astrofísica, Facultad de Física, Pontificia Universidad Católica de Chile, Casilla 306, Santiago 22, Chile.

⁸ Chinese Academy of Sciences South America Center for Astronomy and China-Chile Joint Center for Astronomy, Camino El Observatorio 1515, Las Condes, Santiago, Chile

⁹ Kavli Institute for Astronomy and Astrophysics, Peking University, Beijing 100871, People's Republic of China

¹⁰ Department of Physics, ETH Zurich, Wolfgang-Pauli-Strasse 27, CH-8093 Zurich, Switzerland

¹¹ Scuola Normale Superiore, Piazza dei Cavalieri 7, I-56126 Pisa, Italy

¹² Helsinki Institute of Physics, P.O. Box 64, 00014 Helsinki, Finland

¹³ Department of Physics and Astronomy, Clemson University, Clemson, SC 29634-0978, USA

¹⁴ Institute for Particle Physics and Astrophysics, Department of Physics, ETH Zurich, Wolfgang-Pauli-Strasse 27, CH-8093 Zurich, Switzerland

¹⁵ Department of Astronomy, Kyoto University, Oiwake-cho, Sakyo-ku, Kyoto 606-8502, Japan

¹⁶ U.S. Naval Observatory, 3450 Massachusetts Avenue NW, Washington, DC 20392, USA

Received 2017 December 15; revised 2018 April 6; accepted 2018 April 9; published 2018 May 14

Abstract

We characterize the environments of local accreting supermassive black holes by measuring the clustering of AGNs in the *Swift*/BAT Spectroscopic Survey (BASS). With 548 AGN in the redshift range $0.01 < z < 0.1$ over the full sky from the DR1 catalog, BASS provides the largest, least biased sample of local AGNs to date due to its hard X-ray selection (14–195 keV) and rich multiwavelength/ancillary data. By measuring the projected cross-correlation function between the AGN and 2MASS galaxies, and interpreting it via halo occupation distribution and subhalo-based models, we constrain the occupation statistics of the full sample, as well as in bins of absorbing column density and black hole mass. We find that AGNs tend to reside in galaxy group environments, in agreement with previous studies of AGNs throughout a large range of luminosity and redshift, and that on average they occupy their dark matter halos similar to inactive galaxies of comparable stellar mass. We also find evidence that obscured AGNs tend to reside in denser environments than unobscured AGNs, even when samples were matched in luminosity, redshift, stellar mass, and Eddington ratio. We show that this can be explained either by significantly different halo occupation distributions or statistically different host halo assembly histories. Lastly, we see that massive black holes are slightly more likely to reside in central galaxies than black holes of smaller mass.

Key words: galaxies: active – galaxies: halos

1. Introduction

Studying the large-scale environments of active galactic nuclei (AGNs) is important for understanding the growth of supermassive black holes (SMBHs) and how they coevolve with their host galaxies (e.g., Kormendy & Ho 2013). Clustering is a powerful tool in statistically determining the typical dark matter halo in which AGNs reside, as well as how they occupy their halos. Coupled with a sensible model of halo mass assembly, this can constrain fueling mechanisms (i.e., mergers versus secular evolution) and feedback scenarios, providing selection effects are properly taken into account.

Previous studies of AGN clustering using soft X-ray and optically selected samples have found somewhat discrepant results for the typical host halo mass of AGNs. Luminous quasars drawn from wide-area optical surveys appear to lie in smaller halos ($M_h \sim 10^{12.5} M_\odot h^{-1}$, where $h = H_0/100 \text{ km s}^{-1} \text{ Mpc}^{-1}$) than moderate-luminosity X-ray AGN found in deeper

surveys ($M_h \sim 10^{13} M_\odot h^{-1}$) across a wide range of redshift (e.g., Croom et al. 2005; Gilli et al. 2005, 2009; Allevato et al. 2011, 2014; Krumpe et al. 2012; Ross et al. 2009; Shen et al. 2009). Additionally, it is not clear whether unobscured and obscured AGNs (either defined by their column density or optical classification) have the same clustering statistics, in accordance with the unified model, or if they tend to reside in different environments due to different accretion modes or due to being two stages of one evolutionary track, as claimed by recent (but discordant) studies (Allevato et al. 2014; Villarroel & Korn 2014; Mendez et al. 2016; DiPompeo et al. 2017a). However, they all probe different volumes, host galaxy properties, and luminosity ranges, making the comparison of various studies difficult (see, e.g., Mendez et al. 2016). Additionally, the picture may be confused because a large number of obscured AGNs have been missed in optical and soft X-ray surveys due to dust and gas obscuration from the torus and/or host galaxy. Population synthesis models of the Cosmic X-ray background indicate that a significant fraction of SMBH accretion occurs in obscured environments (Treister et al. 2004, 2012), meaning

¹⁷ Zwicky Fellow.

¹⁸ JSPS Fellow.

obscured AGNs are a vital population to consider in a full model of halo, galaxy, and SMBH (co-)evolution. Hard X-ray selection (>10 keV) can remedy this obscuration-related bias, as the majority of energetic photons are able to pass through large columns of gas and dust, up to Compton-thick levels ($N_{\text{H}} \approx 10^{24} \text{ cm}^{-2}$; Ricci et al. 2015). In addition, hard X-ray selection is extremely efficient, as there are very few contaminants, including the host galaxy.

The Burst Alert Telescope (BAT; Barthelmy et al. 2005; Krimm et al. 2013) instrument on the *Swift* satellite (Gehrels et al. 2004) has surveyed the entire sky to unprecedented sensitivity in the 14–195 keV band (Baumgartner et al. 2013; Oh et al. 2018). Local AGNs detected by BAT include the obscured and/or low-luminosity AGNs missed by optical detection, as well as the rare high-luminosity AGNs only found in wide-area surveys, so that BAT AGNs can solve some of the aforementioned issues with previous AGN clustering studies. Cappelluti et al. (2010) were the first to measure the clustering of *Swift*/BAT AGNs using a sample of 199 AGNs in the 36-month catalog (Ajello et al. 2009), but had uncertain results due to the small sample size. While they did find a dependence in X-ray luminosity, it was most likely a selection effect due to the strong redshift dependence inherent in any small flux-limited sample.

In this study, we have more than doubled the sample by using the 70-month *Swift*/BAT AGN catalog (Baumgartner et al. 2013), along with spectroscopic information from the *Swift*/BAT Spectroscopic Survey (BASS; Koss et al. 2017), to constrain the AGN halo occupation distribution (HOD) for 499 BASS AGNs in the redshift range $0.01 < z < 0.1$. To improve the statistics, we cross-correlate the AGNs with local 2MASS galaxies that trace the underlying dark matter distribution. Additionally, we investigate the environmental dependence of AGN parameters like obscuring column density and black hole mass, while matching distributions in X-ray luminosity, redshift, stellar mass, and Eddington ratio.

Krumpe et al. (2018) recently published a similar, independent clustering analysis of *Swift*/BAT AGNs, in which they analytically fit the cross-correlation function with 2MASS galaxies. They also divided their sample by optical classification (Type 1 or Type 2) from Baumgartner et al. (2013), as well as by observed X-ray luminosity. However, detailed X-ray spectral fitting (Ricci et al. 2017a) allows us to estimate the *intrinsic* absorption-corrected luminosity for each AGN, which differs strongly from the observed value at $N_{\text{H}} > 10^{23.5} \text{ cm}^{-2}$. The BASS DR1 release (Koss et al. 2017) also includes 46 new redshifts for a spectroscopic completeness of over 95%, and provides column densities for each 836 AGNs, which is the best method for understanding whether an AGN is obscured or not. Our study also differs from Krumpe et al. (2018) in how we fit models; namely, we populate dark matter halos statistically from *N*-body simulations (using the *Halotools* software package; Hearn et al. 2017). Because this allows a straightforward correction for catalog incompleteness, we use an extended redshift range ($z < 0.1$ rather than $z < 0.037$) for better number statistics (499 AGNs compared to 274 in Krumpe et al. 2018), and we do not have to rely on assumptions from analytic models. The simulation-based approach also allows us to look beyond halo mass to other halo parameters like halo concentration, in order to investigate effects such as assembly bias. In this paper we challenge the idea that AGN clustering is driven only by the typical mass of its dark matter halo.

This paper is organized as follows: we describe the data selection of the BASS AGN and 2MASS galaxies in Section 2; our method for measuring the correlation function and fitting it with a halo model is described in Section 3; Section 4 presents the results for the full AGN sample, as well as the dependence on obscuration and black hole mass; we discuss our findings in Section 5, and summarize them in Section 6. We assume flat Λ CDM cosmology ($\Omega_m = 0.3$, $\Omega_{\Lambda} = 0.7$, $H_0 = 100 h^{-1} \text{ km s}^{-1} \text{ Mpc}^{-1}$, $h = 0.7$), and errors quoted are 1σ unless otherwise stated.

2. Data

2.1. AGN Sample

BASS consists of 836 local AGNs from the *Swift*-BAT 70-month catalog (Koss et al. 2017; Ricci et al. 2017a), selected by their hard X-ray emission (14–195 keV), which has the benefit of being unbiased toward obscuration. BASS comprises the largest, most unbiased sample of local AGNs to date, and there is an abundance of complementary multi-wavelength ancillary data.¹⁹

Each AGN has soft X-ray data from *Chandra*, *XMM-Newton*, *Suzaku*, or *Swift*/XRT, so that the full X-ray spectra have been modeled (0.3–150 keV; Ricci et al. 2017a). These give the obscuring column (N_{H}) and intrinsic X-ray flux for each AGN, from which bolometric luminosities have been estimated using a fixed hard X-rays bolometric correction to the 14–195 keV luminosities ($L_{\text{bol}} = 8 L_{14-195\text{keV}}$; Koss et al. 2017).

