Photometric selection and redshifts for quasars **Kilo-Degree Survey Data Release 4**

Szymon Nakoneczny - 17.10.2021

The catalogs

http://kids.strw.leidenuniv.nl/DR3/ quasarcatalog.php

arXiv.org > astro-ph > arXiv:1812.03084

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Astrophysics > Instrumentation and Methods for Astrophysics

[Submitted on 7 Dec 2018 (v1), last revised 9 Apr 2019 (this version, v2)]

Catalog of quasars from the Kilo-Degree Survey Data Release 3

S. Nakoneczny, M. Bilicki, A. Solarz, A. Pollo, N. Maddox, C. Spiniello, M. Brescia, N.R. Napolitano

> Changes to the DR4 release: 1. Added photo-zs and near-IR 2. Increased the magnitude coverage to r < 23.5

http://kids.strw.leidenuniv.nl/DR4/ quasarcatalog.php

arXiv.org > astro-ph > arXiv:2010.13857

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[Submitted on 26 Oct 2020]

Photometric selection and redshifts for quasars in the Kilo-Degree Survey Data Release 4

S.J. Nakoneczny, M. Bilicki, A. Pollo, M. Asgari, A. Dvornik, T. Erben, B. Giblin, C. Heymans, H. Hildebrandt, A. Kannawadi, K. Kuijken, N.R. Napolitano, E. Valentijn





Agenda

1. Introduction

1. Quasars in photometry

2. Methodology

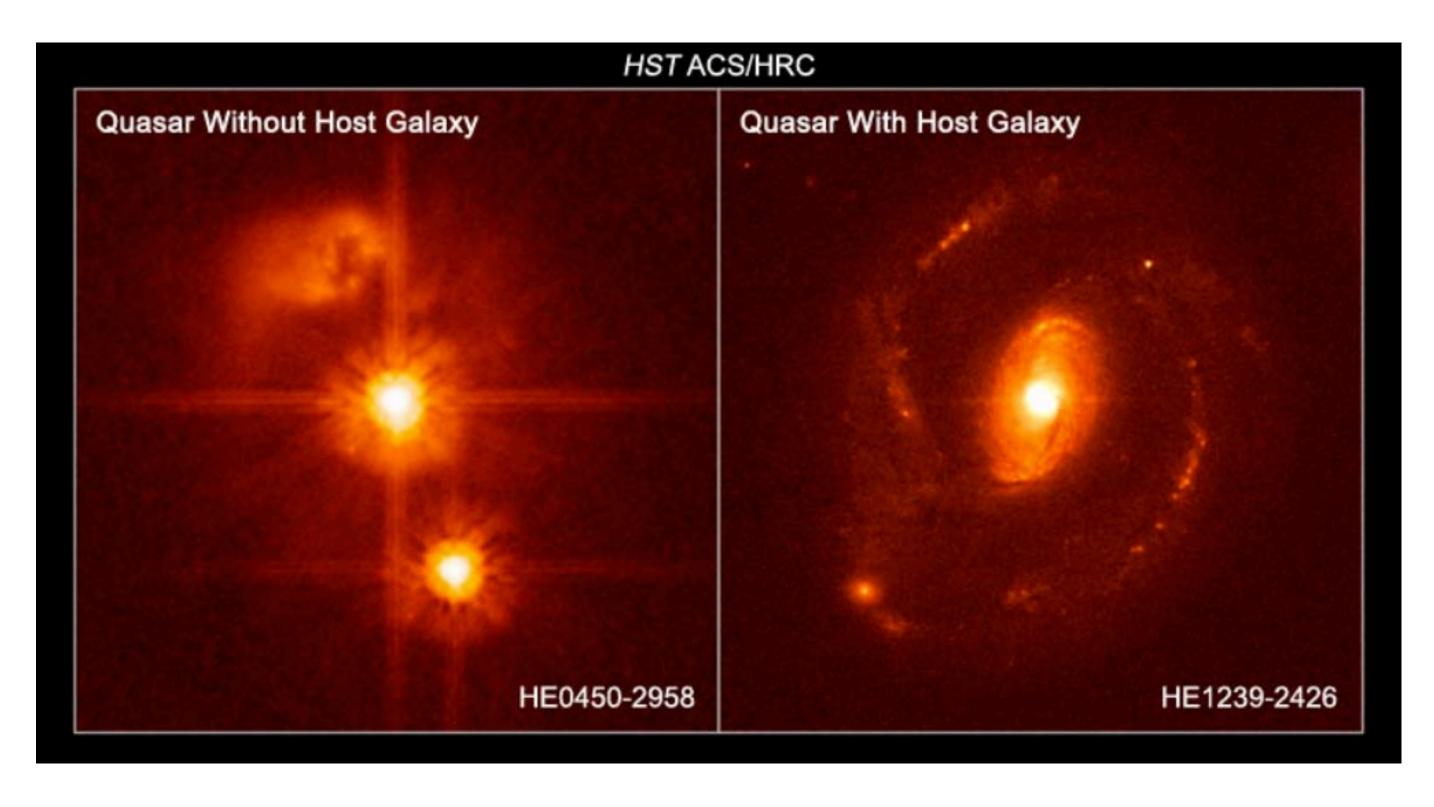
- 1. Data (research and visualisation)
- 2. Machine learning (validation)

3. Results

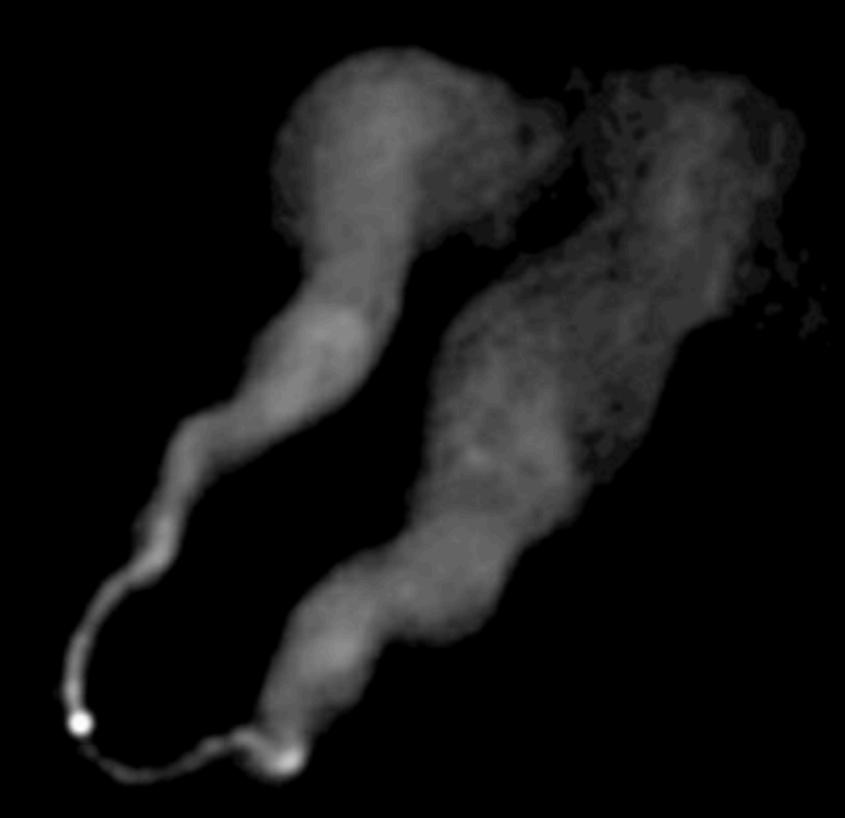
- 1. Experiments
- 2. Final catalog (testing)

