5[^] EUCLID ITALIAN MEETING

Giulia Rodighiero (*University of Padova*)

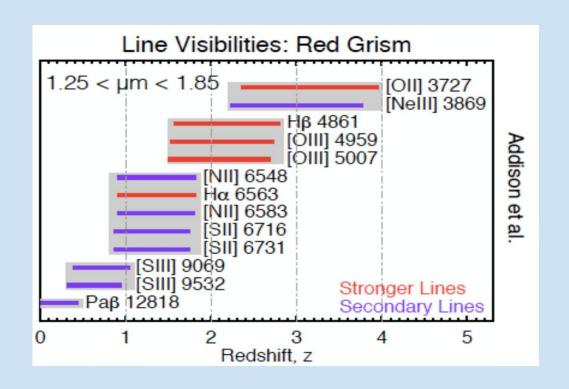
for

GAEV-SWG WP10 – high-z galaxies

24th of February 2022

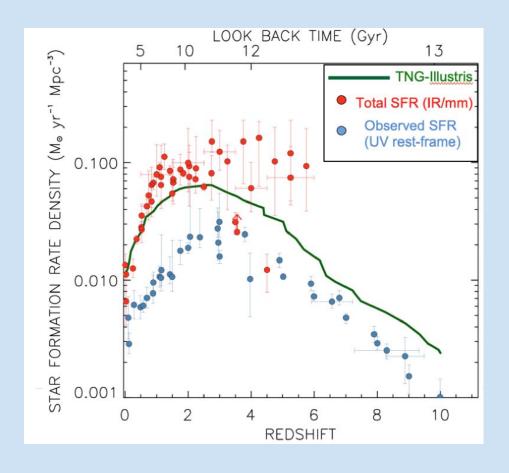
WP10 identity:

- start of activities: December 2020
- ~30 participants
- four telecons
- high-z galaxies ⇒ above z~4 and below z~7!!



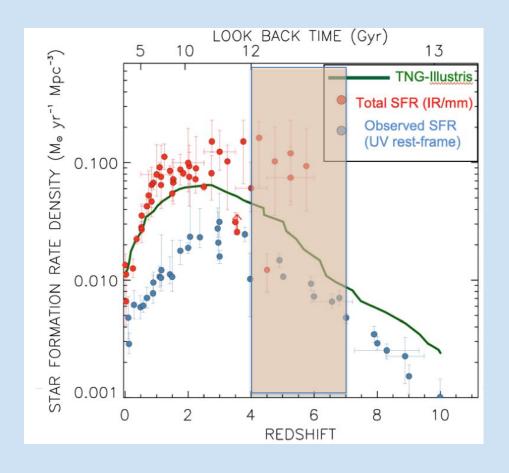
Goal of this Euclid SWG Galaxy Evolution Work Package:

To understand the potential contribution of Euclid in revealing a class of sources that are likely to represent the bulk population of massive galaxies that have been missed from previous surveys and are probably the progenitors of the largest present-day galaxies in massive groups and clusters.

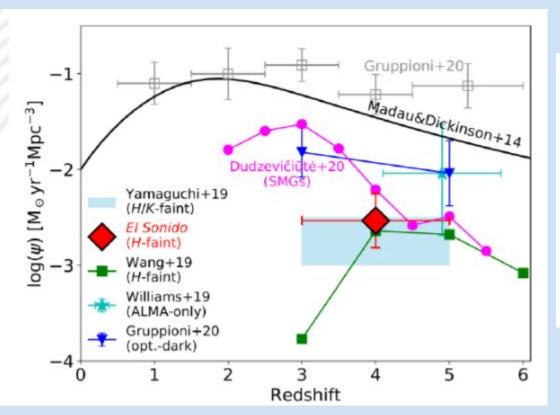


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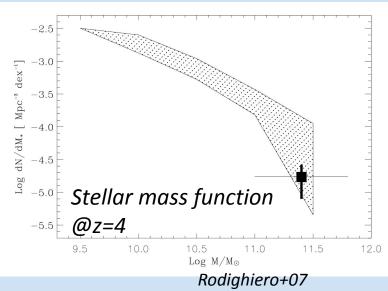
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Contribution of "HST" dark sources to the stellar mass density (selection from IRAC, ALMA, radio....):



Different population from LBGs!



