Photometric Redshift Biases from Galaxy Evolution

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ABSTRACT. Proposed cosmological surveys will make use of photometric redshifts of galaxies that are significantly fainter than any complete spectroscopic redshift surveys that exist to train the photo-z methods. We investigate the photo-z biases that result from known differences between the faint and bright populations: a rise in active galactic nucleus (AGN) activity toward higher redshift, and a metallicity difference between intrinsically luminous and faint early-type galaxies. We find that even very small mismatches between the mean photometric target and the training set can induce photo-z biases large enough to corrupt derived cosmological parameters significantly. Our results suggest that a metallicity shift of \( \sim 0.003 \) dex in an old population, or contamination of any galaxy spectrum with \( \sim 0.2\% \) AGN flux, is sufficient to induce a \( 10^{-3} / C_0 \) bias in photo-z. These results highlight the danger in extrapolating the behavior of bright galaxies to a fainter population, and the desirability of a spectroscopic training set that spans all of the characteristics of the photo-z targets, i.e., extending to the 25th mag or fainter galaxies that will be used in future surveys.

Online material: color figures

1. INTRODUCTION

Photometric redshifts have been used extensively in studies of galaxy evolution, particularly for sources that are too faint for feasible spectroscopic detection of absorption or emission lines. Photo-z’s are just beginning to find application to measurements of cosmic structure, particularly in the determination of source redshifts for weak gravitational lensing surveys (e.g., Brown et al. 2003; Massey et al. 2007). Projects in development plan to measure photometric redshifts for \( \sim 10^9 \) galaxies, far more than can be practically determined spectroscopically.

Yet the cosmological measures that weak-lensing surveys have as goals are extremely sensitive to biases or other miscalibrations of the photo-z scale: biases as small as \( \approx 10^{-3}(1 + z) \) will overwhelm the statistical errors on cosmology in these experiments (Huterer et al. 2006). The photo-z technique has never been subject to such demands on its accuracy, so we need to examine carefully the strategies for achieving them.

A photo-z algorithm must somehow be taught to convert broadband fluxes into redshifts. This can be done by fitting to theoretical or empirical templates for galaxy spectra (e.g., Benítez 2000) or by completely empirical machine-learning techniques, e.g., neural nets (Collister & Lahav 2004), that are trained with a sample of galaxies of known spectroscopic redshift. There is a difficulty, however, in that many imaging surveys intend to measure photo-z’s from galaxies that are too faint to obtain the spectroscopic information needed to form templates or obtain spectroscopic training redshifts. One must therefore take on some degree of faith the assertion that a photo-z algorithm trained on brighter galaxies will return sufficiently unbiased redshifts for galaxies beyond the spectroscopic limit.

We investigate in this article the sizes of biases that we might expect to arise because of differences between the faint-galaxy population and the brighter spectroscopic population. In particular,

- Fainter galaxies will tend to be at higher redshifts, where AGNs are more frequent or brighter than at low redshift. Furthermore, optically based spectroscopy is inefficient in the “desert” between \( 1.5 \lesssim z \lesssim 3 \), where quasar activity peaks. We can ask, therefore, whether the different AGN characteristics of the fainter galaxies could bias the photo-z’s trained on low-z galaxies.
- For elliptical galaxies, there is a well-known color-magnitude relation. As we move to lower luminosity galaxies in this red sequence, stellar populations traverse a path in the age-metallicity plane. Stellar populations of the luminous galaxies of different Hubble types that would typically be used for photo-z training will not cover the same locus in the age-metallicity plane. Would photo-z’s trained on bright galaxies produce biased redshifts for fainter ellipticals? We investigate this by examining the metallicity dependence of photo-z results.


for old stellar populations when training on the bright Hubble sequence. There are additional complications from the age-metallicity degeneracy. Accordingly, we also estimate the age dependent behavior of photo-z biases.

In the Rumsfeld parlance, these are “known unknowns” of the faint-galaxy population: a photo-z algorithm might be designed that could compensate for these two differences between faint and bright galaxies. But we will use a photo-z algorithm that is ignorant of these effects, so that they can serve as models for “unknown unknowns” that might bias faint photo-z estimates. In this way we might learn how safe or dangerous it is to extrapolate training sets or templates to fainter magnitudes.

2. PHOTO-Z FITTING TECHNIQUE AND TEMPLATE SETS

Probing AGN contamination biases necessitates the preparation of an appropriate galaxy spectrum and the use of fitting software to determine the photometric redshift $z_p$. Contaminated galaxies are modeled by introducing a quasi-stellar object (QSO) spectrum as an impurity to the spectral energy distributions (SEDs) of galaxies. The Le Phare$^1$ software package is used to compute the photometric redshifts. Le Phare compares theoretical and known source magnitudes, estimating $z_p$ via $\chi^2$ minimization (Arnouts et al. 1999, 2002).