Introduction

Left: quasar (in the center) and a star



Right: quasar and a spiral host galaxy



Active galactic nucleus (AGN) with jets

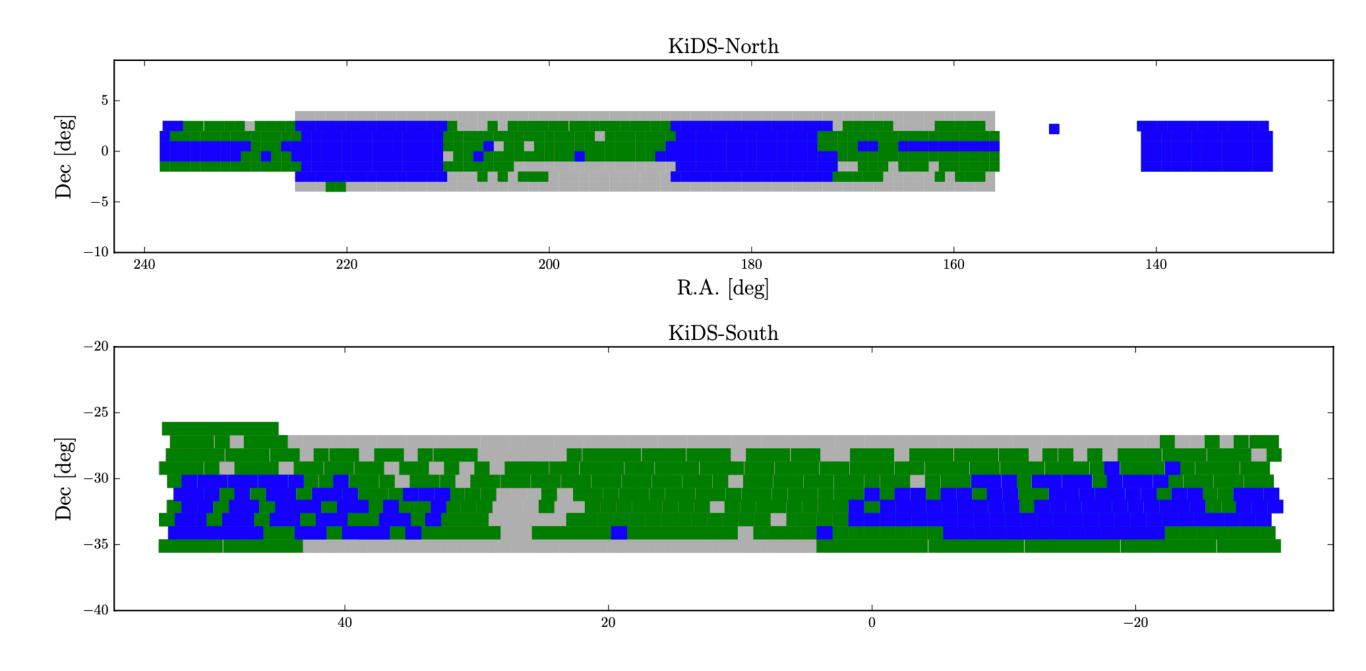




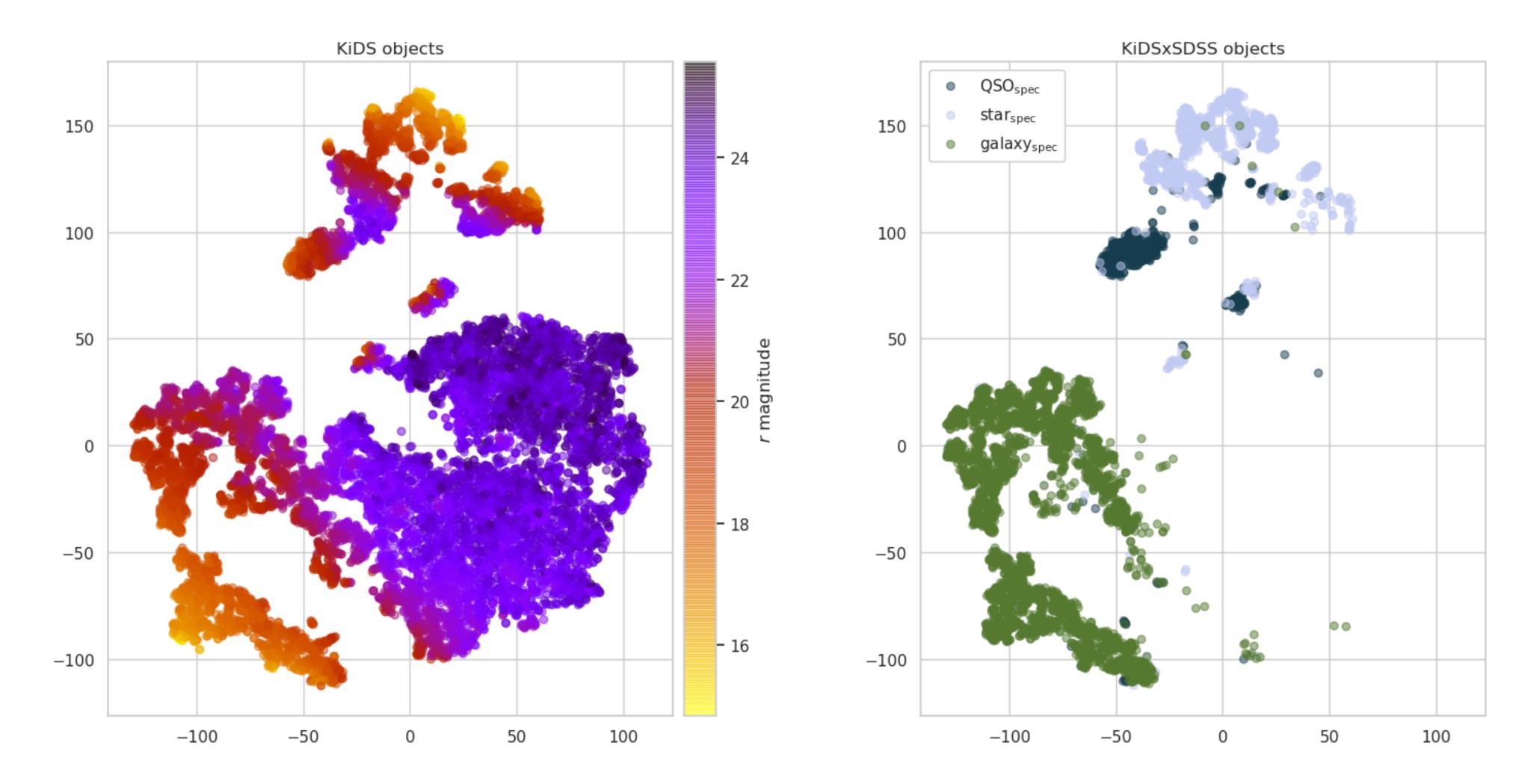
Methodology

Data

- Kilo-Degree Survey (KiDS): optical ugri \bullet
- VISTA Kilo-degree Infrared Galaxy Survey (VIKING): near-infrared ZYJHKs \bullet
- Depth: *r* < 25
- Data release 4: 1006 square-degrees (1350 in the final DR5), ~100M objects
- 9 band detections yield 45M objects