Sun+20 (but see also Talia+20, Enia+22)

These dusty and massive galaxies show remarkable star formation activity but are very rare and faint ⇒ Need for Deep and Wide near-IR surveys to statistically recover this population -> Euclid Deep Survey

High-z populations in the Euclid redshift desert (always thought in combination to ancillary obs):

- Photometric redshifts (including Machine Learning)

- Colours

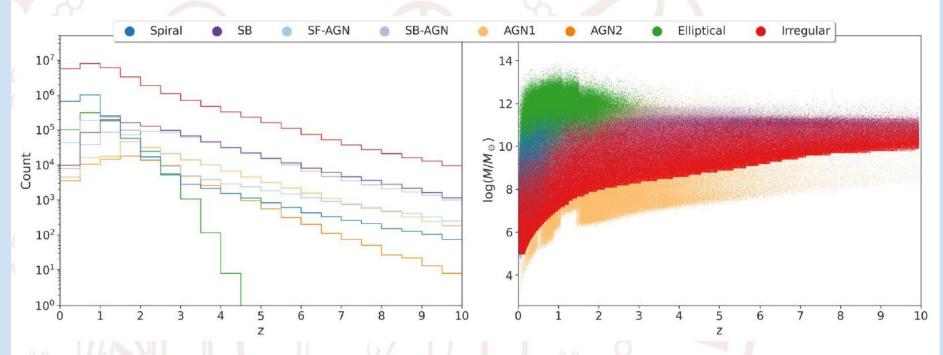
- Drop-outs

- Line emitters embedded in broad-band photometry



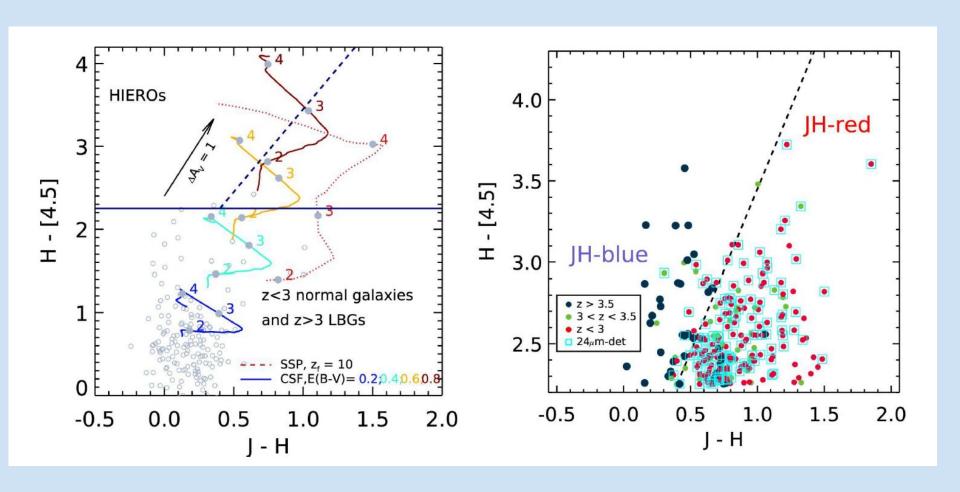
The Euclid Deep Fields Simulated Catalog

The Euclid Deep Survey combination of depth and area results in a simulated catalog with a total of more than 30 million objects with redshift from z ≈ 0 to ≈ 10



- ⇒ SPRITZ simulation (Bisigello et al. 2021)
- •The simulation is built from the Herschel infrared luminosity functions of different galaxy populations, and is based on a wide set of empirical relations to associate a spectral energy distribution and physical properties to each simulated galaxy.

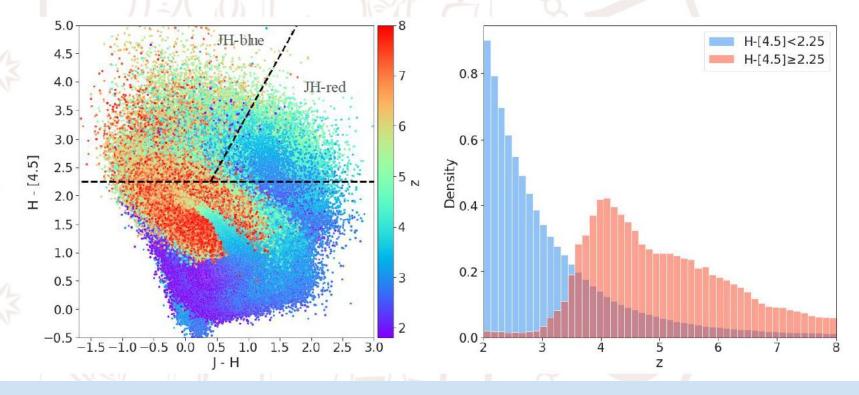
First assessment on HIEROS (Wang+16)