Le Phare fits to a magnitude catalog of template galaxies. In this case, the catalog consists of four known galaxies from Coleman et al. (1980, hereafter CWW) and three simulated galaxies from A. G. Bruzual & S. Charlot (1996, unpublished, hereafter BC96), as well as linear interpolations (via flux) of these. The CWW galaxies are early types, Sbc and Scd spirals, and irregulars, while the BC96 galaxies are 2, 0.5, and 0.05 Gyr in age. Altogether, including interpolations, the catalog is composed of 72 models. We investigate two sets of filters. The first includes four Hubble Deep Field–North (HDF-N) filters (F300W, F450W, F606W, and F814W) and three near-infrared filters (Jbb, H, and K) included in the Le Phare software package. The second set is the five Sloan Digital Sky Survey (SDSS) filter set (u, g, r, i, and z). Le Phare compiles this catalog by calculating the magnitudes for each galaxy SED with each filter at redshifts from 0 to 6 in increments of 0.04. Note that the template galaxies contain no AGN signal.

A second input catalog describes the galaxies for which $z_p$ is to be estimated. We synthesize AGN-contaminated magnitudes by making a QSO template spectrum, normalizing this template to the galaxy SEDs, contaminating the Le Phare templates with a specified fraction of AGN flux, and then calculating apparent magnitudes at various redshifts and AGN contaminations.

A template QSO spectrum over a wide wavelength span is needed in order to synthesize AGN-contaminated SEDs.

Vanden Berk et al. (2001) and Glikman et al. (2006) provide suitable QSO template spectra; the former covers a broad UV–NIR (800.5–8554.5 Å) wavelength range, while the latter extends further into the NIR (5801–35,095.4 Å). We scale the Glikman spectrum to match the Vanden Berk spectrum at 847 nm, where both are well measured, and then concatenate the two.

The QSO contamination is specified by the fraction $0 \leq h \leq 0.2$ of the flux contributed by the QSO. Specifically, for a chosen galaxy template $g(\lambda)$ and our template QSO spectrum $q(\lambda)$ we first determine the normalization $N$, which yields

$$\int_{\lambda_1}^{\lambda_2} g(\lambda) d\lambda = N \int_{\lambda_1}^{\lambda_2} q(\lambda) d\lambda, \quad (1)$$

where the integration bounds $\lambda_1 = 80$ nm to $\lambda_2 = 3.5 \mu m$ span the range over which both spectra are known. For a desired contamination $h$, the galaxy rest-frame SED is taken as

$$f(\lambda) = (1 - h)g(\lambda) + Nhq(\lambda). \quad (2)$$

Apparent magnitudes in the chosen filter sets are calculated using the standard formulas for redshifted sources.

We use the same QSO template spectrum for all levels of contamination. In reality, low-luminosity AGN will have different spectra than bright QSOs, e.g., weaker broad lines; however, the effect of contamination on photo-z’s that we derive will still be crudely correct.

The catalog of AGN-contaminated galaxies contains seven galaxy types—the four CWW templates (of type E, Sbc, Scd, and Irr) and the three BC96 simulated star forming galaxies (denoted I2, I05, and I005 for their ages)—at redshifts in the range $0 \leq z \leq 3$ in steps of 0.1 for SDSS filters and in the ranges $0 \leq z \leq 1$ and $1 \leq z \leq 3$ in steps of 0.1 and 0.2, respectively, for the HDF-N filters, with AGN contaminations in the range $0 \leq h \leq 0.2$ in 0.01 increments.

We run Le Phare to obtain estimates for $z_p$. The quantity of interest is the redshift bias between the nominal and photometric redshifts, to wit, $\Delta z = z_p - z$.

3. BIAS FROM AGN CONTAMINATION

All galaxy types using either filter set have a nontrivial bias $\Delta z$ induced by AGN contamination. We first examine how $\Delta z$ depends on contamination fraction $h$ at fixed $z$ and galaxy type. Figure 1 illustrates the range of behaviors found. The I2 galaxy shows a simple linear bias-contamination response with a well-defined $d(\Delta z)/dh$. In cases like the Scd galaxy at $z = 0.7$, the linear trend gains steps or wiggles, which we believe is induced by the discrete steps in galaxy type that Le Phare takes when finding a best-fit template. In this case we define $d(\Delta z)/dh$ by a least-squares fit through all the data points. Finally there are cases like the high redshift E-type in which the best-fit photo-z becomes “catastrophically” incorrect above some
contamination level. Here the AGN light has moved the galaxy toward some new part of the color manifold of the templates rather than inducing a small differential change. We define $d(\Delta z)/dh$ by manually selecting a region of $h$ that is below the catastrophe and fitting a slope to this region. We essentially ignore the catastrophic failures because we are more interested in biases induced by gradual evolution in the AGN fraction. We will focus henceforth on the slope $d(\Delta z)/dh$ in the vicinity of $h = 0$.