Optical spectroscopy has been obtained for 641 unbeam AGN, providing spectroscopic redshifts that allow for 3D clustering analyses. We assume that the AGNs without spectra (5%) do not systematically affect the clustering of the overall population, as we verified negligible differences between the flux distributions and angular correlation functions with and without their inclusion. Black hole masses have been estimated for 429 AGNs, of which 54% are unobscured and 46% are obscured. Black hole masses from unobscured AGNs were estimated from $\text{H}\beta$ and/or $\text{H}\alpha$ broad lines FWHM (Kaspi et al. 2000; Greene & Ho 2005; Bentz et al. 2009; Trakhtenbrot & Netzer 2012; Mejia-Restrepo et al. 2016); these have uncertainties of 0.3–0.4 dex (e.g., Shen 2013; Peterson 2014). For obscured AGN without broad lines, we relied on the $M_{\text{BH}} - \sigma_*$ relation (Kormendy & Ho 2013), where σ_* was measured by fitting the spectra with host galaxy stellar templates. These black hole mass estimates have slightly larger uncertainties of ~ 0.5 dex (Xiao et al. 2011). Eddington ratios ($\lambda_{\text{Edd}} \equiv L_{\text{bol}}/L_{\text{Edd}}$) were derived from the bolometric luminosities and black hole masses via $L_{\text{bol}}/1.3 \times 10^{38} \text{ erg s}^{-1} [M_{\text{BH}}/M_{\odot}]$. The uncertainties on λ_{Edd} are driven by the large systematic uncertainties on both M_{BH} determinations (up to ~ 0.5 dex, see above) and bolometric corrections. The latter may be of roughly 0.2–0.3 dex, and perhaps involve more complicated uncertainties related to possible trends with luminosity and/or λ_{Edd} itself (Marconi et al. 2004; Vasudevan & Fabian 2007; Jin et al. 2012). More details of the optical spectral analysis can be found in Koss et al. (2017).

Stellar masses of the BAT AGN host galaxies were derived by spectrally de-convolving the AGN emission from stellar emission via SED decomposition. We combined near-IR data

¹⁹ <http://www.bass-survey.com>

from 2MASS, which is more sensitive to stellar emission, with mid-IR data from the AllWISE catalog (Wright et al. 2010), which is more sensitive to AGN emission. Where available, isophotal near-IR magnitudes from the 2MASS XSC were used to capture the most stellar emission, and the corresponding AllWISE elliptical magnitudes were used. We then converted the magnitudes to the AB system, and corrected for Galactic reddening using $E(B-V)$ estimates from Schlafly & Finkbeiner (2011). We used the low-resolution SED templates from Assef et al. (2010) to decompose the BAT AGN host galaxies into a linear combination of an AGN plus early-type (E), continuously star-forming (Sbc), and starburst galaxies (Im). To convert the luminosities of the galaxy components to masses, we obtained their stellar mass coefficients by fitting them with the Blanton & Roweis (2007) stellar population synthesis templates. The templates were convolved with the 2MASS/WISE system responses, and fit to the data via weighted non-negative least-squares, where for the weights we use the inverse variances of the data. Finally, we include AGN reddening by performing the SED decompositions along a logarithmically spaced grid of $E(B-V)$ values, choosing the value that yields the lowest χ^2 . To estimate random errors in the stellar mass uncertainties, we re-fit each source several times, each time removing one of the seven photometric data points (jackknife resampling), and permuting the remaining magnitudes by their uncertainties. This produced random errors of about 0.06 dex. There is also a component of scatter introduced by using the Assef et al. (2010) templates, which have three stellar components, instead of the five original Blanton & Roweis (2007) stellar population templates. By fitting the NASA-Sloan Atlas photometry with the Assef et al. (2010) templates, we estimate that there is an additional scatter term of about 0.08 dex, which we add in quadrature to the random error term provided above. Finally, the absolute stellar mass uncertainty for masses estimated using near-IR photometry is approximately a factor of two (Bell & de Jong 2001). We therefore estimate that our stellar mass uncertainties are about 0.32 dex, on average.

We selected AGNs in the redshift range $0.01 < z < 0.1$ with intrinsic (i.e., absorption-corrected) $L_{2-10\text{keV}} > 10^{42.5} \text{ erg s}^{-1}$, to remove any bias in peculiar velocities of low-redshift objects, as well as improve the AGN completeness for the luminosity range. The upper redshift limit was imposed to match the maximum redshift of the galaxy sample that we cross-correlated with the AGNs. After these cuts, the final number of AGNs in our sample is 499, and their distribution of redshift versus X-ray luminosity is shown in Figure 1. The luminosities are comparable to those of well-studied, higher redshift AGNs from pencil-beam X-ray surveys (e.g., COSMOS; Civano et al. 2016; Marchesi et al. 2016).

In addition to this full sample, we made subsamples in two bins of N_{H} (threshold = 10^{22} cm^{-2}) and two bins of black hole mass (threshold = $10^8 M_{\odot}$). Because the statistics are insufficient to use volume-limited samples, different luminosity subsamples automatically probe different volumes and host galaxy stellar masses. We therefore do not examine the clustering dependence on X-ray luminosity. Additionally, to avoid selection effects between the two bins of N_{H} and M_{BH} , we matched the subsamples in their distributions of redshift and X-ray luminosity. Specifically, we defined five bins of z ; then for each bin, we randomly selected N AGNs of the sample, with the larger number of sources in the bin to match the

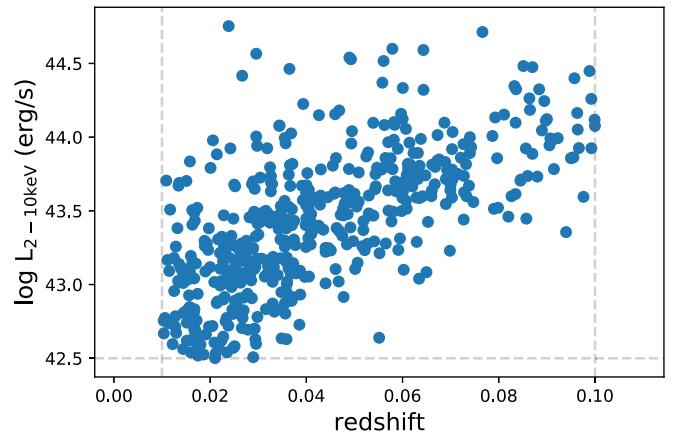


Figure 1. Log of the intrinsic 2–10 keV luminosity vs. redshift for the 499 BASS AGN at redshift $0.01 < z < 0.1$. The sample spans 2 decades in luminosity, but as in all flux-limited samples, there is a strong redshift-luminosity correlation.

Table 1
AGN Subsamples and Their Characteristics, Including the Number of AGNs, the Median Black Hole Mass, the Median 2–10 keV Luminosity (after Correcting for Absorption), and the Average Redshift of Each

AGN Sample	Threshold	N	\tilde{M}_{bh}	$\tilde{L}_{2-10\text{keV}}$	$\langle z \rangle$
L-limited (Full)	$L_X > 10^{42.5} \text{ erg s}^{-1}$	499	8.0	43.4	0.04
$\lambda_{\text{Edd}}\text{-limited}$	$\lambda_{\text{Edd}} > 0.01$	245	7.9	43.5	0.04
Obscured	$N_{\text{H}} \geq 10^{22} \text{ cm}^{-2}$	186	8.2	43.4	0.04
Unobscured	$N_{\text{H}} < 10^{22} \text{ cm}^{-2}$	186	7.7	43.4	0.04
Small M_{bh}	$M_{\text{bh}} \leq 10^8 M_{\odot}$	102	7.6	43.4	0.04
Large M_{bh}	$M_{\text{bh}} > 10^8 M_{\odot}$	102	8.4	43.4	0.04

Note. Black hole mass is in log units of M_{\odot} , and luminosity is in log units of erg s^{-1} .

number of sources (N) in the other sample. We then repeated the process for luminosity, with 10 bins of $\log L_{2-10\text{keV}}$. The total numbers were 186 AGNs in each bin of N_{H} and 102 AGNs in each bin of M_{BH} . Each random selection provided consistent results, as did using the derived bolometric luminosities rather than L_X . The characteristics of these subsamples are summarized in Table 1.

2.2. Galaxy Sample

Using a dense sample of galaxies as tracers of the underlying dark matter distribution greatly boosts AGN clustering statistics (e.g., Coil et al. 2009). We therefore cross-correlated our AGN sample with galaxies from the 2MASS Redshift Survey (Huchra et al. 2012), as the redshift range is close to that of the AGN sample ($z_{\text{peak}} \sim 0.03$). Selected based on their K -band magnitude, $K_s \leq 11.75$, the galaxies are spectroscopically complete and cover 91% of the sky (the Galactic plane is excluded; $|b| > 8^{\circ}$).

We estimated stellar masses of the 2MASS galaxies by employing a universal mass-to-light ratio (M/L) between K -band luminosity and stellar mass, as K -band M/L does not significantly vary with mass at $z = 0$ (Lacey et al. 2008), nor is it sensitive to dust content. We use an absolute solar K_s band magnitude of 3.29 (Blanton & Roweis 2007) to obtain the luminosities and fit our measured autocorrelation function for M_*/L_{K_s} , as described in Section 3.3. The random error associated with using a single M/L is about 0.3 dex (Bell &

de Jong 2001). However, we only use the resulting distribution of stellar mass in our model, and we verified that convolving the distribution with this error does not change our results.

We used the full flux-limited sample for maximal statistics, and corrected for incompleteness as a function of stellar mass when modeling the galaxies via the process described in Section 3. We excluded 2MASS galaxies that are also in the BASS AGN catalog (to within $3''$; 361 sources) so that the cross-correlation measurement was between two independent catalogs. In total, we used 38,567 galaxies in the redshift range $0.01 < z < 0.1$.

