



Reproducibility in the Era of Data Driven Science

software, data, publications

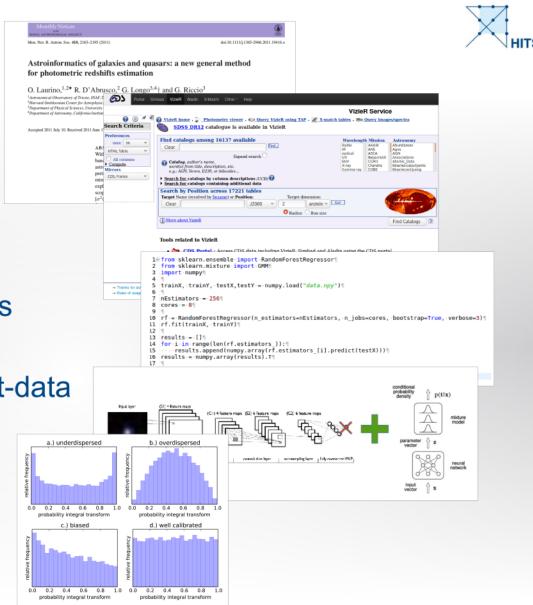
result = nearestNeigbours[self._skip.]
if self._identity == 1 and result[0][0
return result

][i]]), result[

Reproducibility

multiple aspects:

- publications
- data / catalogs
- software / code / algorithms
- parameters / training- / test-data
- statistical methods



Reproducibility





multiple aspects:



Photometric Redshift Estimation



Monthly Notices

of the ROYAL ASTRONOMICAL SOCIETY





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Astroinformatics of galaxies and quasars: a new general method for photometric redshifts estimation

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ABSTRACT

With the availability of the huge amounts of data produced by current and future large multiband photometric surveys, photometric redshifts have become a crucial tool for extragalactic astronomy and cosmology. In this paper we present a novel method, called Weak Gated Experts (WGE), which allows us to derive photometric redshifts through a combination of data mining techniques. The WGE, like many other machine learning techniques, is based on the exploitation of a spectroscopic knowledge base composed by sources for which a spectroscopic value of the redshift is available. This method achieves a variance $\sigma^2(\Delta z) = 2.3 \times 10^{-4}$ [$\sigma^2(\Delta z) = 0.08$, where $\Delta z = z_{\text{phot}} - z_{\text{spec}}$] for the reconstruction of the photometric redshifts

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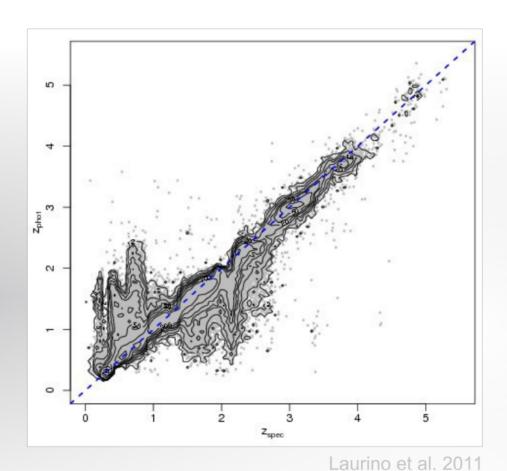
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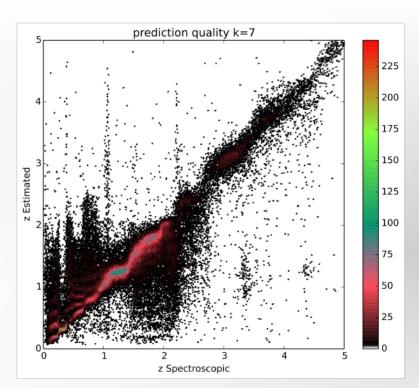
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Photometric Redshift Estimation