Figure 2 shows $d(\Delta z)/dh$ from the low-contamination domain in both filter sets. The different templates show a variety of behaviors, and some are erratic due to discreteness in the Le Phare fitting, particularly with the SDSS filter set, which offers poor photo-z constraints in some domains due to lack of NIR coverage.

The important point, however, is that the bias responses $d(\Delta z)/dh$ are $\approx \pm 0.5$ over most of the color and galaxy-type domain. A 1% contamination with AGN light causes a photo-z bias of $\Delta z \sim 0.005$ if the photo-z algorithm is using uncontaminated templates. Application of photo-z’s to cosmic-shear measurements requires that biases be held below $\sim 0.001$, which means that the mean AGN contamination of the target sample would need to be determined to $\ll 1\%$ in order for the resultant photo-z biases to make insignificant impact on the inferred cosmology. We will not know the faint-galaxy AGN fraction to anywhere near this precision, so this would be a major problem.

Of course we are using a naive photo-z estimator: one could include a QSO template in the fitting process (as in the option Le Phare offers, or as described in Wolf et al. [2001], which includes object classification), or attempt to construct an empirical training set of bright galaxies that spans the range of AGN contamination of the faint set. This would greatly reduce the photo-z bias in our simple simulated catalog. The behavior of real AGN is, however, much more complex than our simple single QSO template, and it is likely that the faint population has different AGN spectra than the bright galaxies that might serve as templates. We simply point out that photo-z biases can be very sensitive to AGN light, and it is risky to extrapolate AGN characteristics to a new population.

4. AGE-METALLICITY MODELS

We use stellar population models to estimate the redshift bias induced by a metallicity mismatch between target galaxies and Le Phare’s templates (Le Phare uses the same extended CWW catalog used previously). The galaxy evolution simulation code GALAXEV$^2$ is used here to generate a magnitude catalog for galaxies with near-solar metallicity with a $z = 0$ age of 10 Gyr. We use the standard model as described in Bruzual & Charlot (2003, hereafter, BC03). Specifically, the galaxy models employed here use the STELIB and BaSeL spectral libraries, a simple stellar population (SSP) star formation rate, and

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$^2$ Available at http://www2.iap.fr/users/charlot/bc2003/.
the Chabrier initial mass function, and the 1994 Padova evolutionary tracks (per recommendation of the code’s authors).

GALAXEV produces a table of magnitudes and ages at various redshifts for a given galaxy model and filter set. The same SDSS and infrared supplemented HDF-N filters from the AGN contamination study are used to calculate these magnitudes. GALAXEV produces colors for metallicities $Z/Z_\odot = 0.2$, 0.4, 1.0, and 2.5 (models m42, m52, m62, and m72 in GALAXEV, respectively). We also produce colors for metallicities 0.3, 0.7, and $1.75Z_\odot$ by interpolating between these four GALAXEV spectra.

We further employ GALAXEV to investigate biases due to age misestimates. We generate solar-metallicity models with $z = 0$ ages of 11, 9, and 8 Gyr. Note that the fiducial model here is still the 10 Gyr old solar-metallicity galaxy, but in contrast to what is done in the metallicity study, we now hold the metallicity fixed and vary the age.

5. BIAS FROM METALLICITY INDETERMINACY

Calculating the bias due to a small metallicity mismatch requires removing the fiducial bias present resulting from inherent differences between the GALAXEV spectra and Le Phare’s catalog of templates. At each redshift, we take an elliptical galaxy with solar metallicity as our benchmark. The solar bias, $\delta z_\odot$, is defined as the difference between the solar model’s redshift and Le Phare’s estimate: $\delta z_\odot = z_{\odot, \text{mod}} - z_{\odot, \text{phot}}$. For any particular metallicity, the bias due to metallicity mismatch is then $\Delta z = \delta z_\odot$. Hence by definition we assign zero metallicity bias to a solar-metallicity population, for any chosen filter set, population age, and redshift.