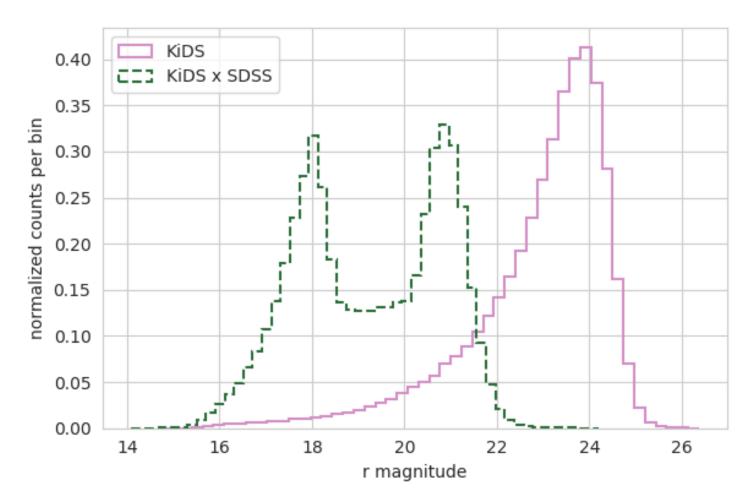


Extrapolation problem

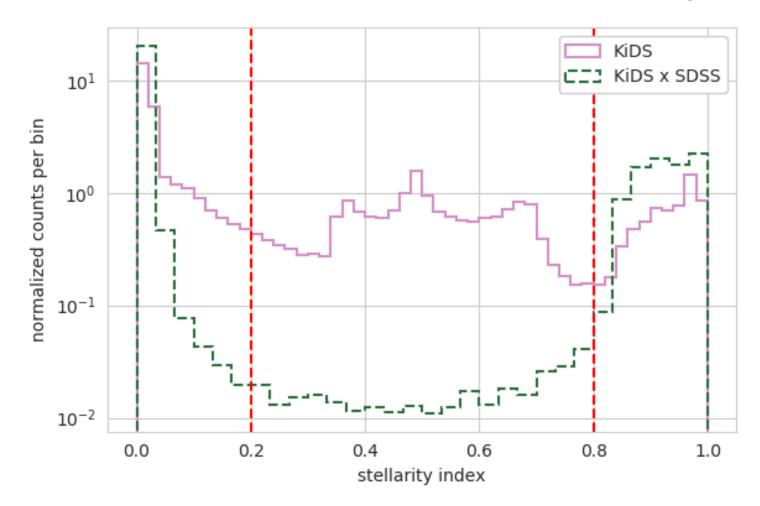


t-SNE projection of **KiDSxSDSS data**. Left: r magnitude, right: SDSS spectroscopic classification.

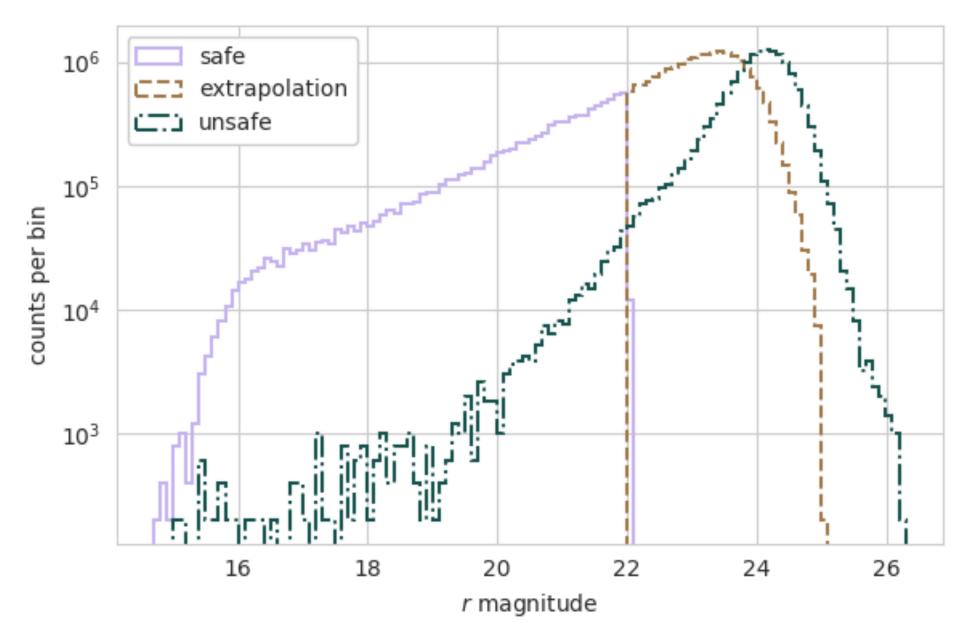
Inference subsets



Normalized histograms of the r magnitude for the KiDS inference data and the KiDS \times SDSS training set.

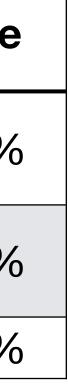


Normalized histograms of Sextractor's stellarity index for training and inference datasets.