 $RMSE(\Delta z_{norm}) = 0.19$ $MAD(\Delta z_{norm}) = 0.041$

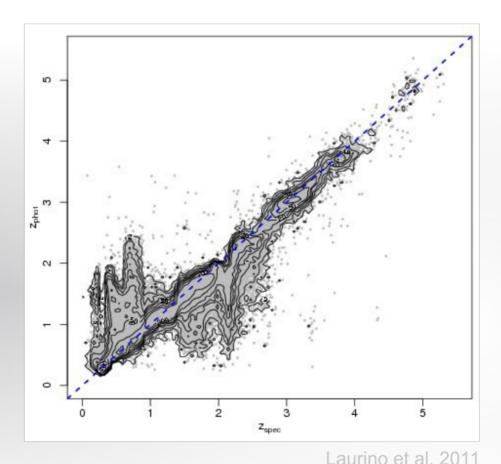


$$RMSE(\Delta z_{norm}) = 0.25$$

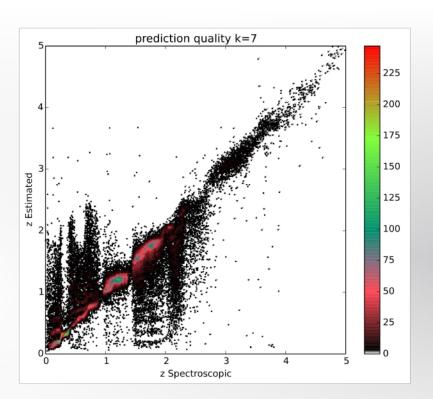
 $MAD(\Delta z_{norm}) = 0.048$

Photometric Redshift Estimation





 $RMSE(\Delta z_{norm}) = 0.19$ $MAD(\Delta z_{norm}) = 0.041$



$$RMSE(\Delta z_{norm}) = 0.22$$

 $MAD(\Delta z_{norm}) = 0.038$

Publishing Training- / Test-data



more than just:

- SDSS DR7 etc.
- VizieR ID
- select statement

required:

- catalogs
- labels / annotations
- data-bases
- reference data-sets
- scripts / pre-processing

```
SELECT TOP 200000
p.objid,p.ra,p.dec,p.u,p.g,p.r,p.i,p.z,
p.psfMag_u, p.psfMag_g, p.psfMag_r, p.psfMag_i,
p.psfMag_z, p.modelMag_u, p.modelMag_g, p.modelMag_r,
p.modelMag_i, p.modelMag_z, s.specobjid, s.class,
s.z AS redshift
INTO mydb.DR9_galaxies_with_modMag
FROM PhotoObj AS p
JOIN SpecObj AS s ON s.bestobjid = p.objid
WHERE s.z BETWEEN 0 AND 6.0
AND s.class = 'GALAXY'
```

ORDER BY NEWID()

Publishing and Sharing Code



simple code / simple example !?

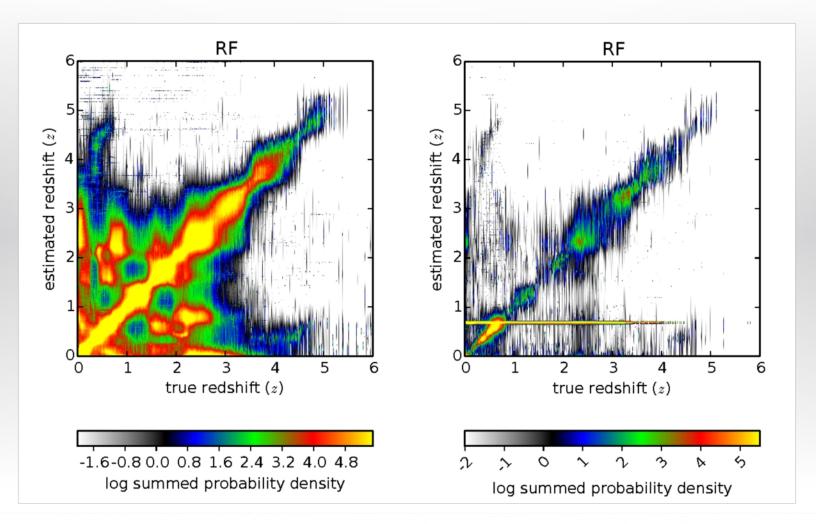
quantile regression forest for probabilistic predictions

```
1 from sklearn.ensemble import RandomForestRegressor
2 from sklearn.mixture import GMM¶
3 import numpy¶
5 trainX, trainY, testX,testY = numpy.load("data.npy")¶
7 nEstimators = 256¶
8 cores -- 8¶
10 rf = RandomForestRegressor(n estimators=nEstimators, n jobs=cores, bootstrap=True, verbose=3)¶
ll rf.fit(trainX, trainY)¶
12
13 results = []¶
14 for i in range(len(rf.estimators )):¶
15 ····results.append(numpy.array(rf.estimators [i].predict(testX)))¶
16 results = numpy.array(results).T¶
17 ¶
18 for i in range (len(testY)):¶
```