3. Method

3.1. Correlation Function Measurement

The quantitative measure of clustering is the two-point correlation function, which quantifies the excess probability over a random distribution that a pair of objects are separated by a given distance (r). We used the Landy–Szalay estimator (Landy & Szalay 1993):

$$\xi(r) = \frac{D_1 D_2(r) - D_1 R_2(r) - R_1 D_2(r) + R_1 R_2(r)}{R_1 R_2(r)}, \quad (1)$$

where DD, DR, and RR correspond to the data–data, data–random, and random–random pairs, respectively. For an autocorrelation (ACF) measurement, the subscripts correspond to the same data set, while they represent two different data sets for a cross-correlation. The random catalogs for each data set have the same selection function as the data survey. Rather than using the AGN ACF, which has large uncertainties because of the rather small AGN sample, we cross-correlated the AGNs with the larger galaxy sample to improve statistics.

We created a random AGN sample with the same selection as the BASS survey by using the *Swift*/BAT sensitivity map. We first randomized the position of each random AGN on the sky, and then assigned it a flux drawn from the flux distribution of the data. If the flux was greater than the sensitivity at that position, we kept that specific randomly generated AGNs; otherwise we omitted it. We then assigned it a redshift drawn from the redshift distribution of the data, smoothed with a Gaussian kernel with $\sigma_z = 0.2$. We repeated this process for each AGN subsample (e.g., each bin in black hole mass or in absorbing column density). Due to the low number density of the data, we made each random AGN sample ~ 100 times larger than the corresponding BASS sample.

For the galaxy random catalog, we assumed that the sensitivity is uniform across the sky and randomized the angular positions, excluding the galactic plane ($|b| < 8^\circ$). We assigned each random galaxy a redshift drawn from the distribution in the real data, also smoothed with a Gaussian kernel with $\sigma_z = 0.2$. The redshift distributions of the galaxies, AGNs, and random samples are shown in Figure 2. The number of random galaxies is 20 times more than the number of 2MASS Redshift Survey galaxies.

We measured ξ in bins of r_p and π (distances perpendicular and parallel to the line of sight, respectively) using the pair counter from the publicly available software CorrFunc (Sinha & Garrison 2017), which counts the number of pairs of galaxies in a catalog separated by r_p and π . We then projected through redshift space to eliminate any redshift-space

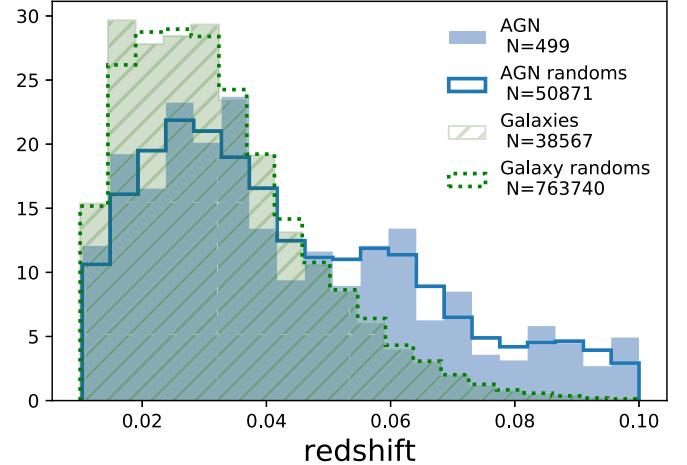


Figure 2. Normalized redshift distributions of the AGN and galaxy samples, along with their respective random catalogs. The redshift distribution of each population is well matched by its randomly positioned counterparts.

distortions, to get the projected correlation function:

$$w_p = 2 \int_0^{\pi_{\max}} \xi(r_p, \pi) d\pi. \quad (2)$$

The value of π_{\max} was chosen such that the amplitude of the projected correlation function converges and gets noisier for any higher values. We found this to be $60 \text{ Mpc } h^{-1}$ for our sample, which is a commonly used value for π_{\max} .

We calculated the covariance matrix via the jackknife resampling method,

$$C_{i,j} = \frac{M}{M-1} \sum_k [w_k(r_{p,i}) - \langle w(r_{p,i}) \rangle] \times [w_k(r_{p,j}) - \langle w(r_{p,j}) \rangle], \quad (3)$$

where we split the sample into $M = 25$ sections of the sky, and computed the cross-correlation function when excluding each section (w_k). We chose $M = 25$ so that the patches were large enough to probe the largest r_p scale at our minimum redshift, yet numerous enough to create a normal distribution. We quote the errors on our measurement as the square root of the diagonals: $\sigma_i = \sqrt{C_{i,i}}$.

3.2. Model Formulation

In the hierarchical model of structure formation, galaxies reside in dark matter halos, which have gravitationally collapsed at the peaks of the underlying dark matter distribution. In this context, clustering statistics of galaxies depend only on the cosmology (how dark matter halos cluster; the two-halo term, dominant on scales $\gtrsim 1 \text{ Mpc } h^{-1}$) and how the galaxies occupy their dark matter halos (one-halo term; $\lesssim 1 \text{ Mpc } h^{-1}$), which depends on their formation and evolution. We consider two kinds of models to describe the latter: a HOD model and a subhalo model, described in the following sections.

In both cases, we used *Halotools* (Hearin et al. 2017) to compute the model cross-correlation functions. This software populates dark matter halos from an N -body simulation with a model and computes the two-point statistics for the resulting galaxy mock catalog. Because we cross-correlated AGNs with galaxies, we first created a mock sample with the same

clustering statistics as the 2MASS galaxies (described in Section 3.3). We then used this simulated galaxy sample to cross-correlate with the AGN mock derived from the model. The average and median halo masses of the AGN sample were calculated empirically from the AGN mock.

We did the analysis with two redshift $z = 0$ halo catalogs (based on the ROCKSTAR halo-finder; Behroozi et al. 2013) from different simulations, both included in `Halotoools`: first, the Bolshoi–Planck simulation (Riebe et al. 2013), which has a resolution of $1.35 \times 10^8 M_{\odot} h^{-1}$ and a box size $L = 250 \text{ Mpc } h^{-1}$ using Planck 2013 cosmological parameters (Planck Collaboration et al. 2016); second, the Consuelo simulation, which has a larger volume ($L = 420 \text{ Mpc } h^{-1}$) but poorer resolution ($2 \times 10^9 M_{\odot} h^{-1}$), with WMAP5 cosmology (Komatsu et al. 2009). The results are consistent with each other; however, since the Bolshoi–Planck simulation is complete down to haloes of mass $M_{\text{vir}} \sim 10^{11} M_{\odot}$, we quote results from that analysis, which is better able to constrain minimum halo mass.

3.2.1. HOD Model

The HOD formalism (e.g., Cooray & Sheth 2002) describes the probability that N galaxies (or AGN) reside in a host halo of mass M_h . To first order, this can be described as the average number of galaxies per host halo as a function of halo mass, $\langle N \rangle(M_h)$.

The HOD can be disaggregated into centrals and satellite galaxies, where the total HOD is sum of the two components $\langle N \rangle = \langle N_c \rangle + \langle N_s \rangle$. We used a simple parametrization for the AGN HOD, derived from the Zheng et al. (2007) model:

$$\langle N_c \rangle(M_h) \propto \Theta(M_h - M_{\text{min}}), \quad (4)$$

$$\langle N_s \rangle(M_h) \propto \left(\frac{M_h - M_{\text{min}}}{M_1} \right)^{\alpha}, \quad (5)$$

where Θ is the Heaviside step function, M_{min} is the minimum halo mass to host a central AGN, M_1 is typical halo mass that starts hosting satellites, and α is the power-law slope of the satellites. We assumed $\log(M_1/M_{\text{min}}) = 1.2$, which is the case for galaxies with $M_r < -20 \text{ mag}$ (Zehavi et al. 2011), and we left M_{min} and α as two free model parameters. The normalization of the HOD is not constrained by the correlation function. We searched for the best-fit model by stepping through $\log M_{\text{min}} - \alpha$ parameter space in 0.1 unit increments ($11.2 < \log M_{\text{min}} < 12.8$; $-0.5 < \alpha < 1.5$), where at each step we averaged five model realizations, and found where the correlated χ^2 was minimum:

$$\chi^2 = \sum_{i,j} [w_{\text{obs}}(r_{\text{p},i}) - w_{\text{mod}}(r_{\text{p},i})] \times C_{\text{eff},i,j}^{-1} \times [w_{\text{obs}}(r_{\text{p},j}) - w_{\text{mod}}(r_{\text{p},j})], \quad (6)$$

where w_{obs} and w_{mod} correspond to the correlation function of the real data and mock data. Because the model has sample variance uncertainty from the finite simulation volume, C_{eff} is the sum of the covariance matrices from the data and simulation (Zheng & Guo 2016). The simulation covariance matrix was also estimated via Jackknife resampling, by splitting the simulation box into 125 cubes. We report the best-fit parameters in Section 4.

For each realization of the HOD model, `Halotoools` populates the host halos with the mock central galaxies and adds satellites according to an NFW profile (Navarro et al. 1996). This is done only for the largest virialized halos in the catalog (i.e., ignoring the subhalo information).

3.2.2. Subhalo Model

The second type of model assumes a one-to-one relation between the galaxies and all halos and subhalos. We used the Behroozi et al. (2010) model based on abundance matching, which assumes that stellar mass predominantly determines the clustering of the sample via the stellar-to-(sub)halo mass relation. This model has been calibrated and tested with galaxy observations, and so it provides an additional check to see if AGNs occupy halos in the same way as inactive galaxies (i.e., based primarily on their stellar mass).

For this model, the `Halotoools` software populates a mock galaxy at the center of each halo *and* subhalo, and assigns it a stellar mass based on the peak mass of that (sub)halo. The mock galaxies in the center of each host halo correspond to the centrals galaxies, while the mocks in the subhalos correspond to the satellites.