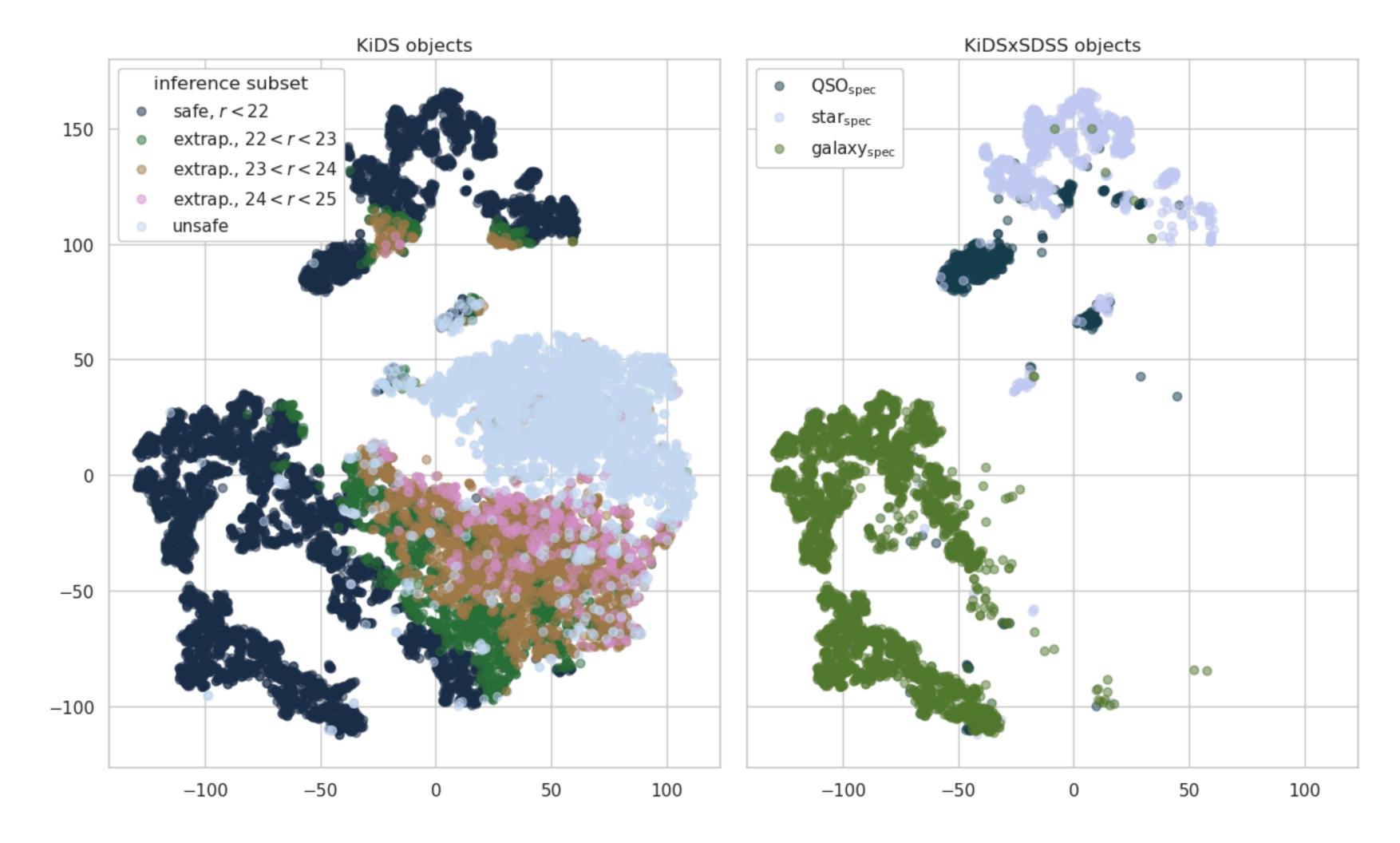


Distribution of inference subsets over the *r* magnitude.

subset name	r magnitude	stellarity index	size
safe	r < 22	not in (0.2, 0.8)	21%
extrapolation	22 < r < 25	not in (0.2, 0.8)	45%
unsafe	r > 25	in (0.2, 0.8)	34%



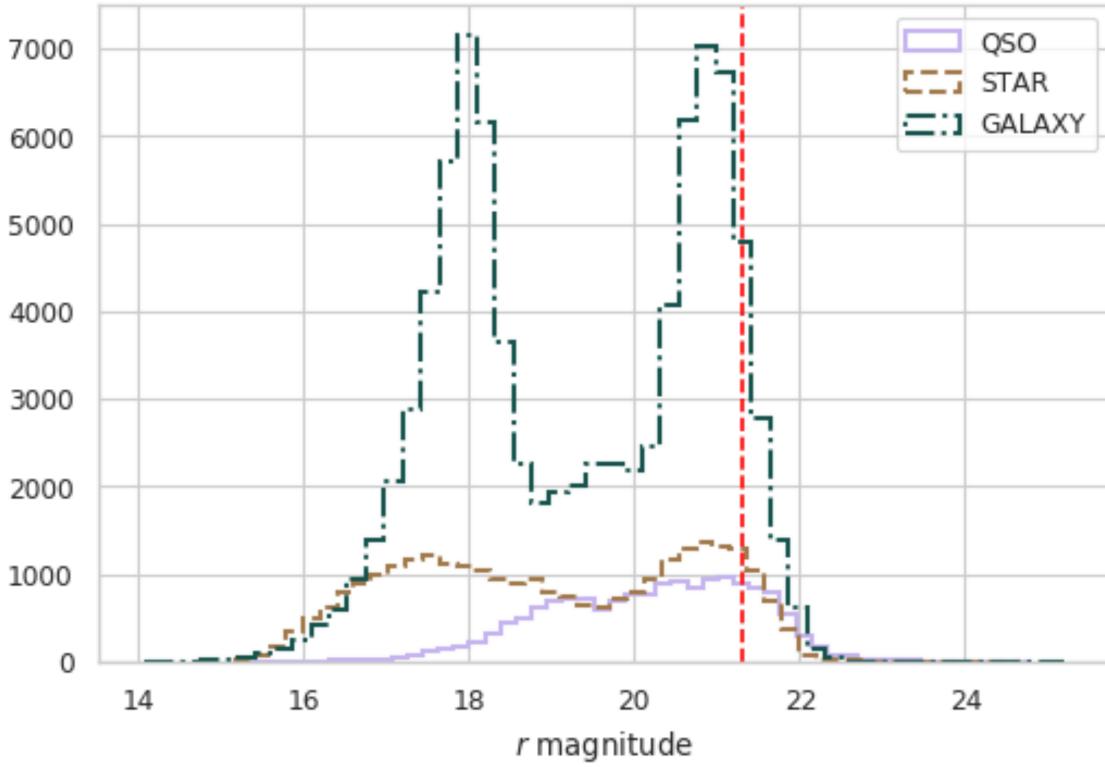
Inference subsets



t-SNE projection of KiDSxSDSS data. Left: inference subsets, right: SDSS spectroscopic classification.

