

# EXTRACTING PHYSICAL RULES FROM ENSEMBLE MACHINE LEARNING FOR THE SELECTION OF RADIO AGN

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# Results mostly from...

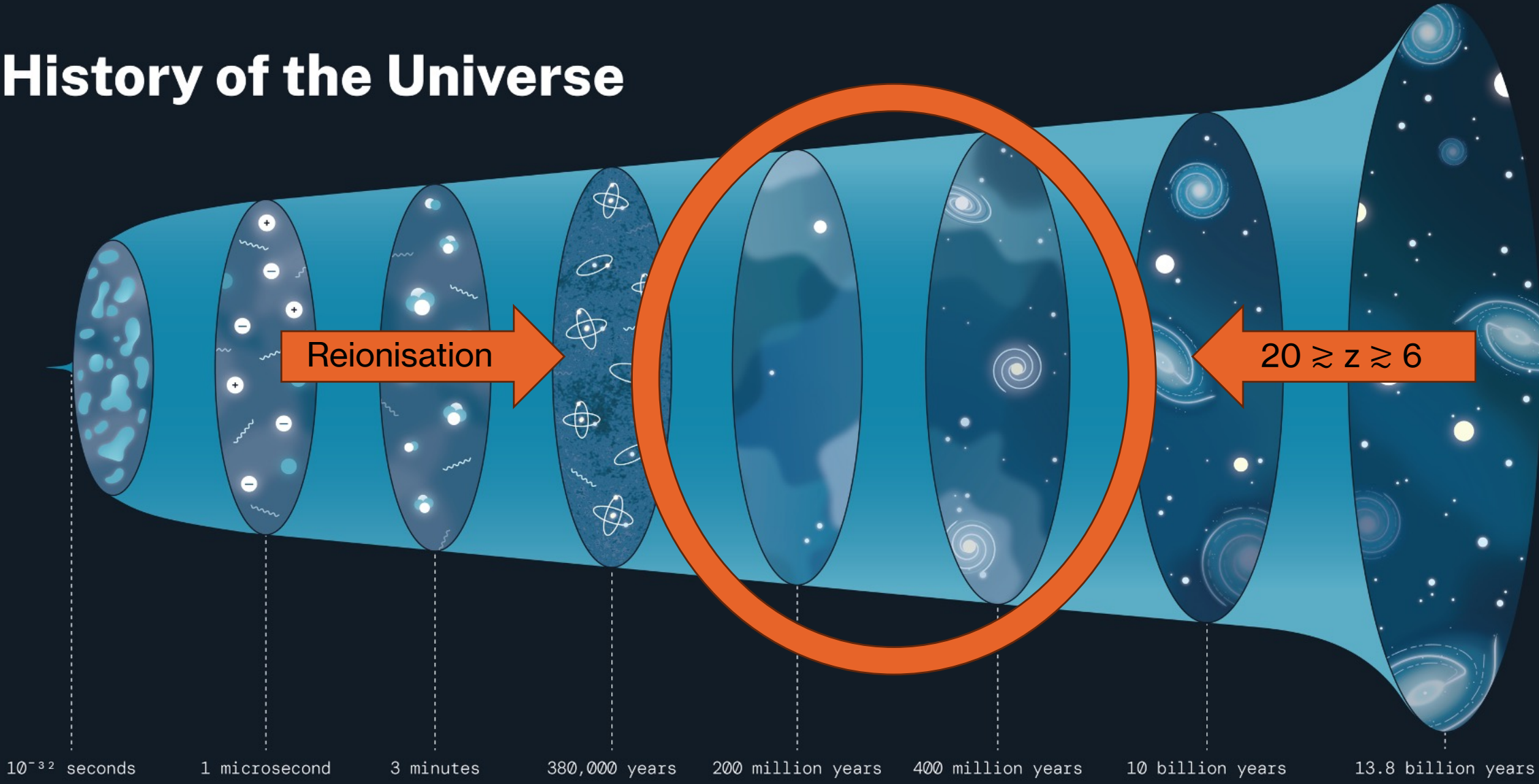
‘Selection of powerful radio galaxies with  
machine learning’