This method allows us to correct for the incompleteness of the AGN catalog in the following way: we populated the halos from our halo catalog with galaxies according to the Behroozi model, and then divided the stellar mass distribution of the resulting galaxy mock sample with the stellar mass distribution of the BASS AGN. We normalized it to obtain the incompleteness fraction as a function of stellar mass. We then assigned random values between 0 and 1 to the mock galaxies, and masked out any mock whose value fell below the incompleteness fraction for its assigned stellar mass. Consequently, the resulting mock sample had the same stellar mass distribution as the data.

This subhalo-based model is approximately equivalent to the HOD model that assumes $\alpha = 1$, as the number of subhalos (and hence satellite galaxies above a threshold luminosity) scales with halo mass. However, it is not biased by incompleteness in flux-limited catalogs. For this model there are no parameters to fit; rather, we simply assess how well the model agrees with the data.

3.3. Galaxy Mock Creation

We used the subhalo model to create a mock galaxy sample with the same stellar mass distribution as the full flux-limited 2MASS galaxy catalog. We fit for the K_s -band M/L , by comparing the resulting mock autocorrelation function with the data. The best-fit value was found to be 0.6 in solar units ($\chi^2_{\nu} = 0.754$). The masked autocorrelation function using the Bolshoi–Planck halo catalog with the best-fit M/L ratio is shown in Figure 3, along with the autocorrelation function of the 2MASS galaxies. We found that using M/L ratios of upper and lower bounds of the 99% confidence limits of the fit does not significantly change our results. Both simulations produced consistent results.

4. Results

The results for both models are summarized in Table 2 for each AGN sample.

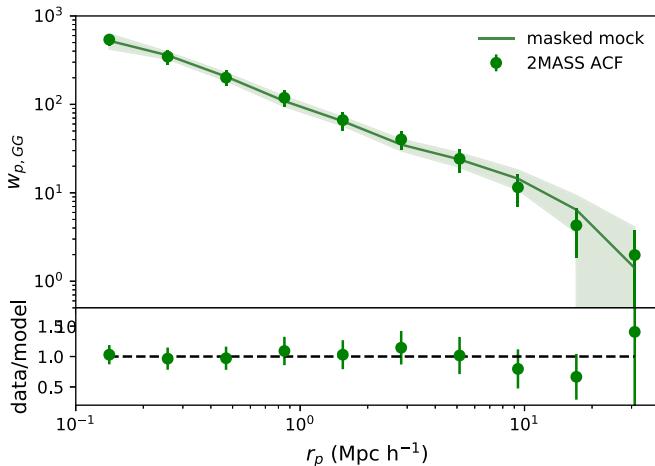


Figure 3. Projected autocorrelation function of 2MASS galaxies compared with the mock sample created using the Behroozi et al. (2010) model and 2MASS selection function.

4.1. Full AGN Sample

The left panel of Figure 4 shows the projected cross-correlation function for the full AGN sample and the corresponding HOD model fit: $\log M_{\min}/M_{\odot} h^{-1} = 12.4^{+0.2}_{-0.3}$, $\alpha_{\text{AGN}} = 0.8^{+0.2}_{-0.5}$. We find that the average dark matter halo mass in which AGNs reside is $\log M_h/M_{\odot} h^{-1} = 13.4 \pm 0.2$, and the median mass is $\log M_h/M_{\odot} h^{-1} = 12.8 \pm 0.2$, from empirical measurement of the mocks from the 1σ best-fit HOD region. This is consistent with the measurement in Cappelluti et al. (2010), as well as in Krumpe et al. (2018). The smoothed contour map of the fit to the two-parameter HOD is shown in the right panel. While the associated significances of the contour levels should be taken with caution, we verified that the projected probability distributions for each parameter are nearly Gaussian, which we use to quote the errors on the best-fit parameters. The satellite power-law slope is consistent with that of the local inactive galaxy population ($\alpha \sim 1$), but favors $\alpha < 1$. We stress that this HOD fit is using the full flux-limited sample, which is incomplete for all AGN luminosities. Thus the derived HOD pertains to AGN with median bolometric luminosity of $10^{44.7} \text{ erg s}^{-1}$ at $z \sim 0.04$. However, we are able to compare our full cross-correlation measurement with an Eddington ratio-limited sample ($\lambda_{\text{Edd}} > 0.01$), which has been suggested to have a more unbiased HOD (Jones et al. 2017) than a luminosity-limited sample. Although the statistics are poorer due to only a fraction of the sources having black hole mass estimates, we find it also agrees with our best-fit HOD model (Figure 4).

Figure 4 also shows the results from our subhalo model analysis, which agrees well with the data ($\chi^2_{\nu} = 1.6$), despite there being no free parameters. The advantage of the subhalo model is that it takes into account the incompleteness of the catalog. The median halo mass, $\log M_h/M_{\odot} h^{-1} \sim 12.3$, is lower than was found with the HOD model (12.8) because of the proper treatment of incompleteness (i.e., taking into account the smaller mass galaxies and halos that were missed). We therefore conclude that AGNs, on average, do not live in special environments compared with the overall galaxy population, as our only assumption was that the host galaxy stellar mass distribution of the AGN sample drives its clustering via the stellar-to-subhalo mass relation.

4.2. Clustering versus Obscuration

Figure 5 shows the cross-correlation function of unabsorbed ($N_{\text{H}} < 10^{22} \text{ cm}^{-2}$) versus absorbed ($N_{\text{H}} \geq 10^{22} \text{ cm}^{-2}$) AGNs with their corresponding HOD fits. The luminosity and redshift distributions of the two bins are shown.

While the two-halo terms of the data seem consistent with each other, the obscured AGNs appear more clustered on small scales (by $\sim 3\sigma$), consistent with recent studies of narrow-versus broad-line AGNs in the Sloan Digital Sky Survey (SDSS; Jiang et al. 2016) and in *Swift*/BAT AGNs (Krumpe et al. 2018). This was also seen using the full sample (without matching luminosity distributions) and using different N_{H} thresholds up to (but not including) 10^{23} cm^{-2} .

The stellar mass distributions shown in Figure 5 are very similar, so this cannot cause the difference in clustering. The subhalo model for unobscured AGN (dotted blue line) is inconsistent with the data (blue dots) ($\chi^2_{\nu} > 4$), another indication that factors beyond host galaxy stellar mass are determining the clustering signal.

The $\Delta\chi^2$ contour plots of the separate HOD fits for obscured and unobscured AGNs are also shown in Figure 5; the shapes of the HODs differ by more than 4σ . M_{\min} and α are different for each: unobscured AGNs have smaller minimum halo mass and a shallower satellite-term slope. This would suggest that unobscured AGNs tend to be in central galaxies while obscured AGNs are more likely to be in satellites. The corresponding average dark matter halo masses are $\log M_h/M_{\odot} h^{-1} = 13.5 \pm 0.2$ for obscured AGNs and $\log M_h/M_{\odot} h^{-1} = 12.6 \pm 0.3$ for unobscured AGNs. The finding that obscured AGNs live in larger mass halos than their unobscured counterparts agrees with recent results of angular clustering studies of infrared-selected *WISE* AGN (Hickox et al. 2009; DiPompeo et al. 2014, 2017a). It is inconsistent, however, with clustering studies of Type 1 versus Type 2 X-ray-selected AGN in COSMOS (Allevato et al. 2014), although these studies probed AGN at higher redshift ($z \sim 1$) and different luminosity ranges. This inconsistency could also be due to the host galaxies; the Type 1 sample had systematically higher luminosities, indicating they most likely had larger host galaxy stellar masses, which may explain why a larger bias for Type 1 AGN was found. The clustering properties of unobscured AGNs are also consistent with the halo masses found for optical quasar samples across a wide range of redshift (e.g., Croom et al. 2005; Ross et al. 2009; Shen et al. 2009).

The distinctly different halo masses of obscured and unobscured AGNs could be due to intrinsic differences between the two types. It has been suggested that (Compton-thin) obscured AGNs tend to have lower Eddington ratios than unobscured AGNs, since the covering factor depends on mass-normalized accretion rate (e.g., Ricci et al. 2017c). This would cause our sample of obscured AGNs to have systematically larger black holes than the unobscured AGNs since we matched their luminosities, which we verified with their M_{BH} distributions. To test if this is biasing the result, we considered the objects that have black hole mass and accretion rate estimates ($\sim 75\%$ of the AGN sample analyzed), and measured the clustering of Compton-thin ($N_{\text{H}} < 10^{23.5} \text{ cm}^{-2}$) obscured and unobscured AGNs after matching distributions of Eddington ratio rather than luminosity (Figure 6). The differences between the two types are still present, suggesting something else determines the environmental differences. However, it should be noted that the black hole mass determination for obscured AGNs is less precise than for unobscured AGNs.

Table 2
Halo Model Parameters for Each AGN Subsample, for Both the HOD and Subhalo Models

AGN Sample	HOD Model					Subhalo Model		
	\tilde{M}_h	$\langle M_h \rangle$	M_{\min}	α	χ^2_ν	\tilde{M}_h	$\langle M_h \rangle$	χ^2_ν
Full	$12.8^{+0.2}_{-0.1}$	$13.4^{+0.1}_{-0.3}$	$12.4^{+0.2}_{-0.3}$	$0.8^{+0.2}_{-0.5}$	1.5	12.3	13.3	1.6
Obscured	$12.9^{+0.3}_{-0.7}$	$13.5^{+0.2}_{-0.2}$	$12.5^{+0.2}_{-0.8}$	$1.1^{+0.4}_{-0.2}$	1.9	12.3	13.3	2.1
Unobscured	$12.0^{+0.2}_{-0.3}$	$12.6^{+0.2}_{-0.3}$	11.4 ± 0.2	$0.4^{+0.2}_{-0.4}$	1.0	12.3	13.3	4.5
Small M_{bh}	$12.6^{+0.2}_{-0.9}$	$13.4^{+0.2}_{-0.9}$	$12.1^{+0.4}_{-1.0}$	$0.9^{+0.2}_{-0.4}$	2.6	12.2	13.3	2.1
Large M_{bh}	$12.8^{+0.2}_{-0.4}$	$13.2^{+0.2}_{-0.3}$	$12.4^{+0.2}_{-0.4}$	$0.2^{+0.5}_{-0.4}$	0.6	12.4	13.3	1.6

Note. All masses are in log units of $M_\odot h^{-1}$.