Figure 3 illustrates the biases from metallicity offsets as a function of redshift between $z = 0$ and $z = 0.7$. Five models are present in the plot; each is 10 Gyr old at $z = 0$ with metallicities $0.2$–$2.5Z_\odot$. As the models get more metal rich, the redshift bias increases, i.e., the photometric redshift gets larger than the model’s redshift. In the redshift range that was studied, the super- and subsolar models both had biases of up to $0.2$ for metallicities $\approx \pm 0.5$ dex away from solar. This suggests that biases of $\approx 0.001(1 + z)$ would result from shifts of only $\approx 0.003$ dex in the mean metallicity of an old population away from the metallicity of the training set. For comparison, current modeling of age and metallicity variation along the red sequence of SDSS galaxies suggests a $\approx 0.3$ dex change in metallicity across four magnitudes of galaxy luminosity (Bernardi et al. 2005). Therefore the biases in photo-z’s of faint red-sequence galaxies can be highly significant if the photo-z’s are trained exclusively on brighter red-sequence members.

6. BIAS FROM AGE INDETERMINACY

Biases from age estimates errors are computed in the same manner as in the metallicity case. The intrinsic bias from the benchmark 10 Gyr solar-metallicity galaxy, $\delta z_{10 \text{ Gyr}}$, must be subtracted from the total, $\delta z_{\text{mod}}$, to arrive at the bias attributable to age mismatching:

$$\Delta z_{\text{age}} = z_{\text{mod}} - z_{\text{phot}} - \delta z_{10 \text{ Gyr}} = \delta z_{\text{mod}} - \delta z_{10 \text{ Gyr}}.$$

Figure 4 portrays the induced biases in both filter sets. Compared to the metallicity-induced biases, these show little structure outside of the feature near $z = 0.4$ for the SDSS filters (also found in the metallicity plots), perhaps suggesting that the photo-z’s are more sensitive to metallicity offsets. For the HDF-N filter set, the bias only becomes comparable at higher $z$ when the age offset is 2 Gyr. In both studies, the HDF-N sets exhibit a slightly wider range in biases, which may mean that extra bands may be more confusing for Le Phare if the templates and data are not initially well matched.

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**Fig. 3.—** Metallicity-induced biases are shown from rest frame to $z = 0.7$ for HDF-N (left) and SDSS (right) filters. Each model’s metallicity is shown on the plot in solar units. The dashed and dotted lines serve to distinguish between the models in the crowded regions. The smoother behavior of the HDF-N + IR plot is attributable to the supplementary IR bands—the HDF-N data acquire jagged profiles when the analysis is done sans the IR bands. Indeed, these plots (not shown) resemble the SDSS plots; while the overall range of the bias decreases, the curves behave more erratically.

**Fig. 4.—** Age-induced biases for solar-metallicity models with $0 \leq z \leq 0.7$ for the HDF-N (left) and SDSS (right) filter sets. Model ages are distinguished by gray scale. As with the metallicity-induced bias, a reanalysis of the HDF-N data with NIR data excluded yields plots comparable to the SDSS. See the electronic edition of the PASP for a color version of this figure.
7. CONCLUSION

Photometric redshifts resemble the real redshifts only as closely as the training sets mimic the objects of interest. In the cases of AGN activity and metallicity variation, we find that even very small mismatches between the mean photometric target and the training set can induce photo-z biases large enough to corrupt significantly the validity of derived cosmological parameters. In particular, our findings suggest that a metallicity shift of $\sim$0.003 dex in an old population, or contamination of any galaxy spectrum with $\sim$0.2% AGN flux, is sufficient to induce a $10^{-3}$ bias in photo-z. Miscalculating the age of a galaxy by 2 Gyr may also result in similar a bias. In a real survey, we can expect differences far larger than this between the faint photometry survey targets and the bright spectroscopic targets used for photo-z training.

While simple models are used here to study redshift biases, the results indicate a need to develop training sets that encompass the full range of behavior of the photo-z target population. Our study uses a worst-case situation in that the photo-z algorithm is given no information on the physical effects underlying the bias (AGN or a luminosity-metallicity correlation). In reality the training sets can include galaxies with AGN of varying brightness and some galaxies of lower luminosity, which would ameliorate the biases we have found. However the extreme sensitivity of the photo-z to these effects suggests that it will be dangerous to extrapolate from the behavior of bright training-set galaxies to a target photo-z sample that extends to significantly higher redshift and lower luminosity because even subtle differences in spectral behavior can lead to important biases. Furthermore we have investigated two known differences between the faint and bright populations, but there may be differences that have not yet been discovered because there are no detailed spectroscopic surveys of the galaxies fainter than 24th magnitude that will comprise the bulk of future weak-lensing surveys. These results highlight the desirability of training photo-z’s with complete spectroscopic surveys to $\sim$25 mag, so that the training set is representative of the galaxies for which photo-z’s are being obtained.

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