Carvajal et al. 2023 (A&A)

arXiv:2309.11652

# History of the Universe

Credit: NASA 2022

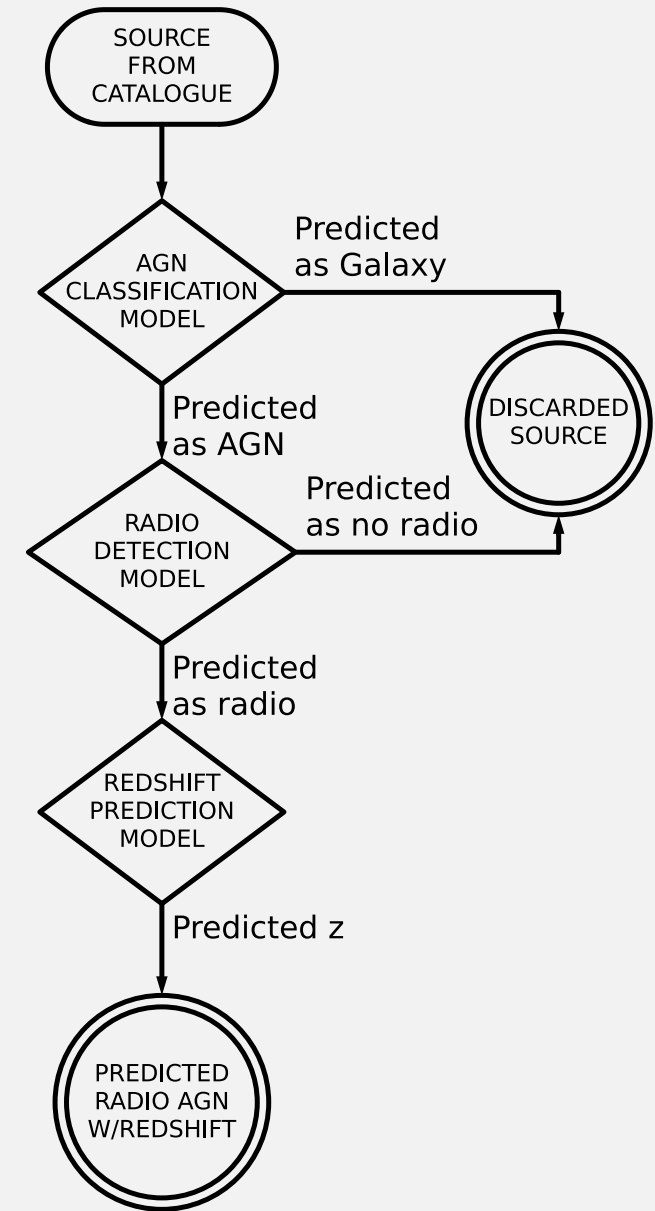


$10^{-32}$  seconds    1 microsecond    3 minutes    380,000 years    200 million years    400 million years    10 billion years    13.8 billion years

<b>Inflation</b>	<b>First Particles</b>	<b>First Nuclei</b>	<b>First Light</b>	<b>First Stars</b>	<b>Galaxies &amp; Dark Matter</b>	<b>Dark Energy</b>	<b>Today</b>
Initial expansion	Neutrons, protons, and electrons form	Helium and hydrogen form	The first atoms form	Gas and dust condense into stars	Galaxies form in dark matter cradles	Expansion accelerates	Humans observe the universe