4.2.1. Role of Assembly Bias

Another possible difference of clustering between obscured and unobscured AGNs could be related to the host halo assembly history rather than the total halo mass, an effect known as assembly bias (e.g., Gao et al. 2005; Dalal et al. 2008). In general, there could be a connection between the mass assembly onto the host halo and the mass assembly onto the central black hole, and additionally, a link between obscuration and halo age can constrain whether obscuration is an evolutionary AGN phase. Also, if mergers are significant for AGN obscuration, it is possible that the merging of the subhalos, relating to the assembly of the host halo, would leave an imprint on the clustering signal. It has been shown that the amount of substructure in halos of a given mass depends on formation epoch (e.g., Gao et al. 2005; Dalal et al. 2008), as subhalos in early-forming hosts have more time to fall toward the center and thus are more concentrated. Therefore, if unobscured AGNs reside preferentially in halos formed late, then both the one-halo and two-halo terms of their correlation function would be reduced. This is quite a different explanation than suggested by the HOD analysis (that they are preferentially central galaxies).

We investigated this scenario with a simple toy model: we populated the halo catalog with our subhalo model and then split the sample by the NFW concentration of their host halos ($c \equiv r_{\text{vir}}/r_s$, where r_{vir} is the virial radius and r_s is the NFW scale radius), which correlates with halo formation time (e.g., Wechsler et al. 2002). We assumed there is a maximum threshold concentration to host an unobscured AGN, and a minimum threshold concentration to host an obscured AGN. While reality is likely to be more complicated, this simple model can explain the overall trend. We found that the obscured sample is best fit by $c > 10 \pm 2$ ($\chi^2_\nu = 1.8$), and the unobscured sample is best fit by $c < 13 \pm 2$ ($\chi^2_\nu = 0.9$; the median concentration of the mock sample is $c \sim 10$). Figure 7 shows the projected cross-correlation function of both models, compared to the obscured and unobscured AGN samples, and Table 3 summarizes the best-fit parameters. There is good agreement with the data, even with such a simple model. The average halo concentrations of each mock sample are $c \sim 8.5$ (unobscured) and $c \sim 27$ (obscured), which correspond to the formation epochs of $z \sim 1$ and $z \sim 5.5$, respectively (see Wechsler et al. 2002, who define the formation epoch as the time when the halo mass accretion rate, $d \log M_h/d \log a$, falls below 2).

From this exercise, we see that obscured AGNs do not necessarily reside in more massive halos than unobscured AGNs (concentration is inversely proportional to mass); rather, it is possible that unobscured AGNs instead prefer halos with low concentration and/or later formation epochs. Evidence of this was seen in a comparison between narrow- and broad-line AGNs in SDSS; Type 2 AGNs seem to reside in groups that are more centrally concentrated (Jiang et al. 2016). It remains unclear whether the distribution of satellites or the halo formation epoch would be driving this preference. Note that these results are the opposite of what we would expect for the evolutionary picture in which a merger-triggered AGNs is first obscured and then evolves into an unobscured phase; in that case, the obscured AGNs would reside in the most recently formed halos, with much smaller difference in average host halo formation epoch between each sample.

4.3. Clustering versus Black Hole Mass

Figure 8 shows the results of the correlation function measurements and HOD fitting for the AGN sample divided into two bins of black hole mass. We again randomly down-sampled each bin in order to avoid selection effects. The differences between the two samples are not significant; there is a $\sim 1\sigma$ difference in α , in the sense that the satellite slope is shallower for large black hole masses than for the smaller ones, with best-fit values 0.2 ± 0.5 and 0.9 ± 0.3 , respectively. A satellite power-law slope of 0 is consistent with the population residing purely in central galaxies; this is within the uncertainties, given such large black hole mass bin sizes and the limited sample size. (While the χ^2_ν is large for the small black hole bin (2.6), it should be noted that it becomes 1.3 with the same best-fit parameters after removing one data point.)

While the correlation between black hole mass and halo mass (e.g., Silk & Rees 1998; El-Zant et al. 2003; Booth & Schaye 2010) would predict that the larger black hole bin would have a larger bias, as was found in Krumpe et al. (2015), we find no significant difference. The median halo masses for each bin are $\log M_h/M_\odot h^{-1} = 12.6 \pm 0.3$ and 12.8 ± 0.3 , for small and large black holes, respectively.

Our results may suggest that larger black holes are less likely to reside in satellite galaxies, which would make sense, assuming a correlation between the mass of the black hole and the mass of its host subhalo. More data are needed to conclusively confirm this.

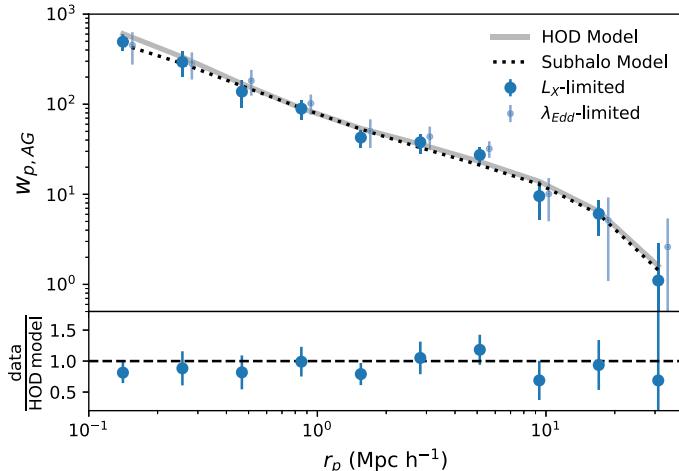


Figure 4. Left: projected cross-correlation function of 2MASS galaxies and BASS AGNs (blue points), with the best-fit HOD model (gray solid line) and subhalo model (black dotted line) for the 2–10 keV luminosity-limited sample. The lower panel shows the data divided by the HOD model. An AGN sample limited by Eddington ratio ($L/L_{\text{Edd}} > 0.01$; light blue points) is consistent with the same models. Right: contour map of the HOD fit, showing the $\Delta\chi^2 = 2.3$ and 6.2 levels, for the Bolshoi–Planck catalog (solid lines) and the Consuelo halo catalog (dotted lines).

5. Discussion

5.1. Environments of Local AGN

We have cross-correlated hard X-ray selected AGNs with 2MASS near-infrared-selected galaxies to constrain how an unbiased sample of local AGNs occupy their halos. Analyzing the sample in terms of an HOD model, we find that the number of AGN hosted in a halo roughly scales with halo mass, as is the case for the overall galaxy population. This is inconsistent with the notion that AGNs are predominantly in central galaxies (e.g., Starikova et al. 2011; Richardson et al. 2013), as our results suggest a significant fraction of AGN are in satellites. This agrees with several recent studies (Allevato et al. 2012; Oh et al. 2014; Silverman et al. 2014). Additionally, using a subhalo-based model that corrects for catalog incompleteness, we find that the host galaxy stellar mass distribution can determine the environments of AGNs on average, via the stellar mass-subhalo mass relation (Behroozi et al. 2010). This was also found when comparing predictions of this model with the weak gravitational lensing signal of X-ray-selected AGNs in COSMOS (Leauthaud et al. 2015).

The typical halo mass found for the BASS AGN with our HOD analysis, $\log M_h/M_\odot h^{-1} = 12.8$, lies between those typically found for soft X-ray-selected AGNs ($\log M_h/M_\odot h^{-1} \sim 13$) and optically selected AGNs ($\log M_h/M_\odot h^{-1} \sim 12.5$), and thus is broadly consistent with earlier results from AGN clustering studies across a large range of luminosity and redshift (Croom et al. 2005; Gilli et al. 2005, 2009; Ross et al. 2009; Shen et al. 2009; Allevato et al. 2011, 2014; Krumpe et al. 2012). The typical halo masses of BASS AGNs correspond to galaxy group environments.

5.2. Obscured versus Unobscured Environments

We split our sample in two bins of N_{H} to test whether AGNs with different column densities (i.e., obscured versus unobscured) live in different environments, for samples matched in luminosity and redshift, in order to avoid bias in the observed volume; we note that the host galaxy stellar mass distributions are also similar (Figure 6). We find differences in their correlation functions, predominantly on small scales. Our HOD

fits suggest that obscured AGNs live in more massive halos and in denser environments than unobscured AGNs.

The simplest unification models attribute obscuration to the circumnuclear material absorbing the radiation produced in the broad-line region. In that case, whether the AGN is observed as obscured or unobscured depends only on viewing angle (Urry & Padovani 1995), which means the halo-scale environments should be the same (statistically) for both populations. Although we now know that circumnuclear geometry is not the only factor, as there is a dependence of luminosity and λ_{Edd} on covering factor (e.g., Ricci et al. 2017c), the analysis of our matched samples shows that these factors are not biasing our results.

Large column densities can also come from the host galaxy (e.g., from a random molecular cloud that happens to lie along the line of sight or from the orientation of the galaxy disk). In the present case, only about 5% of the sample lies in an edge-on host galaxy, and the results do not change when those AGNs are removed. Another possibility is from inflowing gas following a merger (e.g., Hopkins et al. 2008; Kocevski et al. 2015; Ricci et al. 2017b). In general, because the probability of galaxy interactions depends on the environment (e.g., Shen et al. 2009; Jian et al. 2012), it is possible that either major mergers or smaller galaxy interactions play a role in causing the clustering difference.