Photometric Selections

MOSTLY BASED ON THE MASTER THESIS OF THEO SIGNOR IN PHYSICS OF DATA (UniPD)

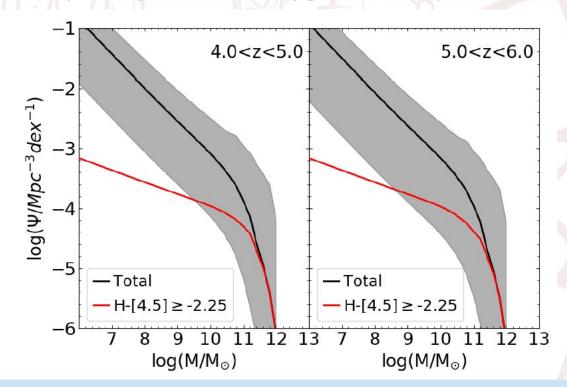
- First, we check the simulated catalog compatibility with a set of observed photometric diagnostics available from the literature (Laigle et al. 2016; Daddi et al. 2004; Wang et al. 2016; van Mierlo et al. 2022, in prep.)
- In particular, we check the **distributions of magnitudes**, **SED types** and **redshifts**, as a function of different color-color plots.
- HIEROs (extremely red objects; old or dusty galaxies at z>3) color selection: H-[4.5]>2.25





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courtesy Laura Bisigello

Photo-z - Data

Gradient Boosted Trees (XGBoost) are implemented to predict the redshift of galaxies within the Euclid Deep survey simulated catalog, based on multi-band photometry.

The Dataset consists of

- Fluxes in 11 bands: VIS, NISP/Y, NISP/J, NISP/H, Rubin/u, Rubin/g, Rubin/r, Rubin/i, Rubin/z,IRAC/3.6μm, IRAC/4.5μm bands;
- Redshift z
- SED Type

Band	5σ Depth	2σ Depth		
VIS	28.2	29.2		
NISP/Y	26.3	27.3		
NISP/J	26.5	27.5		
NISP/H	26.4	27.4		
Rubin/u	25.15	26.1		
Rubin/g	26.35	27.34		
Rubin/r	26.45	27.4		
Rubin/i	25.75	26.7		
Rubin/z	25	26		
IRAC/3.6μm	24.5	25.5		
$IRAC/4.5\mu m$	24.5	25.5		



Photo-z - Training Set Size

In real-world observations, one will have no choice regarding the size of the training set. However, when forecasting future surveys observations, it is useful to assess what **dimension of the training set** is required to obtain a certain **redshift prediction performance**.

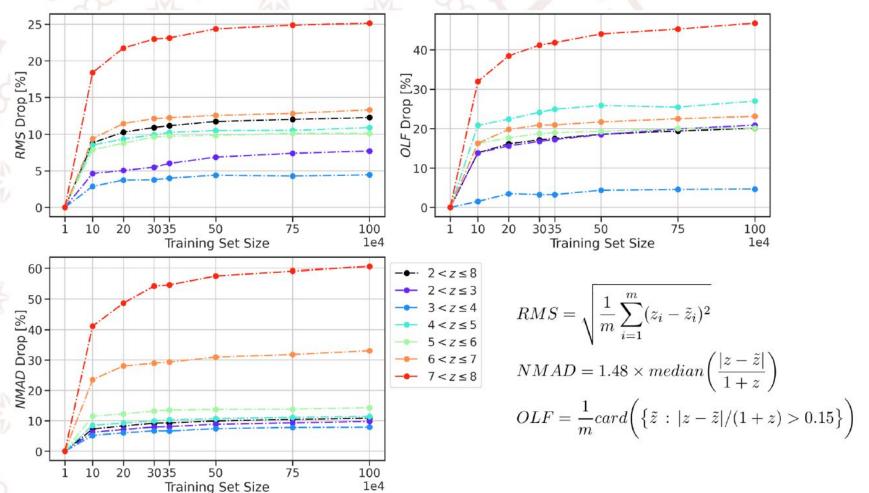
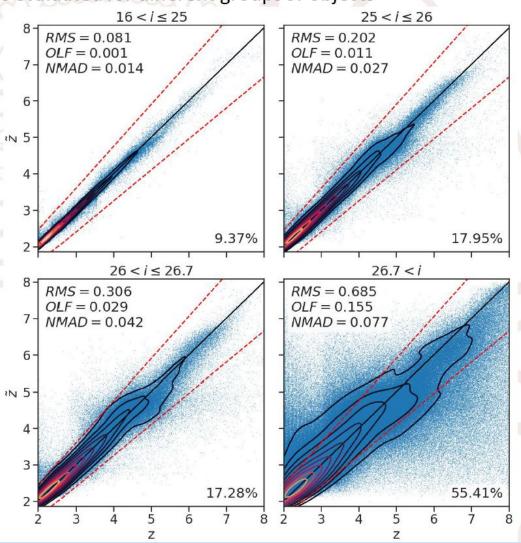