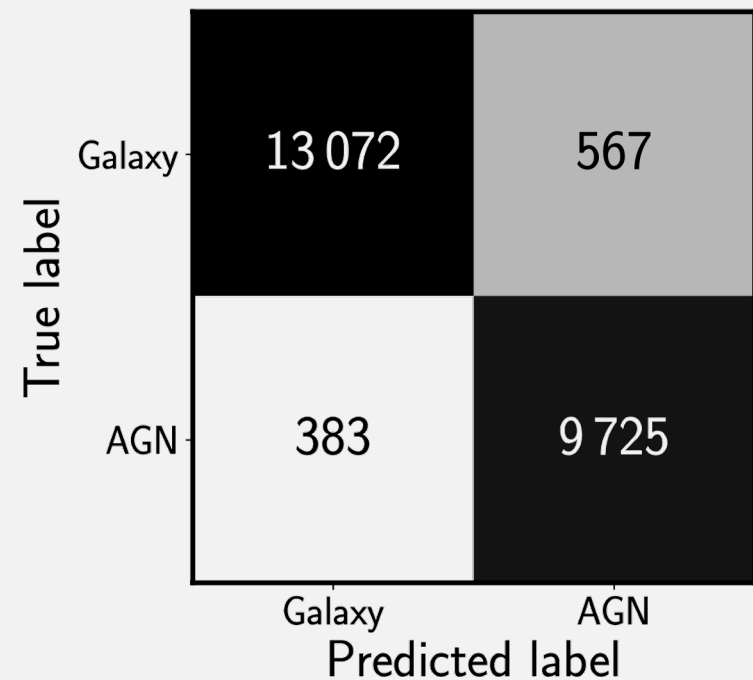
# Prediction Pipeline

- AGN/Galaxy + Radio-detection + Redshift.
- Model stacking: CatBoost, XGBoost, RF, ET, GBC/GBR (tabular data).
- Training: Photometry from IR-detected sources in HETDEX Spring Field.
- WISE + Pan-STARRS + 2MASS + LoTSS detection.
- Validation: HETDEX & Stripe 82