Testing Using internal data structure

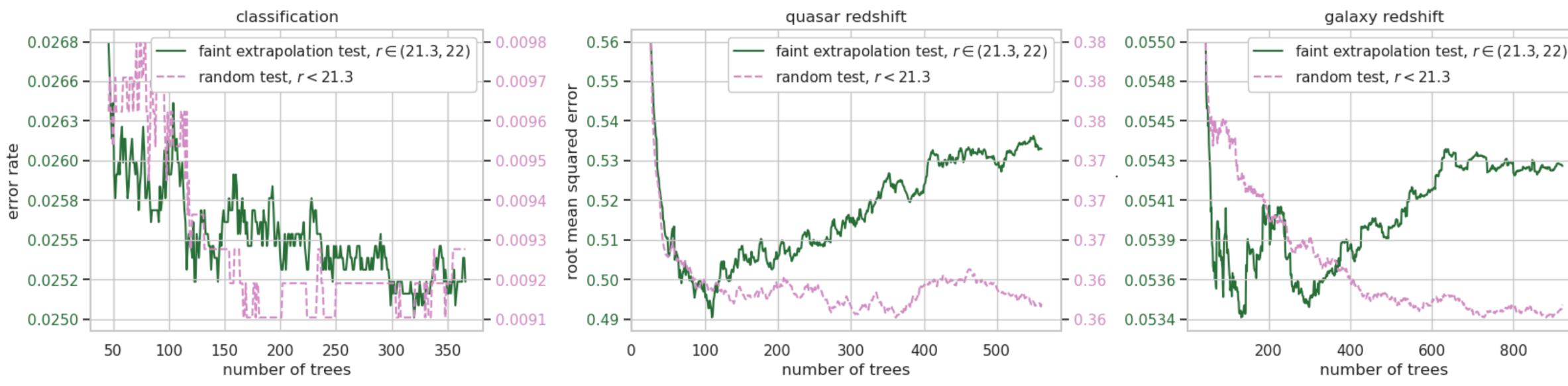
- Model training is based on random test
- Parameters tuning is based on random and faint extrapolation tests
- New models are trained for inference with parameters derived from experiments (except number of epochs/trees)



Distribution of spectroscopic classes in the training dataset. The red dashed line at r = 21.3 separates data used for the faint extrapolation test.

Experiment results

Testing Method comparison



Learning histories for XGBoost. Left: classification, center: QSO redshift, right: galaxy redshift.

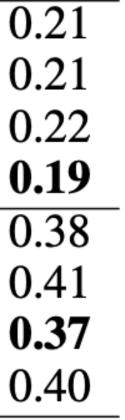
- 0.0297 - 0.0296 - 0.0295 - 0.0294 - 0.0293 - 0.0292 - 0.0291 - 0.0290

Model selection Experimental comparison

		classificati	ion		redshif	ft in QS	O _{spec}	redshif	ft in QS	O _{photo}
test	model	accuracy	purity	recall	MSE	R2	z error	MSE	R2	z error
random	RF	99.00%	97.44%	94.31%	0.12	85%	0.018 ± 0.14	0.12	84%	0.032 ± 0
	XGB	99.09%	97.85%	94.75%	0.13	84%	0.017 ± 0.15	0.13	83%	0.030 ± 0
	ANN	98.98%	96.93%	94.67%	0.10	88%	$\textbf{0.009} \pm \textbf{0.12}$	0.11	85%	0.023 ± 0
	clf XGB, z ANN	99.09%	97.85%	94.75%	0.10	88 %	$\textbf{0.009} \pm \textbf{0.12}$	0.10	87 %	0.020 ± 0
faint	RF	97.44%	96.12%	92.37%	0.31	31%	0.019 ± 0.25	0.33	31%	0.046 ± 0
extrap.	XGB	97.44%	96.48%	92.12%	0.27	39%	0.036 ± 0.23	0.34	29%	0.077 ± 0
	ANN	97.27%	96.52%	90.89%	0.22	51%	-0.0004 ± 0.19	0.28	39%	0.042 ± 0
	clf XGB, z ANN	97.44%	96.48%	92.12%	0.22	51%	-0.0004 ± 0.19	0.31	35%	0.050 ± 0

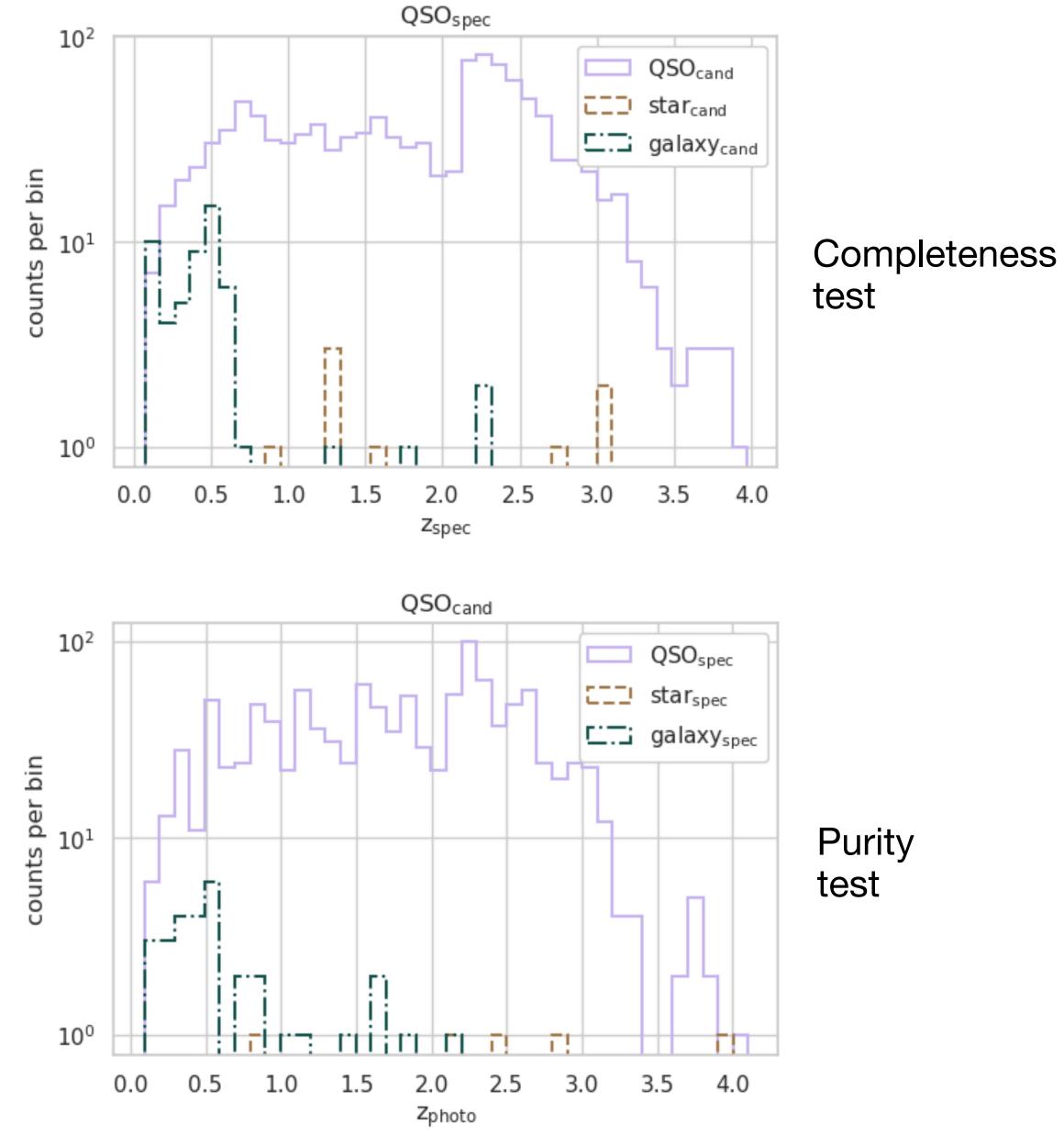
Final method: classification network + QSO redshift networ