Photo-z - Results

Following a Bayesian optimization for the xgboost hyperparameters, the test set performances are evaluated for different groups of objects





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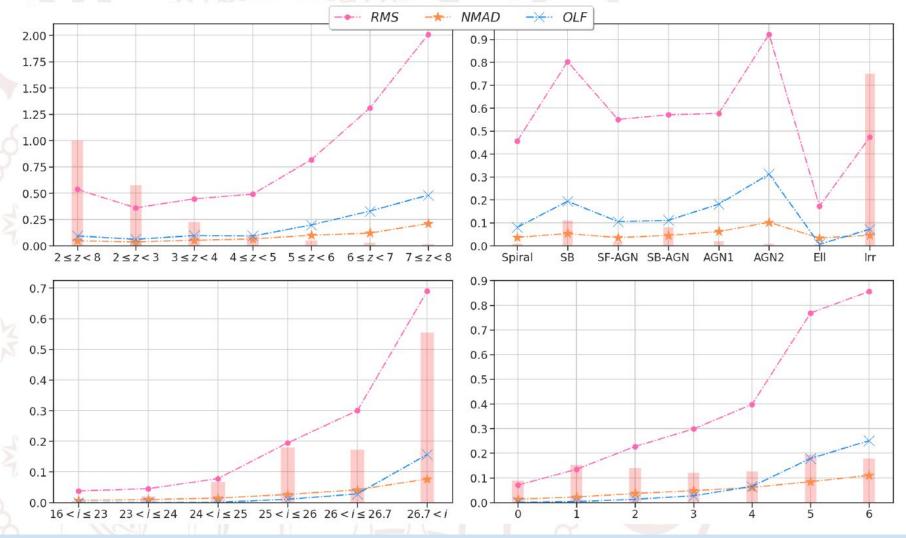




Photo-z - Results

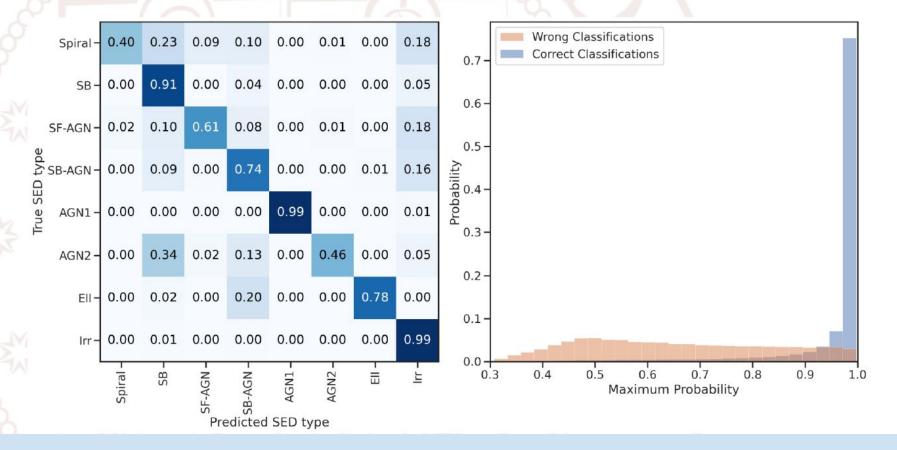
- To give a contest to the results obtained, they are compared to previous photometric redshift performances, in a similar z-range.
- This comparsion is made with the results reported in Laigle 2016, where the precision of photometric redshifts obtained via the SED fitting technique using 28 bands over the COSMOS2015 catalog is assessed againts spectroscopic ones
- Clearly the performances are very comparable

	i-band Magnitude	NMAD		OLF			
		COSMOS Star-Forming	COSMOS Quiescent	This Work	COSMOS Star-Forming	COSMOS Quiescent	This Work
	$16 < i \le 21$	0.007	0.005	0.005	0.005	0.0	0.0
	$21 < i \le 22$	0.008	0.007	0.006	0.006	0.003	0.0
	$22 < i \le 23$	0.01	0.01	0.007	0.017	0.006	0.0
	$23 < i \leq 24$	0.022	0.027	0.009	0.067	0.06	0.0
1	$24 < i \le 27$	5 //	0.054	0.048	-	0.189	0.096