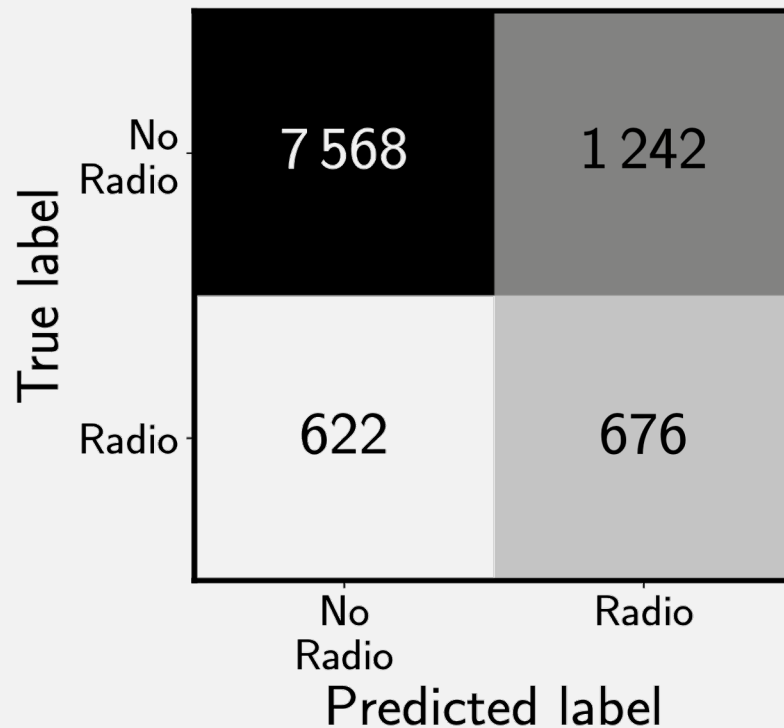


Carvajal et al., 2023

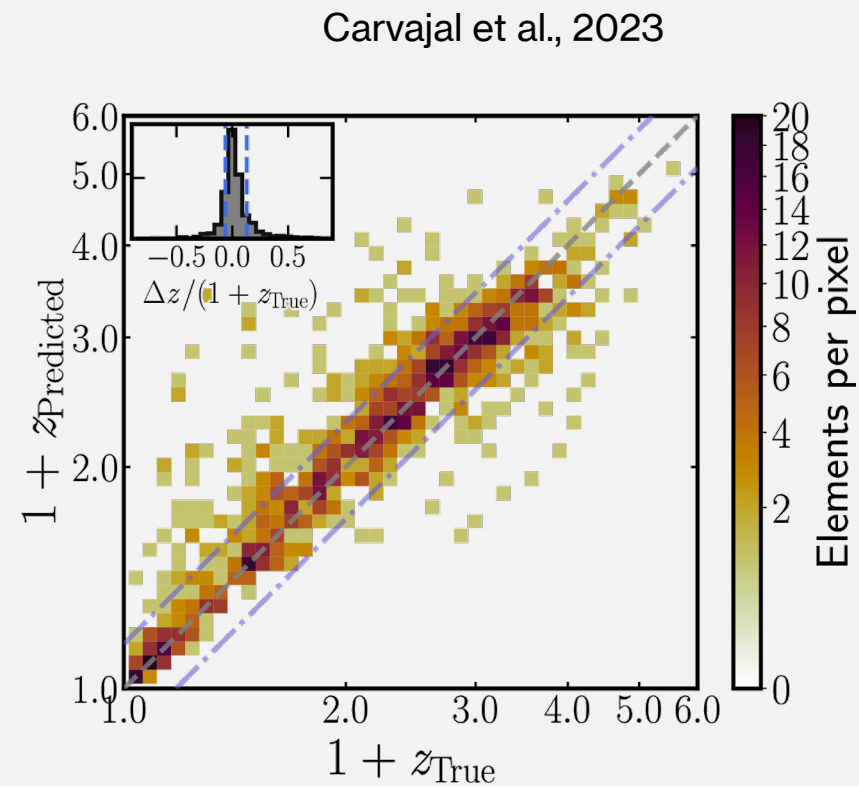
# RESULTS



Completeness: 96%  
Precision: 95%



Completeness: 52%  
Precision: 35%



$\sigma_{\text{NMAD}}$ : 7%  
Outlier fraction: 19%

# Feature Importances: SHAP values

- Properties with higher predicting power.
- Possible for individual sources.
- Example with AGN/Galaxy classification: W1-W2 & g-r.

AGN-Galaxy model					
Feature	SHAP value	Feature	SHAP value	Feature	SHAP value
W1_W2	32.458	i_y	5.086	z_y	1.591
g_r	11.583	y_W1	4.639	H_W3	1.048
W1_W3	8.816	band_num	4.050	W4mag	0.514
r_i	7.457	y_W2	3.228	H_K	0.466
i_z	6.741	z_W2	2.348	W3_W4	0.466
r_J	6.613	y_J	1.718	J_H	0.178

Radio detection model					
Feature	SHAP value	Feature	SHAP value	Feature	SHAP value
g_i	14.120	z_W1	6.751	W4mag	2.691
W2_W3	13.201	r_i	5.577	band_num	2.661
g_r	12.955	r_z	5.161	K_W4	0.939
y_J	8.224	i_z	4.512	H_K	0.719
K_W3	7.441	z_y	4.121	J_H	0.190
W1_W2	6.874	y_W1	3.864		

Redshift prediction model					
Feature	SHAP value	Feature	SHAP value	Feature	SHAP value
g_r	32.594	z_y	3.557	W4mag	1.639
y_W1	20.770	y_J	3.010	g_W3	1.479
W2_W3	12.462	band_num	2.595	K_W3	0.853
W1_W2	5.692	i_y	2.381	K_W4	0.451
r_i	4.381	H_K	2.230	J_H	0.146
r_z	3.755	i_z	2.005		