+ QSO redshift network with Gaussian uncertainty



QSO classification

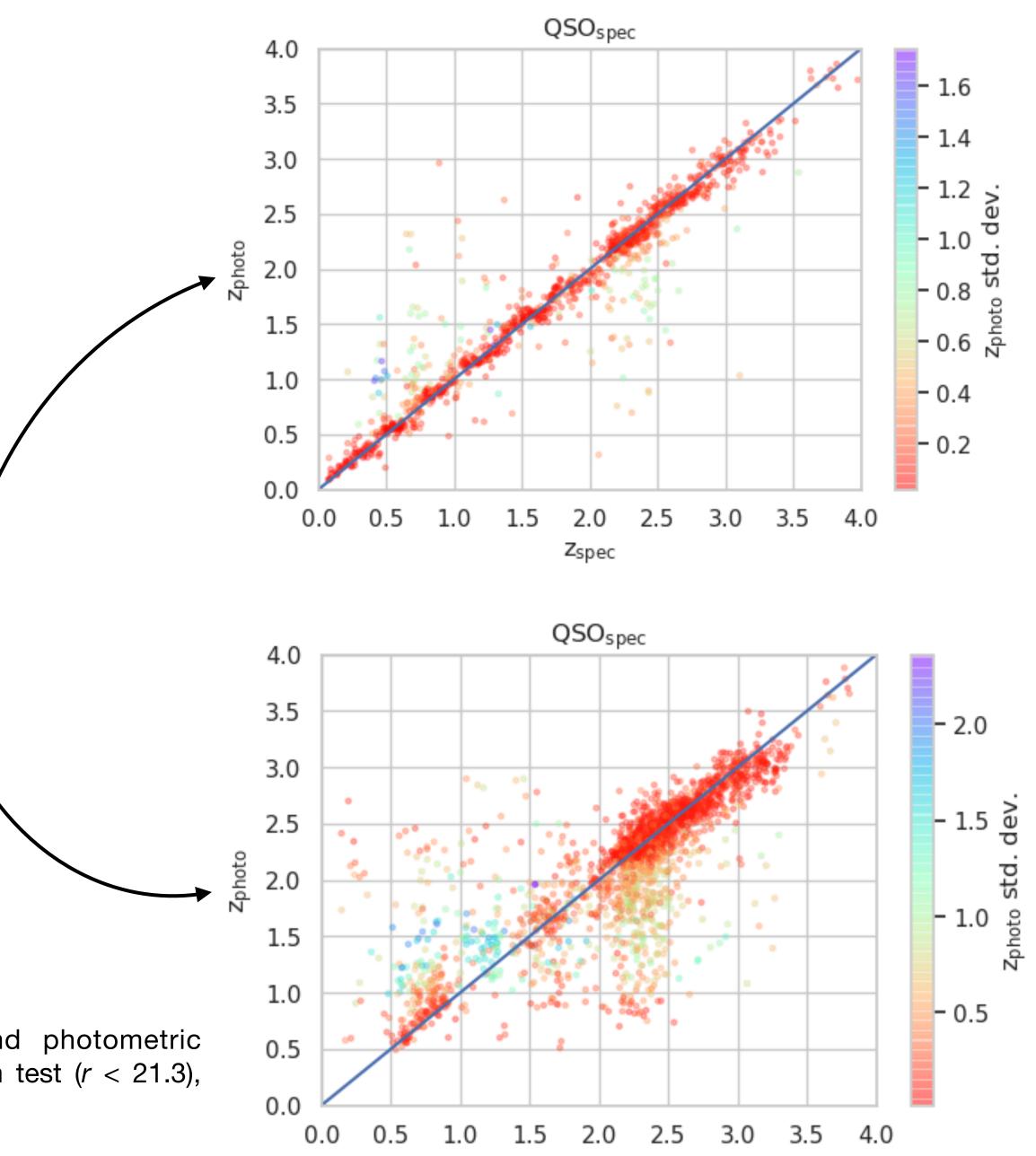
	3-class accuracy	purity	completeness
randomly selected (r < 21.3)	99.0%	96.9%	94.7%
faint extrapolation (21.3 < r < 22)	97.3%	96.5%	90.9%



QSO photo-zs

	$\delta z = \frac{(z_{photo} - z_{spec})}{(1 + z_{spec})}$	
randomly selected (r < 21.3)	0.009 +/- 0.12	
faint extrapolation (21.3 < r < 22)	-0.0004 +/- 0.19	

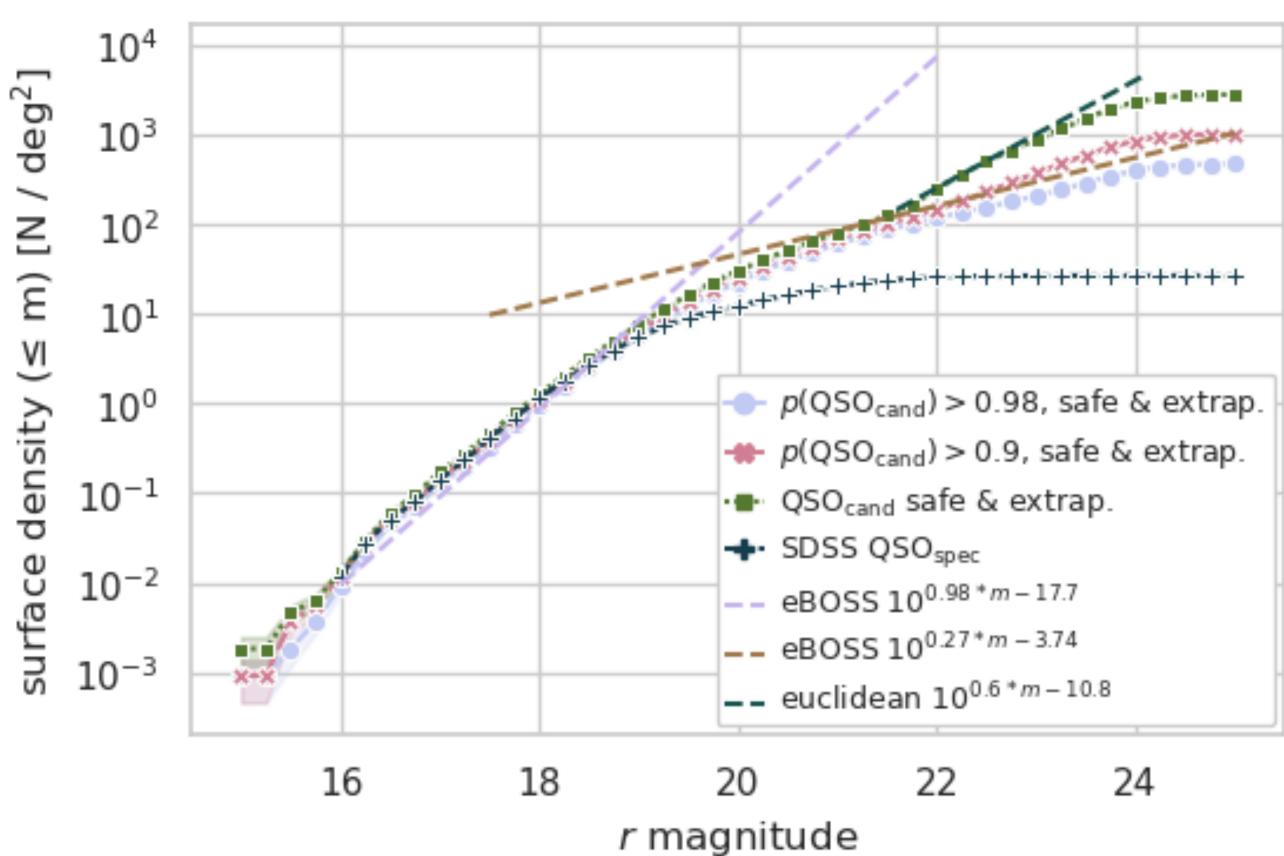
Comparison of the spectroscopic and photometric redshifts on SDSS quasars. *Left:* random test (r < 21.3), *right:* faint extrapolation test (21.3 < r < 22)