DiPompeo et al. (2017b) found a similar clustering difference on large scales with *WISE* infrared-selected AGNs at $z \sim 1$. They interpreted their results as obscuration being an evolutionary phase of merger-driven quasar fueling, in which the quasar is first obscured, followed by an unobscured phase after gas “blow-out.” The resulting observations of obscured AGN living in larger halos would be a selection effect based on this model, by using luminosity-limited samples. Assuming this scenario for our AGN sample, the halo mass differences should be minimal at these low luminosities (~ 0.2 dex of $M_\odot h^{-1}$)—much less than our results based on the HOD analysis. It is unlikely that major mergers trigger these low-luminosity, low-redshift AGN. Indeed, only 8% of BAT AGNs are in the final phases of major mergers, where obscuration is found to peak (Koss et al. 2010).

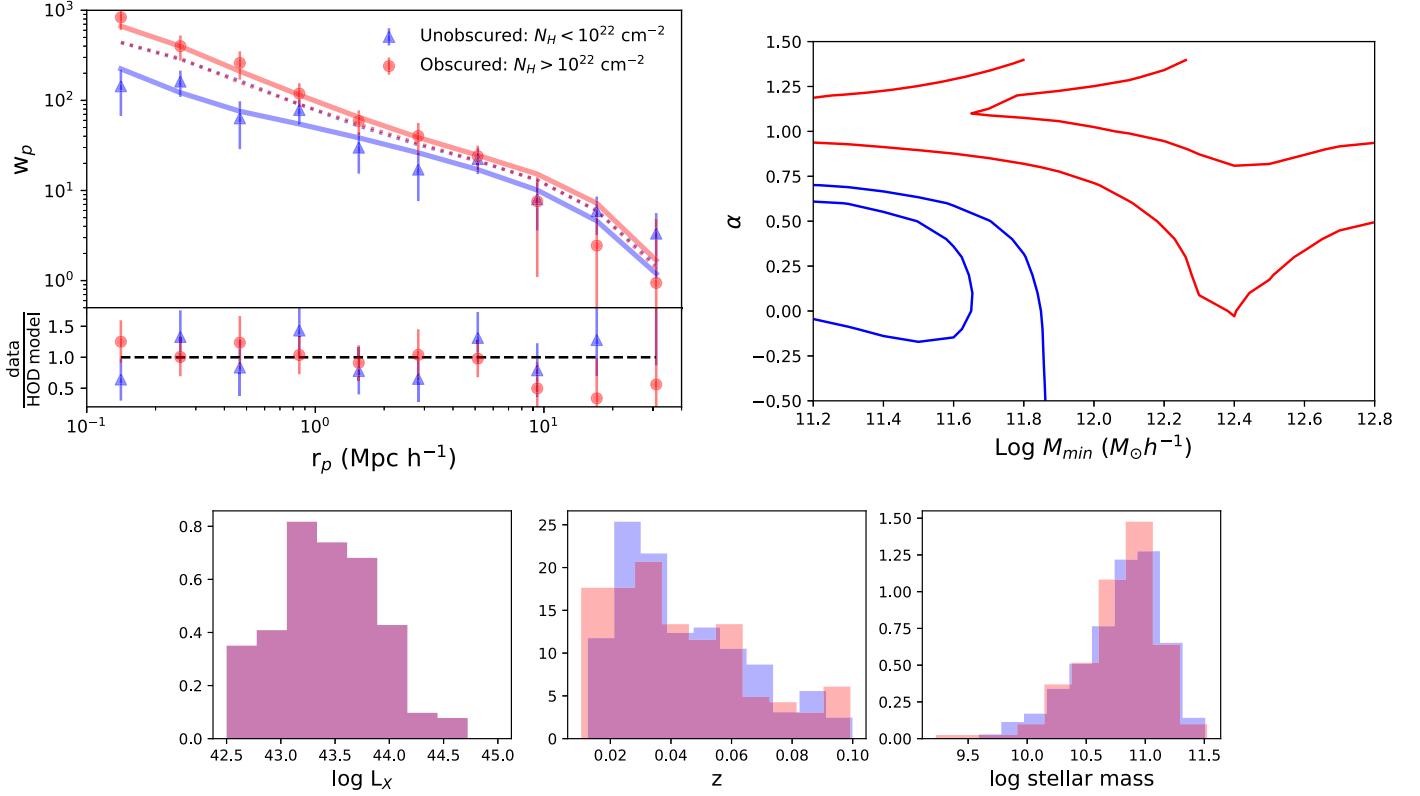


Figure 5. Upper panels: projected cross-correlation function of obscured (red) versus unobscured (blue) AGNs and corresponding HOD model fits (solid lines) and subhalo models (dotted points). While their two-halo terms are consistent with each other, obscured AGNs appear more clustered on scales of the one-halo term. Upper right: $\Delta\chi^2$ contour maps of the HOD fit for unobscured (blue) and obscured (red) AGNs are completely distinct, suggesting some difference between the two populations. Lower panels: matched subsamples have similar distributions of (left) log of the $L_{2-10\text{keV}}$ luminosity, (middle) redshift, and (right) log of the host galaxy stellar mass.

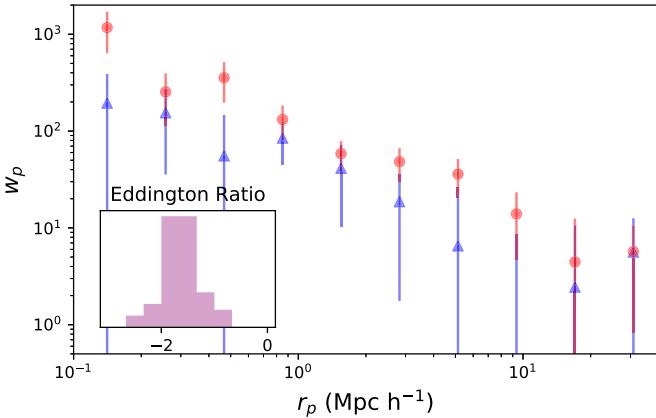


Figure 6. Projected cross-correlation function for Compton-thin obscured (red) and unobscured (blue) AGNs with matched distributions of Eddington ratio (inset). Although this analysis involved only half as many AGNs as in Figure 4, the difference is similar, indicating that black hole mass (and its possible relation to halo mass) is not causing the difference in clustering.

The shallow satellite power-law slope of unobscured AGNs, α , obtained from the HOD analysis, would suggest that the fraction of galaxies that host unobscured AGNs drops as a function of halo mass. This would mean that unobscured AGNs avoid the richest clusters. Because high velocity encounters in the largest clusters disfavor galaxy mergers, perhaps a higher fraction of unobscured AGNs were triggered by an earlier major merger, such that there was sufficient time to clear the surrounding gas and dust. Using the analytical function of the instantaneous galaxy merger rate from Shen (2009), we

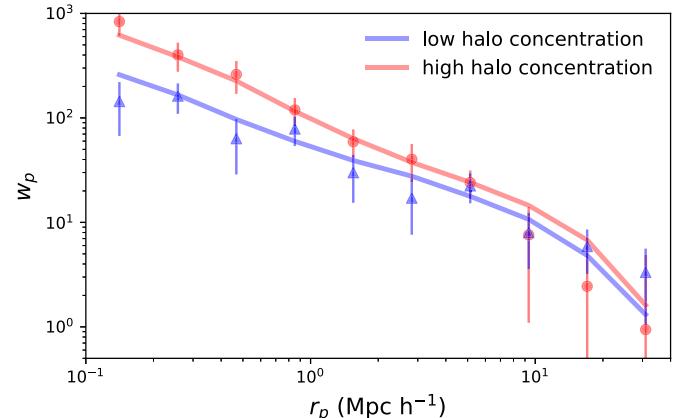


Figure 7. The clustering of obscured (red) and unobscured (blue) AGNs subsamples is well reproduced by a toy subhalo model split by host halo concentration ($c < 13.5$, blue; $c > 10.0$, red). This differs from the HOD interpretation that each population has distinct occupation statistics; rather, each population could reside in halos of statistically different concentrations (and hence different assembly histories).

Table 3
Parameters for the Best-fit Subhalo Models, Which Assume a Threshold Halo Concentration (a Maximum for the Unobscured Sample, and a Minimum for the Obscured Sample)

AGN Sample	c_{\min}	c_{\max}	$\langle c \rangle$	\tilde{M}_h	$\langle M_h \rangle$	χ^2
Obscured	$10.0^{+1.5}_{-2.0}$...	8.5	12.3	13.4	1.8
Unobscured	...	$13.5^{+2.0}_{-2.5}$	27.0	12.3	13.1	0.9