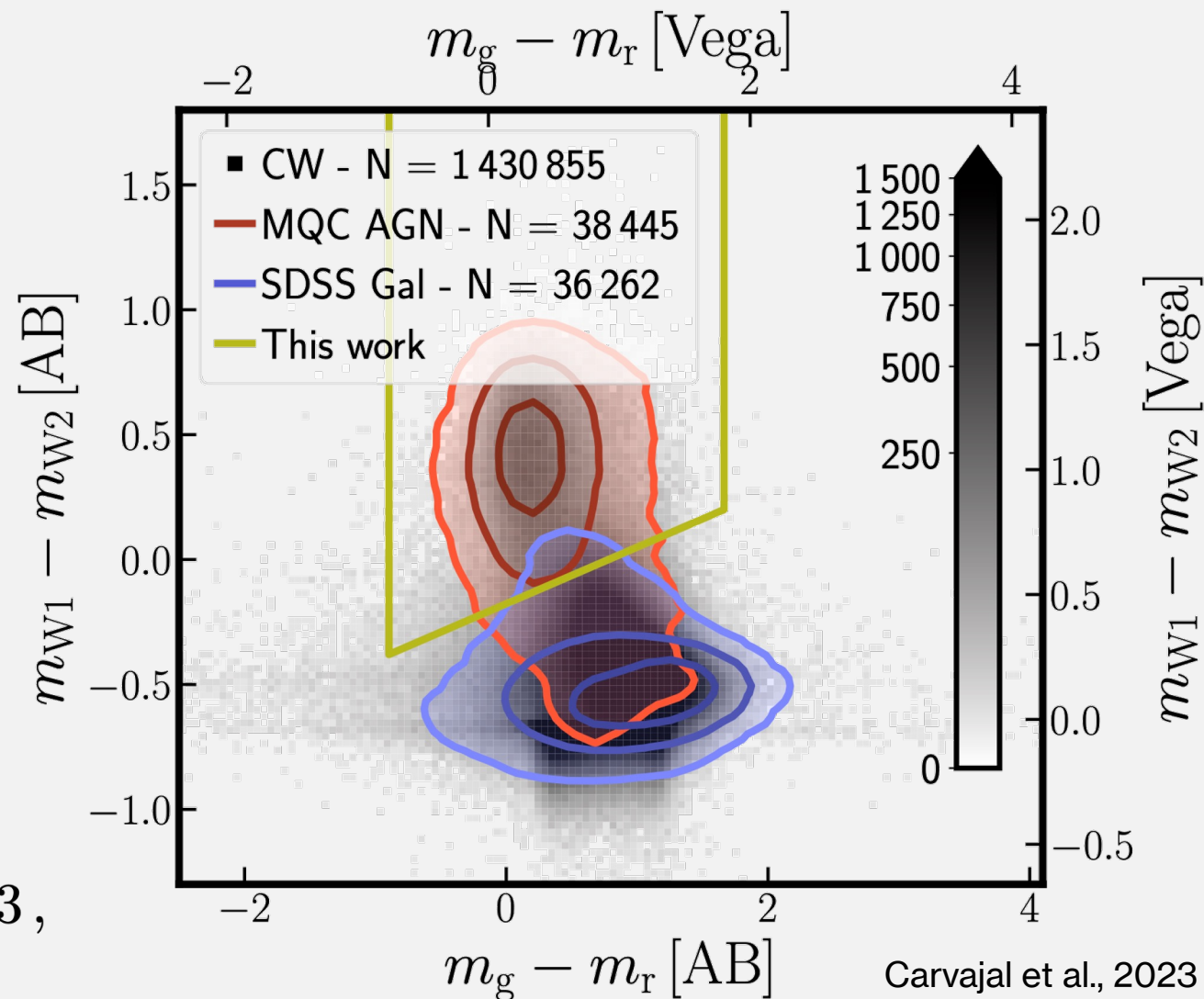
# A NEW COLOUR-COLOUR CRITERION

As efficient as previous IR colour-colour criteria.

$$g - r > -0.76,$$

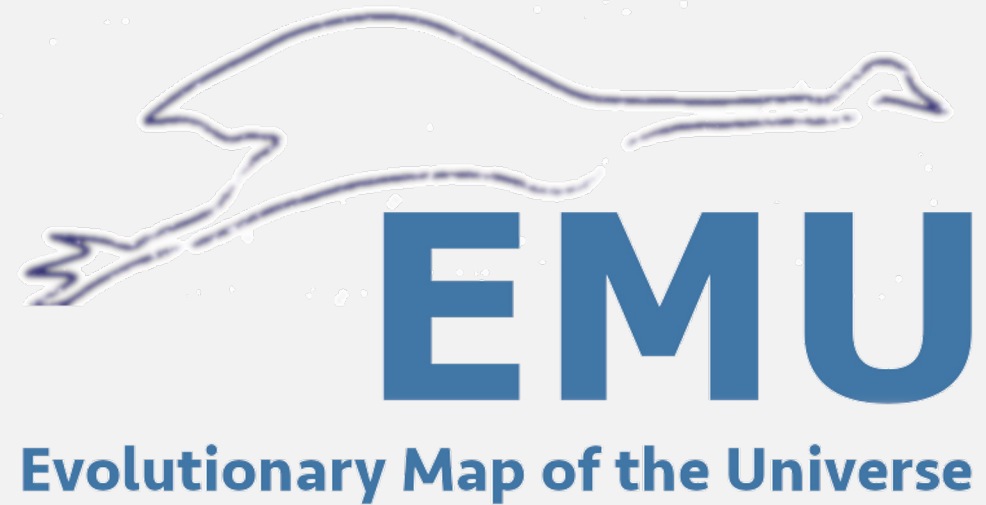
$$g - r < 1.8,$$

$$W1 - W2 > 0.227 \times (g - r) + 0.43,$$



# To summarise...

- (Astro-) Physical rules can be extracted from traditional ML models.
- One step towards AGN (radio) – Galaxy (IR+optical) connection.
- Possible to use in SKA and Pathfinders footprints and more.





**Thank you!**

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