

# 4300430: Introdução à Cosmologia Física

## Final Project: Photometric Redshifts (Photo-zs)

(2016)

In this project you will compute photometric redshifts (photo-zs) from the magnitudes of galaxies, using an Artificial Neural Network (ANN) code. Read the paper by Collister & Lahav 2004 (<https://arxiv.org/abs/astro-ph/0311058>) and download the corresponding ANNz code (<http://www.homepages.ucl.ac.uk/~ucapola/annz.html>). Read the README file, compile the code and run it on the example file.

Download the following files for training sets [here](#) and [here](#) and validation sets [here](#) and [here](#) (to be used for training the network). Download the testing set [here](#) (to be used for applying the network solution and quantifying photo-z errors). All these files contain magnitudes ( $u, g, r, i, z$ ), magnitude errors ( $\sigma_u, \sigma_g, \sigma_r, \sigma_i, \sigma_z$ ) and spectroscopic redshift ( $z^{\text{spec}}$ ).

### Basic Steps:

- Start using the **representative** training and validation files above to train and validate the network solution.
- Run the code to train and validate your photo-z solution and apply this solution to the testing set to compute photo-zs  $z_i^{\text{phot}}$  for  $i = 1 \dots N$  where  $N$  is the number of galaxies in the testing set.
- Define metrics to quantify the quality of the solution, e.g. the photo-z bias  $z^{\text{bias}}$

$$z^{\text{bias}} = \frac{1}{N} \sum_{i=0}^N (z_i^{\text{phot}} - z_i^{\text{spec}}) \quad (1)$$

and the photo-z scatter  $\sigma_z$

$$\sigma_z^2 = \frac{1}{N-1} \sum_{i=0}^N (z_i^{\text{phot}} - z_i^{\text{spec}})^2 \quad (2)$$

- Plot  $z^{\text{phot}}$  versus  $z^{\text{spec}}$  and indicate in the plot the  $y = x$  line as well as the values of the metrics above.
- The metrics above can also be defined in bins of  $z^{\text{spec}}$  or  $z^{\text{phot}}$ . Do this calculation in multiple bins of  $z^{\text{spec}}$  and  $z^{\text{phot}}$  and plot  $z^{\text{bias}}$  and  $\sigma_z$  as a function of  $z^{\text{spec}}$  and  $z^{\text{phot}}$ .
- Plot on a single figure the redshift distributions  $N(z^{\text{spec}})$  and  $N(z^{\text{phot}})$  as histograms for the testing set in bins of redshift (e.g. 20 bins of  $\Delta z = 0.1$  from  $z = 0$  to  $z = 2$ ).

**Investigate how metrics vary as a function of the following changes:**

- Changes in the number of multiple network solutions from different seeds that get combined into the final photo-z solution. Until now you had probably used a single solution.
- Changes in the network configuration, by changing the number of hidden layers and the number of nodes in these hidden layers.
- Changes in the size of the training and validation set used in the training/validation process. Consider fractions of these sets, e.g. 100%, 70%, 50%, 20%, 5% of the points.
- Changes in the maximum redshift used in the training/validation process.
- For the best case found, repeat the computation for **non-representative** training and validation sets. This represents a more realistic situation for real data analysis.
- Up to this point, you used ANNz to output a single number, i.e the galaxy photo-z  $z^{\text{phot}}$ . ANNz can also produce multiple outputs, for instance multiple points in the *redshift distribution*  $p(z_i)$  of a single galaxy. The overall redshift distribution  $N(z)$  can be estimated by summing  $p(z_i)$ 's of all galaxies. Think of how you could use the training and validation sets to estimate  $p(z_i)$ 's with ANNz. (*Hint*: What is the redshift distribution of a perfect spectroscopic redshift  $z^{\text{spec}}$  in the training set?)