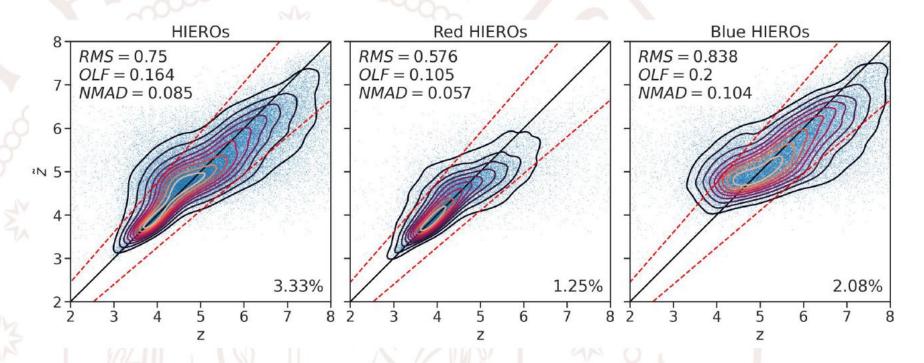


Spectral Type Classification - Results FULL SAMPLE

- A gradient boosting approach was also taken to determine the SED type from photometry.
- Test Set accuracy: 94.4%
- This is clearly a simplified description, given the limited number of SED templates considered in the simulation

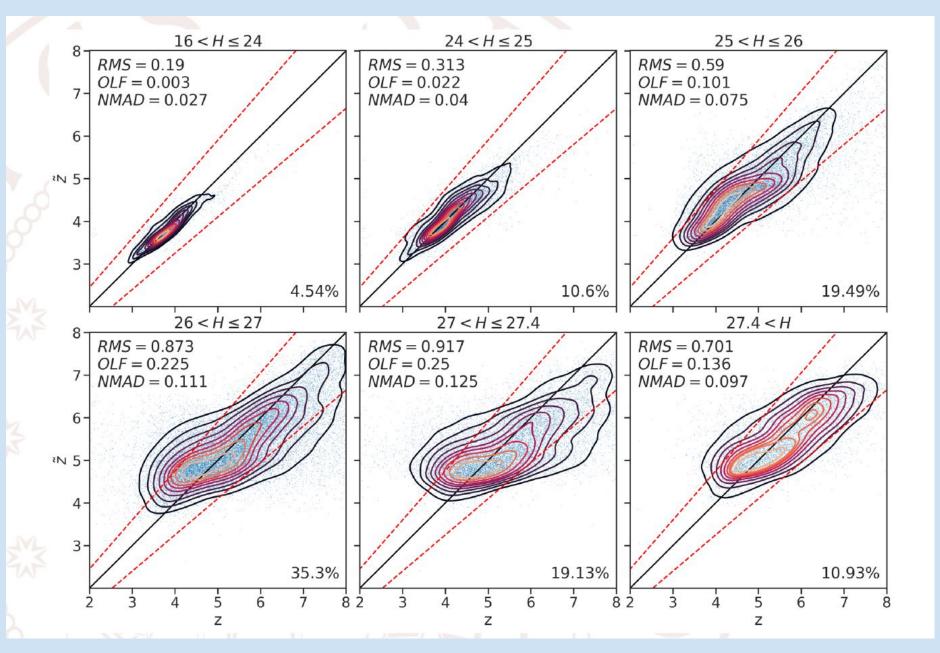


Particular focus in this work is on massive dusty galaxies, the **HIEROs** (H-[4.5]>2.25)



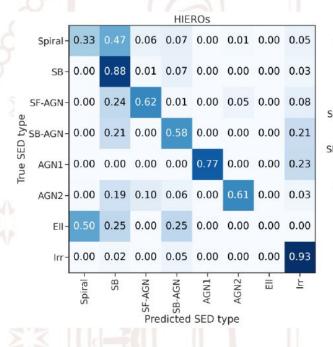
VS

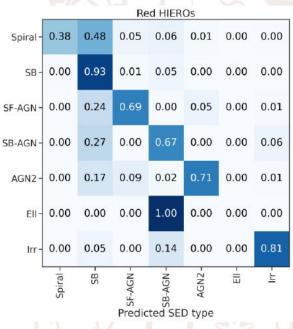
Wang et al. 2016 HIEROs photometric redshifts NMAD=0.11