Publishing and Sharing Code



same code on 8x core and 120x core machine



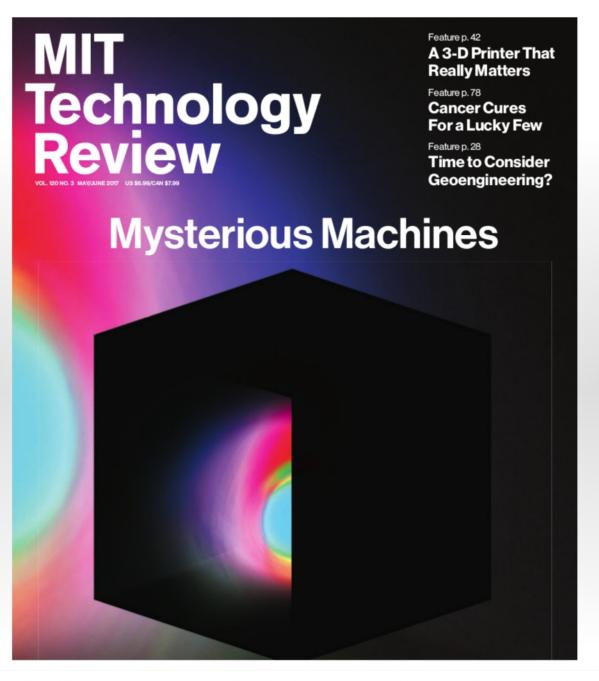
Publishing and Sharing Code



it is mandatory to publish code ...

- ... but how to publish
- environments
 - hardware
 - software / libraries
 - container
- compiler / interpreter
- test-data to verify the code is running as expected
- ... how to preserve and publish functionality?







Machine Learning

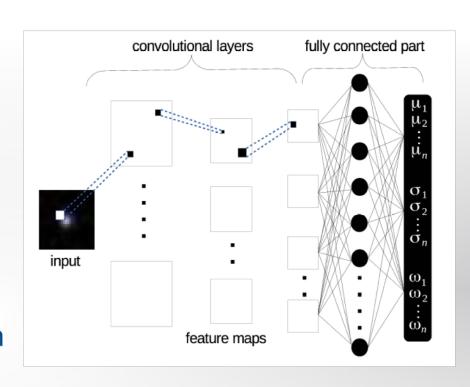


publish all:

- architecture / model
- hyper-parameters

deep-learning:

