Overview of HSC photo-z's

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The question



Omega_M

The problem



redshift

HSC photo-z paper (arXiv: 1704.05988)

Photometric Redshifts for the Hyper Suprime-Cam Subaru Strategic Program Data Release 1

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Received ; Accepted

Abstract

Photometric redshifts are a key component of many science objectives in the Hyper Suprime-Cam Subaru Strategic Program (HSC-SSP). In this paper, we describe and compare the codes used to compute photometric redshifts for HSC-SSP, how we calibrate them, and the typical accuracy we achieve with HSC five-band photometry (*grizy*). We introduce a new point estimator based on an improved loss function and demonstrate that it works better than other commonly used estimators. We find that our photo-*z*'s are most accurate from $0.3 \leq z \leq 1.5$, where we can straddle the 4000Å break. We achieve $\sigma(\Delta z/(1+z)) \sim 0.04$ and an outlier rate of about 10% for galaxies down to i = 25 within this redshift range, which should enable many science cases for HSC-SSP. We also characterize the accuracy of our redshift probability distribution function (PDF) and discover that some codes over/under-estimate the redshift uncertainties, which have implications for N(z) reconstruction. Our photo-*z*'s for the Deep and UltraDeep layers are available in the public data release, while those for the Wide layer will soon be made available. Both our catalog products (such as point estimates) and full PDFs are available from the data release site, https://hsc-release.mtk.nao.ac.jp/.

Key words: surveys, galaxies: distances and redshifts, galaxies: general, cosmology: observations

Our photo-z's for Deep and UltraDeep are publicly available.

Photo-z's for the Wide layer will be made available later this month.

The HSC photo-z team



Masayuki Tanaka Template fitting code (Tanaka 2015)



Bau-Ching Hsieh DEmP (Hsieh+ 2014)



Atsushi Nishizawa MLZ (Carrasco Kind+ 14)



Jun Nakano Deep-learning (no paper yet)



Jean Coupon NNPZ (no paper yet)



Josh Speagle Franken-Z (paper in prep)



Sogo Mineo EPHOR (no paper yet)

Database tables

We have a database table for each code. We let science users choose the code.

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Photo-z production runs

After a data release, the photo-z team prepares

(a) training sample(b) target sample

I let people train their codes and do the production in any way they want, but I ask them to

(1) not to apply stringent cuts on the target sample
 (2) generate P(z) in a common format
 (3) write a release note
 (4) submit photo-z products by a deadline

This works reasonably well, but...

Training sample construction



- The Wide layer goes down to i~26.0.
- No spectroscopic sample goes down to that faint mags...
- We rely on many-band photo-z's from COSMOS.

Training sample construction



- The training sample is combination of spec-z, g/prism-z and high-accuracy photo-z from the entire survey footprint.
- We compute a weight for each object in order to reproduce the Wide galaxy distribution in the multi-color space.
- We suffer from the wiggles in dn/dz in COSMOS.

Code training

We employed the classical hold-out validation:



Each fold contains about 90k objects. One of us did cross-validation to get a sense of the performance and then used essentially all the training sample.

Lesson: ask someone outside of your photo-z group to keep the truth table for the test fold.

COSMOS wide-depth stacks



COSMOS visits taken under various seeing conditions

- 0.7 arcsec (median)
- 1.0 arcsec (worst)

Photo-z risk and the best point estimates

We take a minimum risk approach to make a point photo-z estimate and its photo-z *risk*. See Mineo-sensei's talk for details. We define the loss function as an inverted Lorentzian:

$$L(\Delta z) = 1 - \frac{1}{1 + \left(\frac{\Delta z}{\gamma}\right)^2}$$

$$\Delta z = (z_{\rm phot} - z_{\rm ref})/(1 + z_{\rm ref})$$

We then define a *risk* around a point-estimate:

$$R(z_{\rm phot}) = \int \mathrm{d}z P(z) L\left(\frac{z_{\rm phot} - z}{1 + z}\right)$$

The *best* point estimate is where the *risk* is the smallest:

$$z_{\rm best} = \operatorname{argmin}(R(z_{\rm phot}))$$

Some of our codes are optimized to minimize loss, but we could minimize delta_P(z).

And, it works!

Point Estimator	bias	σ_{conv}	$f_{outlier,conv}$	loss
mean	-0.003	0.075	0.227	0.260
mode	-0.002	0.067	0.213	0.244
median	-0.001	0.066	0.199	0.236
best	-0.003	0.064	0.197	0.233

z_best gives the smallest scatter and outlier rate!

z_risk works better than z_conf (see Mineo-sensei's talk)

Code performance : point estimates



- All galaxies down to i=25 (no clipping)
- Be careful not to interpret the absolute numbers at faint mags

Code performance : point estimates



Peak-peak variation of our photo-z accuracy. Not too bad.

Photo-z performance : PDF

$$PIT(z_{ref}) = \int_0^{z_{ref}} P(z)dz$$



PDF is over-dispersed

PDF is under-dispersed

Photo-z performance : PDF

$$PIT(z_{ref}) = \int_0^{z_{ref}} P(z) dz$$



PDF is over-dispersed

PDF is under-dispersed

PDF is too accurate

The Boldoroi corrections will be applied in our future photo-z's.

Weak-lensing tests

photo-z's for source selection



Stacked WL signal around CAMIRA clusters



color selection vs p(z) cut

$$\int_{z_{\rm lens}+0.2}^{1.3} \mathrm{d}P(z_p) z_p > 0.98$$

dNdz w/ re-weighting -> foreground contamination : ~ 11% (22% for no-cut)



Weak-lensing tests





- different colors : different photo-z codes
- No significant difference between photo-z codes

PDF files are getting very massive...

- The total file size of all the P(z)'s from all the codes amounts to ~1Tbyes (xz compressed).
- Need to compress further! Any good algorithms?

HSC public data release 1: https://hsc-release.mtk.nao.ac.jp/



Public Data Release 1

Welcome to the Hyper Suprime-Cam Subaru Strategic Program Data Release Site!

The first public release of HSC-SSP occurred on 28 February 2017. The release includes over 100 square degrees of deep multi-color data served through dedicated databases and user interfaces. The figures below shows the area covered in this release and the table gives an overview of the data in the three survey layers. Refer to **our survey website** for details of the survey design.