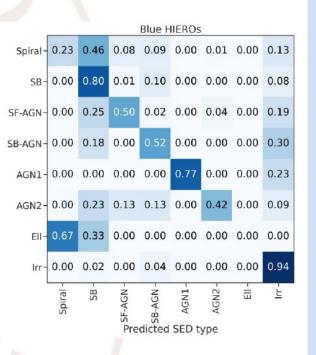


HIEROs Classification HIEROS SAMPLE

A gradient boosting approach was also taken to determine the SED type from photometry. In this case, the trained machine output, when a feature vector is fed to it, is a vector of probability (in this case with 8 entries, each one corresponding to a SED type). The predicted class is thus the one corresponding to the maximum value.

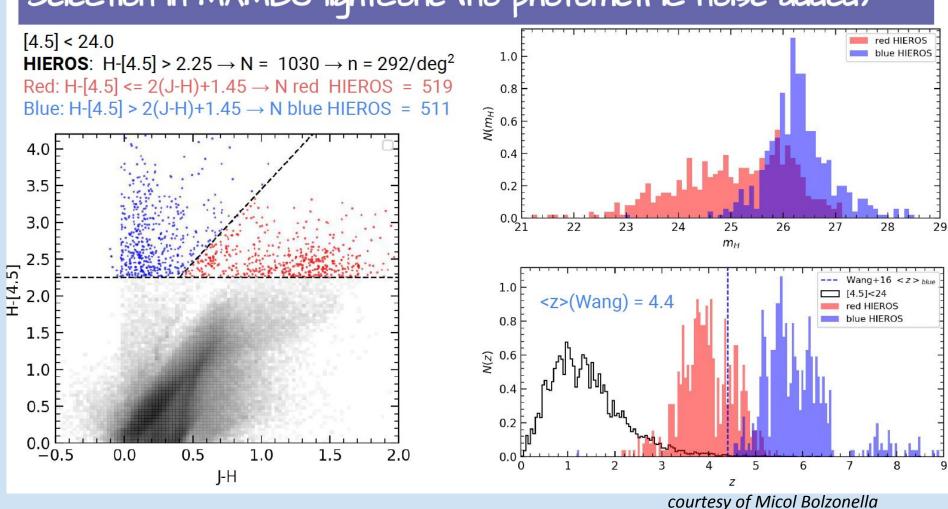




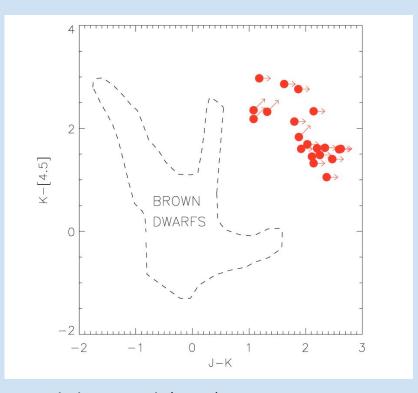


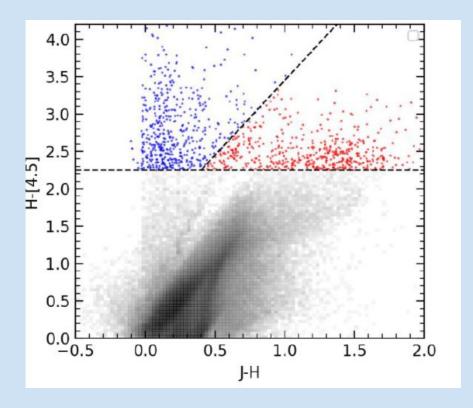
On going/future steps: analysis of different simulations and photo-z from the standard Euclid pipeline (OU-photoz)

Selection in MAMBO lightcone (no photometric noise added)



On going/future steps: Stellar contamination among the reddest drop-outs? check predictions from extensive Mock stellar catalogs (e.g. TRILEGAL, Girardi et al.)





Rodighiero et al. (2007)

Critical discussion with Primeval Universe / Cosmic Dawn team

- Taking advantage of people sitting in both Gal Evol (in particular this WP)
 and PU;
- Offline coordination with the WP leads in PU;
- Need to understand the depth of the multi-lambda photometric surveys available from Cosmic Dawn to address our science cases;
- Discussion about the PL-KPs from PU to see scientific and redshift overlaps with WP10 (and gal evol in general).