- weights / biases
- training-data / data augmentation



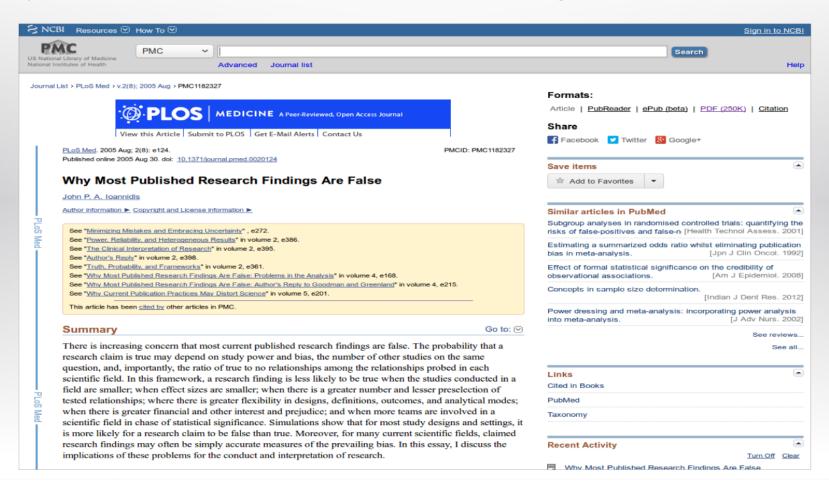
online-learning / streaming algorithms:

how to preserve this?

Statistical Methods



John P. A. Ioannidis: "Why Most Published Research Findings Are False"



Statistical Methods

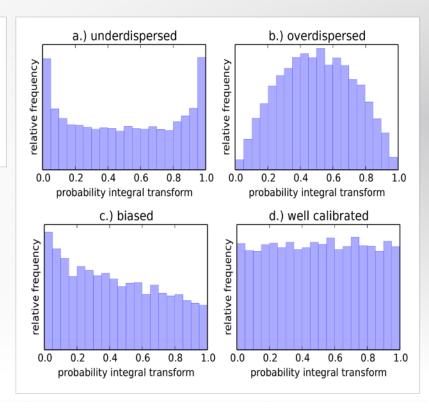


proper scores / scoring rules dealing with uncertainties

continuous ranked probability score

$$CRPS = \frac{1}{N} \sum_{t=1}^{N} crps(CDF_t, z_t),$$
 with $crps(CDF_t, z_t) = \int_{-\infty}^{+\infty} \left[CDF_t(z) - CDF_{z_t}(z)\right]^2 dz$

probability integral transform
 preserve data for different kinds of analysis!



Conclusion



publications

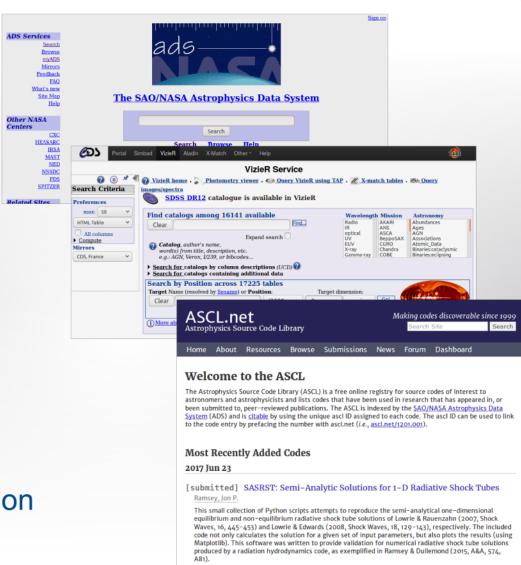
open access

data

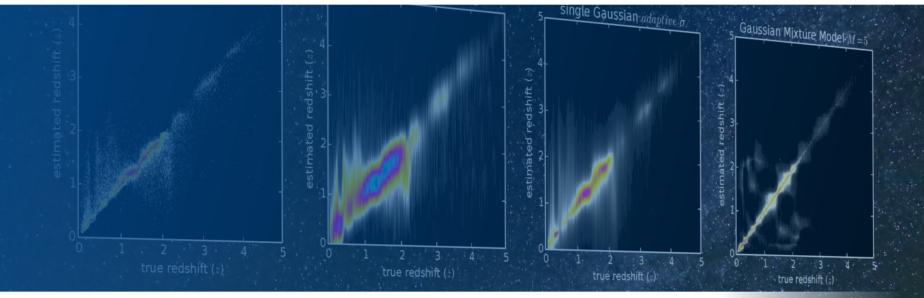
- raw-data / catalogs
- training- / test-data
- reference data-sets
- detailed results

software

- code / repositories
- parameter / configuration
- environment







thanks for your attention!



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