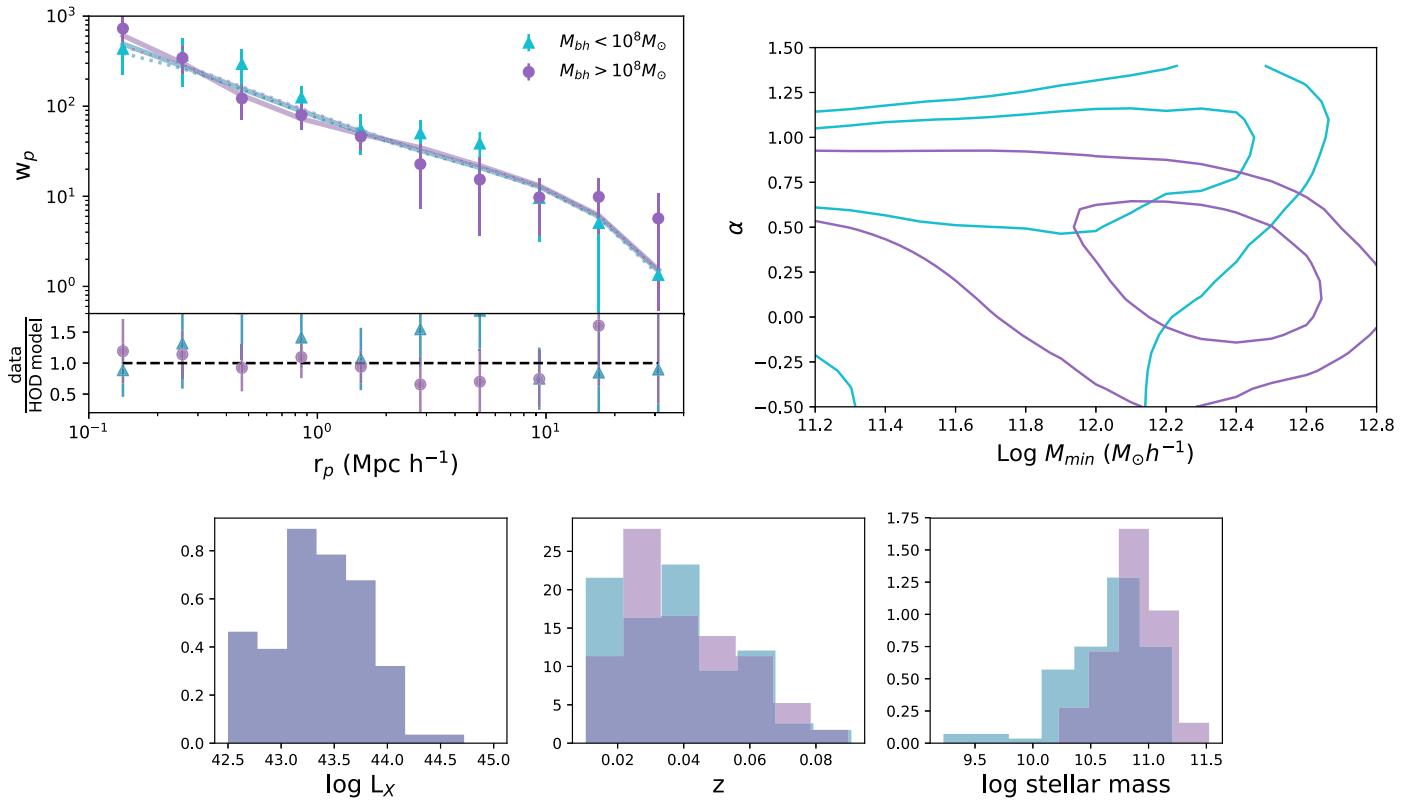


Figure 8. Upper left: projected cross-correlation function in two bins of black hole mass, $M_{\text{bh}} < 10^8 M_{\odot}$ (cyan) and $M_{\text{bh}} > 10^8 M_{\odot}$ (purple), with corresponding HOD model fits (solid lines) and subhalo models (dotted lines). Upper right: $\Delta\chi^2$ contour map of the HOD fit for each mass bin. Lower panels: distributions of the log of the X-ray luminosity (matched), redshift, and log of the host galaxy stellar mass.

estimate that major mergers occur roughly four times more often in halos of the average mass hosting unobscured AGNs than in halos hosting obscured AGNs around $z \sim 0.1$. However, we compared the halo masses for obscured and unobscured AGNs from the 2MASS group catalog (Lu et al. 2016) for the sources with counterparts in 2MASS, and found no evidence that obscured AGNs live in preferentially larger halos.

Alternatively, we have shown that a difference in halo concentration, opposed to differing halo occupation distributions and/or typical halo masses, fits the data equally well. Highly concentrated halos of a given mass would have a high concentration of satellite galaxies, and therefore have a higher probability of galaxy interactions (i.e., minor mergers and encounters, as opposed to major mergers that predominantly occurred at high redshifts). Indeed, Jiang et al. (2016) found that SDSS Type 2 satellites were more concentrated than Type 1 satellites, and Villarroel & Korn (2014) calculated an enhanced number of SDSS Type 2 versus Type 1 companions around Type 2 AGN. The excess of Compton-thick BASS AGNs in mergers would also support this scenario (Koss et al. 2016). However, after removing clear cases of mergers and interactions in the obscured sample by visual inspection, the clustering differences slightly increased rather than decreased—the opposite of what this scenario would predict. Additionally, the unobscured sample is the one that is more inconsistent with the clustering statistics expected for its stellar mass distribution, suggesting unobscured AGNs have more environmental dependencies than obscured AGNs.

Instead, the observed difference between the clustering of obscured and unobscured AGNs may be due to a difference in

their host halo assembly histories. Halo concentration correlates with formation epoch, and so unobscured AGNs tend to reside in halos that were assembled more recently in cosmic time than halos hosting obscured AGNs. This means that the merging of their subhalos, and hence the merging of the galaxies within these subhalos, occurred around $z \sim 1$, opposed to at higher redshift for obscured AGN host halos. Therefore, the progenitors of $z = 0$ unobscured AGNs underwent major merger events statistically more recently than obscured AGNs. If the major mergers triggered a powerful quasar that blew away much of the surrounding gas and dust, then it would explain the lower column densities we see in AGNs in recently formed halos. Perhaps obscured AGN host halos had, on average, a more quiescent history dominated by secular processes, allowing nuclear obscuring material to remain. This scenario, with the different host halo histories rather than AGN triggering processes, explains the distinct clustering signatures we see for unobscured and obscured AGNs at $z \sim 0$. However, it is uncertain if this explanation is consistent with higher redshift studies (e.g., Allevato et al. 2014); an investigation of obscured versus unobscured AGNs clustering with samples of matched stellar mass distributions across a wide range of redshift (and luminosity) is needed.

5.3. Possible Dependence of Environment on Black Hole Mass

There is a small ($\sim 1\sigma$) difference between the clustering of AGNs with black holes of mass $< 10^8 M_{\odot}$ and $> 10^8 M_{\odot}$. The flatter satellite power-law slope indicated by our analysis may suggest that larger black holes tend to lie in central galaxies rather than satellites, while smaller black holes tend to lie in satellites. A correlation between the SMBH and the mass of its

subhalo goes in the right direction. However, more data are clearly necessary to confirm this weak signal.

6. Summary

In this study, we characterized the environments of a sample of accreting SMBHs unbiased toward obscuration by measuring the cross-correlation function of BASS AGN and 2MASS galaxies. We compared our results to mock samples created from simulations in order to model how AGNs occupy their host dark matter halos.

From fitting an HOD model to the cross-correlation function of the full sample, and by comparing with a subhalo model that assumed only stellar mass determines clustering statistics, we concluded that BASS AGNs, on average, occupy dark matter halos consistently with the overall inactive galaxy population.

However, subsamples based on column density and black hole mass have differing clustering statistics. We found that absorbed AGNs reside in denser environments than unabsorbed AGNs, despite no significant difference in their luminosity, redshift, or stellar mass distributions. Our subhalo model analysis suggests they may reside in halos with statistically different concentrations/assembly histories. The alternative interpretation from the HOD analysis—that they have systematically different halo occupation distributions and host halo masses—seems to contradict the finding that stellar mass drives the clustering amplitude. Lastly, we found a hint that a larger fraction of high-mass black holes ($M > 10^8 M_{\odot}$) reside in central galaxies than lower mass black holes.

M.P. would like to thank Andrew Hearin for helpful discussions. M.P., N.C., and C.M.U acknowledge support from NASA-SWIFT GI: Nr. 80NSSC18K0505, NSF grant 1715512, NASA CT Space Grant, and Yale University. M.K. acknowledges support from NASA through ADAP award NNH16CT03C, and C.R. acknowledges support from FONDECYT 1141218, CONICYT PAI77170080, Basal-CATA PFB-06/2007, and the China-CONICYT fund.

Software: CorrFunc (Sinha & Garrison 2017), Halotools (Hearin et al. 2017), Astropy (Astropy Collaboration et al. 2013), Matplotlib (Hunter 2007).

ORCID iDs

M. C. Powell [DOI](https://orcid.org/0000-0003-2284-8603) <https://orcid.org/0000-0003-2284-8603>
 N. Cappelluti [DOI](https://orcid.org/0000-0002-1697-186X) <https://orcid.org/0000-0002-1697-186X>
 C. M. Urry [DOI](https://orcid.org/0000-0002-0745-9792) <https://orcid.org/0000-0002-0745-9792>
 M. Koss [DOI](https://orcid.org/0000-0002-7998-9581) <https://orcid.org/0000-0002-7998-9581>
 A. Finoguenov [DOI](https://orcid.org/0000-0002-4606-5403) <https://orcid.org/0000-0002-4606-5403>
 C. Ricci [DOI](https://orcid.org/0000-0001-5231-2645) <https://orcid.org/0000-0001-5231-2645>
 B. Trakhtenbrot [DOI](https://orcid.org/0000-0002-3683-7297) <https://orcid.org/0000-0002-3683-7297>
 V. Allevato [DOI](https://orcid.org/0000-0001-7232-5152) <https://orcid.org/0000-0001-7232-5152>
 M. Ajello [DOI](https://orcid.org/0000-0002-6584-1703) <https://orcid.org/0000-0002-6584-1703>
 K. Oh [DOI](https://orcid.org/0000-0002-5037-951X) <https://orcid.org/0000-0002-5037-951X>
 K. Schawinski [DOI](https://orcid.org/0000-0001-5464-0888) <https://orcid.org/0000-0001-5464-0888>
 N. Secrest [DOI](https://orcid.org/0000-0002-4902-8077) <https://orcid.org/0000-0002-4902-8077>