Zspec

Final catalog properties

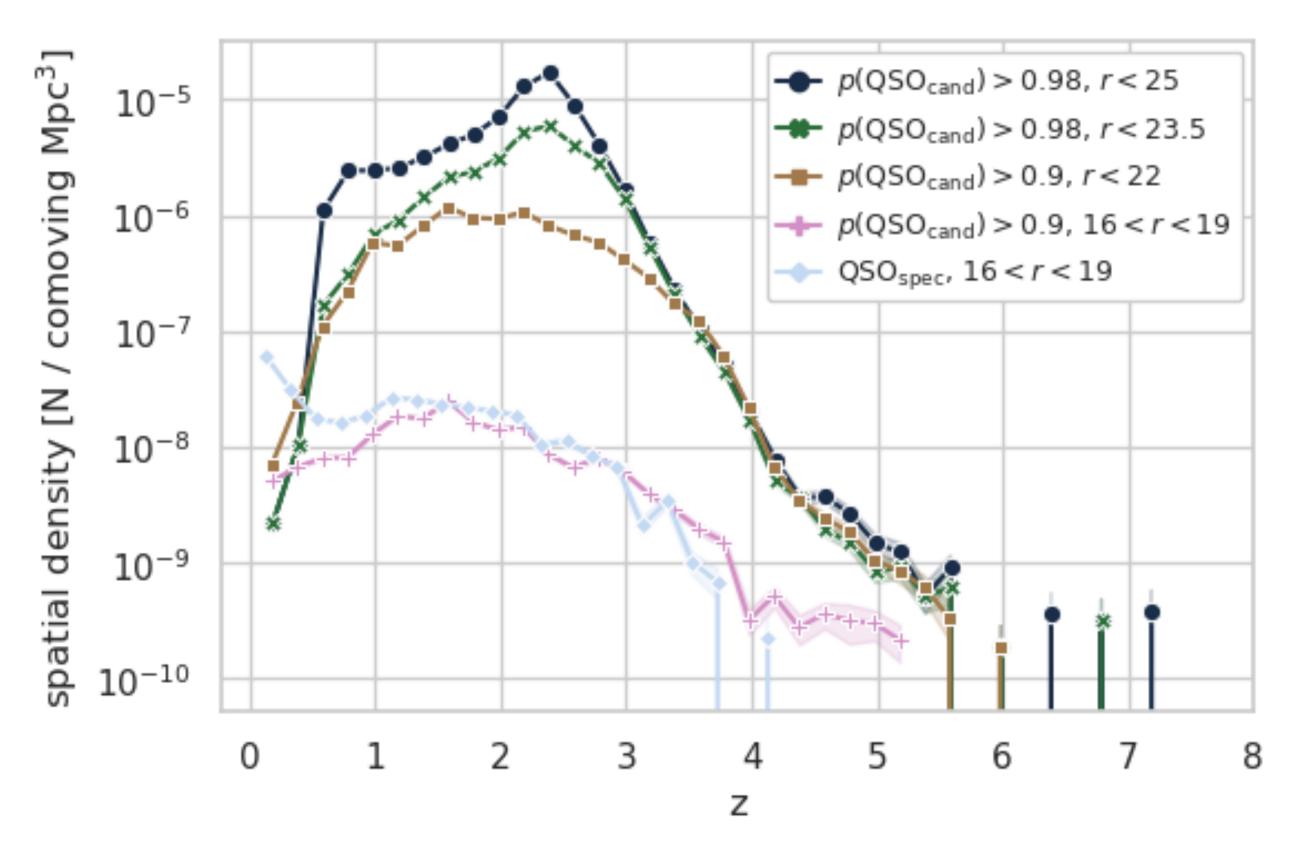
Number counts **Completeness of the catalog**



KiDS photometric QSO catalog at progressing probability cuts

- Straight dashed lines: quasar model from eBOSS
- Expected number counts lacksquareat $p(QSO_{cand}) > 0.98$
- Limited by the extrapolation ulletsubset which stops growing at *r* > 23.5