References

Ajello, M., Costamante, L., Sambruna, R. M., et al. 2009, *ApJ*, 699, 603
 Allevato, V., Finoguenov, A., Cappelluti, N., et al. 2011, *ApJ*, 736, 99
 Allevato, V., Finoguenov, A., Civano, F., et al. 2014, *ApJ*, 796, 4
 Allevato, V., Finoguenov, A., Hasinger, G., et al. 2012, *ApJ*, 758, 47
 Assef, R. J., Kochanek, C. S., Brodin, M., et al. 2010, *ApJ*, 713, 970
 Astropy Collaboration, Robitaille, T. P., Tollerud, E. J., et al. 2013, *A&A*, 558, A33
 Barthelmy, S. D., Barbier, L. M., Cummings, J. R., et al. 2005, *SSRv*, 120, 143
 Baumgartner, W. H., Tueller, J., Markwardt, C. B., et al. 2013, *ApJS*, 207, 19
 Behroozi, P. S., Conroy, C., & Wechsler, R. H. 2010, *ApJ*, 717, 379
 Behroozi, P. S., Wechsler, R. H., & Wu, H.-Y. 2013, *ApJ*, 762, 109
 Bell, E. F., & de Jong, R. S. 2001, *ApJ*, 550, 212
 Bentz, M. C., Peterson, B. M., Netzer, H., Pogge, R. W., & Vestergaard, M. 2009, *ApJ*, 697, 160
 Blanton, M. R., & Roweis, S. 2007, *AJ*, 133, 734
 Booth, C. M., & Schaye, J. 2010, *MNRAS*, 405, L1
 Cappelluti, N., Ajello, M., Burlon, D., et al. 2010, *ApJL*, 716, L209
 Civano, F., Marchesi, S., Comastri, A., et al. 2016, *ApJ*, 819, 62
 Coil, A. L., Georgakakis, A., Newman, J. A., et al. 2009, *ApJ*, 701, 1484
 Cooray, A., & Sheth, R. 2002, *PhR*, 372, 1
 Croton, S. M., Boyle, B. J., Shanks, T., et al. 2005, *MNRAS*, 356, 415
 Dalal, N., White, M., Bond, J. R., & Shirokov, A. 2008, *ApJ*, 687, 12
 DiPompeo, M. A., Hickox, R. C., Eftekharzadeh, S., & Myers, A. D. 2017a, *MNRAS*, 469, 4630
 DiPompeo, M. A., Hickox, R. C., Myers, A. D., & Geach, J. E. 2017b, *MNRAS*, 464, 3526
 DiPompeo, M. A., Myers, A. D., Hickox, R. C., Geach, J. E., & Hainline, K. N. 2014, *MNRAS*, 442, 3443
 El-Zant, A. A., Shlosman, I., Begelman, M. C., & Frank, J. 2003, *ApJ*, 590, 641
 Gao, L., Springel, V., & White, S. D. M. 2005, *MNRAS*, 363, L66
 Gehrels, N., Chincarini, G., Giommi, P., et al. 2004, *ApJ*, 611, 1005
 Gilli, R., Daddi, E., Zamorani, G., et al. 2005, *A&A*, 430, 811
 Gilli, R., Zamorani, G., Miyaji, T., et al. 2009, *A&A*, 494, 33
 Greene, J. E., & Ho, L. C. 2005, *ApJ*, 630, 122
 Hearin, A. P., Campbell, D., Tollerud, E., et al. 2017, *AJ*, 154, 190
 Hickox, R. C., Jones, C., Forman, W. R., et al. 2009, *ApJ*, 696, 891
 Hopkins, P. F., Hernquist, L., Cox, T. J., & Kereš, D. 2008, *ApJS*, 175, 356
 Huchra, J. P., Macri, L. M., Masters, K. L., et al. 2012, *ApJS*, 199, 26
 Hunter, J. D. 2007, *CSE*, 9, 90
 Jian, H.-Y., Lin, L., & Chiueh, T. 2012, *ApJ*, 754, 26
 Jiang, N., Wang, H., Mo, H., et al. 2016, *ApJ*, 832, 111
 Jin, C., Ward, M., & Done, C. 2012, *MNRAS*, 425, 907
 Jones, M. L., Hickox, R. C., Mutch, S. J., et al. 2017, *ApJ*, 843, 125
 Kaspi, S., Smith, P. S., Netzer, H., et al. 2000, *ApJ*, 533, 631
 Kocevski, D. D., Brightman, M., Nandra, K., et al. 2015, *ApJ*, 814, 104
 Komatsu, E., Dunkley, J., Nolta, M. R., et al. 2009, *ApJS*, 180, 330
 Kormendy, J., & Ho, L. C. 2013, *ARA&A*, 51, 511
 Koss, M., Mushotzky, R., Veilleux, S., & Winter, L. 2010, *ApJL*, 716, L125
 Koss, M., Trakhtenbrot, B., Ricci, C., et al. 2017, *ApJ*, 850, 74
 Koss, M. J., Assef, R., Baloković, M., et al. 2016, *ApJ*, 825, 85
 Krimm, H. A., Holland, S. T., Corbet, R. H. D., et al. 2013, *ApJS*, 209, 14
 Krumpe, M., Miyaji, T., Coil, A. L., & Aceves, H. 2012, *ApJ*, 746, 1
 Krumpe, M., Miyaji, T., Coil, A. L., & Aceves, H. 2018, *MNRAS*, 474, 1773
 Krumpe, M., Miyaji, T., Husemann, B., et al. 2015, *ApJ*, 815, 21
 Lacey, C. G., Baugh, C. M., Frenk, C. S., et al. 2008, *MNRAS*, 385, 1155
 Landy, S. D., & Szalay, A. S. 1993, *ApJ*, 412, 64
 Leauthaud, A., J. Benson, A., Civano, F., et al. 2015, *MNRAS*, 446, 1874
 Lu, Y., Yang, X., Shi, F., et al. 2016, *ApJ*, 832, 39
 Marchesi, S., Civano, F., Elvis, M., et al. 2016, *ApJ*, 817, 34
 Marconi, A., Risaliti, G., Gilli, R., et al. 2004, *MNRAS*, 351, 169
 Mejía-Restrepo, J. E., Trakhtenbrot, B., Lira, P., Netzer, H., & Capellupo, D. M. 2016, *MNRAS*, 460, 187
 Mendez, A. J., Coil, A. L., Aird, J., et al. 2016, *ApJ*, 821, 55
 Navarro, J. F., Frenk, C. S., & White, S. D. M. 1996, *ApJ*, 462, 563
 Oh, K., Koss, M., Markwardt, C. B., et al. 2018, *ApJS*, 235, 4
 Oh, S., Mulchaey, J. S., Woo, J.-H., et al. 2014, *ApJ*, 790, 43
 Peterson, B. M. 2014, *SSRv*, 183, 253
 Planck Collaboration, Ade, P. A. R., Aghanim, N., et al. 2016, *A&A*, 594, A13
 Ricci, C., Bauer, F. E., Treister, E., et al. 2017b, *MNRAS*, 468, 1273
 Ricci, C., Trakhtenbrot, B., Koss, M. J., et al. 2017a, *ApJS*, 233, 17
 Ricci, C., Trakhtenbrot, B., Koss, M. J., et al. 2017c, *Natur*, 549, 488
 Ricci, C., Ueda, Y., Koss, M. J., et al. 2015, *ApJL*, 815, L13
 Richardson, J., Chatterjee, S., Zheng, Z., Myers, A. D., & Hickox, R. 2013, *ApJ*, 774, 143
 Riebe, K., Partl, A. M., Enke, H., et al. 2013, *AN*, 334, 691
 Ross, N. P., Shen, Y., Strauss, M. A., et al. 2009, *ApJ*, 697, 1634
 Schlafly, E. F., & Finkbeiner, D. P. 2011, *ApJ*, 737, 103
 Shen, Y. 2009, *ApJ*, 704, 89
 Shen, Y. 2013, *BASI*, 41, 61
 Shen, Y., Strauss, M. A., Ross, N. P., et al. 2009, *ApJ*, 697, 1656
 Silk, J., & Rees, M. J. 1998, *A&A*, 331, L1

Silverman, J. D., Miniati, F., Finoguenov, A., et al. 2014, [ApJ](#), **780**, 67

Sinha, M., & Garrison, L. 2017, Corrfunc: Blazing fast correlation functions on the CPU, Astrophysics Source Code Library, ascl:[1703.003](#)

Starikova, S., Cool, R., Eisenstein, D., et al. 2011, [ApJ](#), **741**, 15

Trakhtenbrot, B., & Netzer, H. 2012, [MNRAS](#), **427**, 3081

Treister, E., Schawinski, K., Urry, C. M., & Simmons, B. D. 2012, [ApJL](#), **758**, L39

Treister, E., Urry, C. M., Chatzichristou, E., et al. 2004, [ApJ](#), **616**, 123

Urry, C. M., & Padovani, P. 1995, [PASP](#), **107**, 803

Vasudevan, R. V., & Fabian, A. C. 2007, [MNRAS](#), **381**, 1235

Villarroel, B., & Korn, A. J. 2014, [NatPh](#), **10**, 417

Wechsler, R. H., Bullock, J. S., Primack, J. R., Kravtsov, A. V., & Dekel, A. 2002, [ApJ](#), **568**, 52

Wright, E. L., Eisenhardt, P. R. M., Mainzer, A. K., et al. 2010, [AJ](#), **140**, 1868

Xiao, T., Barth, A. J., Greene, J. E., et al. 2011, [ApJ](#), **739**, 28

Zehavi, I., Zheng, Z., Weinberg, D. H., et al. 2011, [ApJ](#), **736**, 59

Zheng, Z., Coil, A. L., & Zehavi, I. 2007, [ApJ](#), **667**, 760

Zheng, Z., & Guo, H. 2016, [MNRAS](#), **458**, 4015