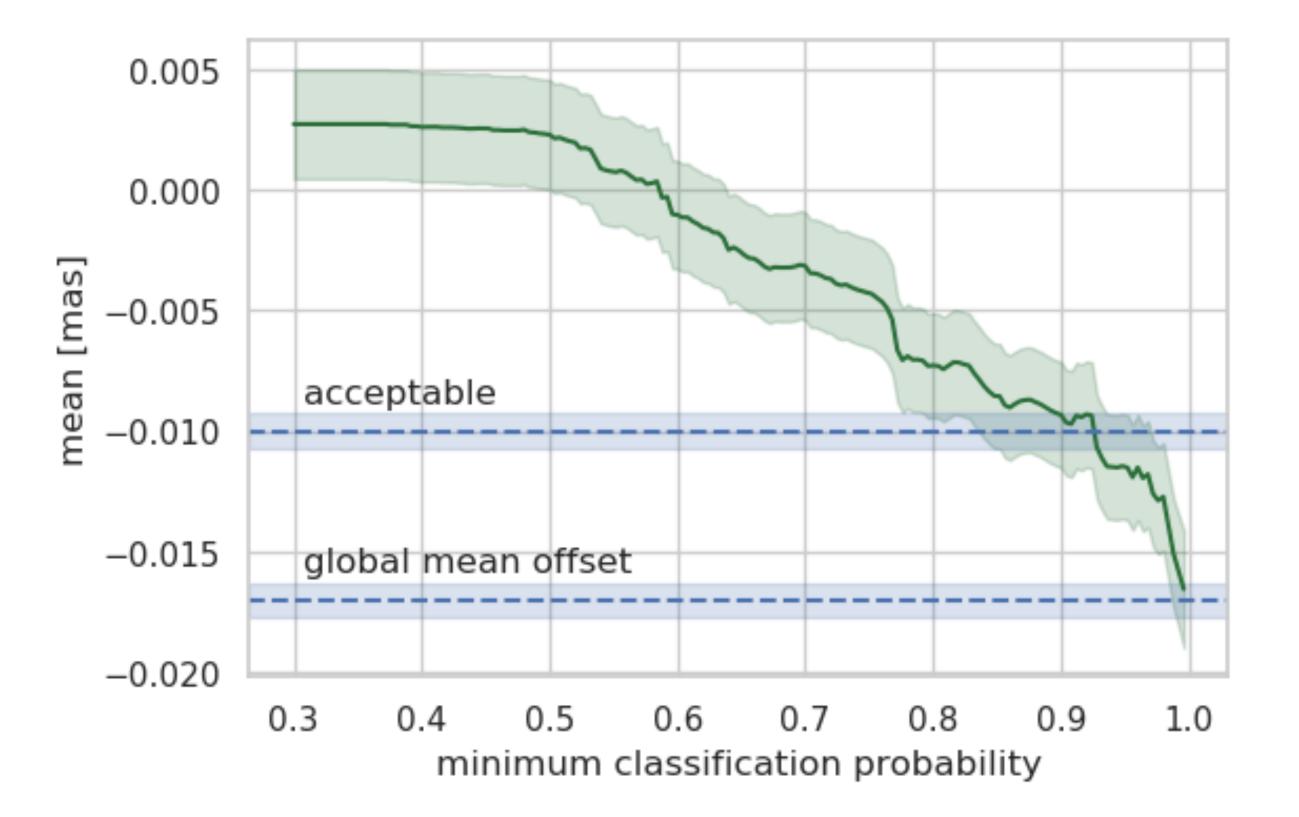
Spatial number density Redshift reliability



Spatial number density for KiDS photometric QSO at progressing magnitudes and corresponding probability cuts.

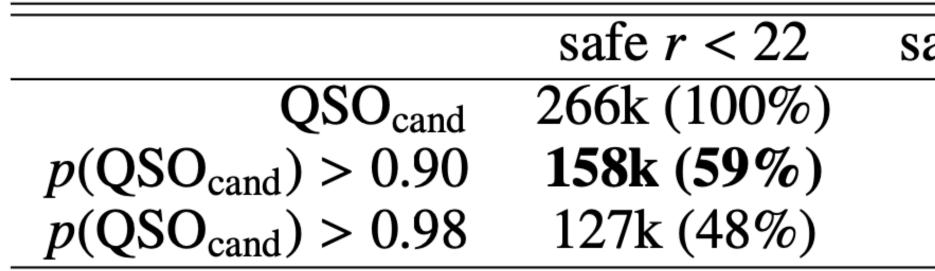
- Good result at the SDSS completeness limit r < 19
- Expected peek at z > 2 for p(QSO_{cand}) > 0.98 and r < 23.5
- Contamination with low-z galaxies at r > 23.5

Parallax test Purity of the catalog



Mean parallax value for photometric quasars as a function of minimum classification probability.

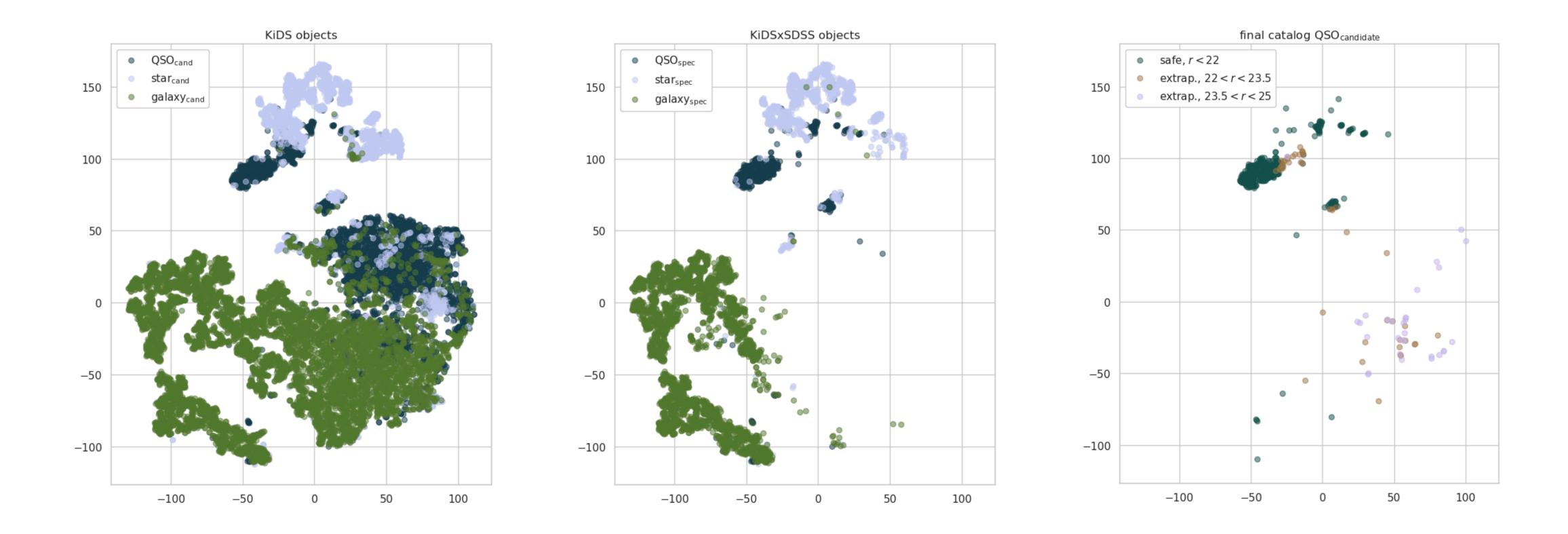
The final catalog Suggested probability cuts



Suggested probability cuts at progressing magnitudes

afe & extrap. <i>r</i> < 23.5	safe & extrap. $r < 25$
1.6M (100%)	3M (100%)
637k (39%)	1.1M (36%)
311k (19%)	507k (17%)

Classification Visualisation



t-SNE projection of KiDS + KiDSxSDSS data. *Left*: photometric classification, *center*: SDSS spectroscopic classification, *right*: final photometric QSO catalog.

Applications

Catalog ready for:

- tomographic study of LSS
- AGN studies (e.g. SED fitting, possible test on photo-zs)

Main takeaways

Key points to successful ML application:

- 1. Data research and visualisations t-SNE visualisations, feature understanding, inference subsets
- 2. Tough model validation method Faint extrapolation test to fit bias vs variance trade-off
- 3. Physically-driven testing approach Number counts and GAIA parallaxes to fit purity vs completeness

Thank you!