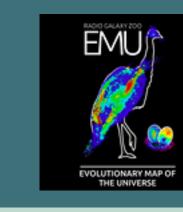


Where astrophysics meets citizen science: Radio Galaxy Zoo EMU (RGZ-EMU)



The making of RGZ-EMU

The Radio Galaxy Zoo for the Evolutionary Map of the Universe Survey (RGZ-EMU; Tang & Vardoulaki et al. in prep.) is a direct spin off from the very successful RGZ citizen science project. With novel methodologies, i.e. complexity and taxonomy, and the combined efforts from specialists and citizen scientists we aim to to tackle scientific questions about radio sources. Data are from the Evolutionary Map of the Universe (EMU) ASKAP [1] Pilot Survey at 944 MHz [2]. It observed 270 deg² of the southern sky, reaching an rms of 25–30 μ Jy/beam and a spatial resolution of 11–18 arcsec. The Selavy algorithm [3] was used to identify \sim 220K sources. The data are fed into the zooniverse.org platform (Fig. 1) into workflows we designed (Fig. 2), for the citizen scientists/zooters to access.

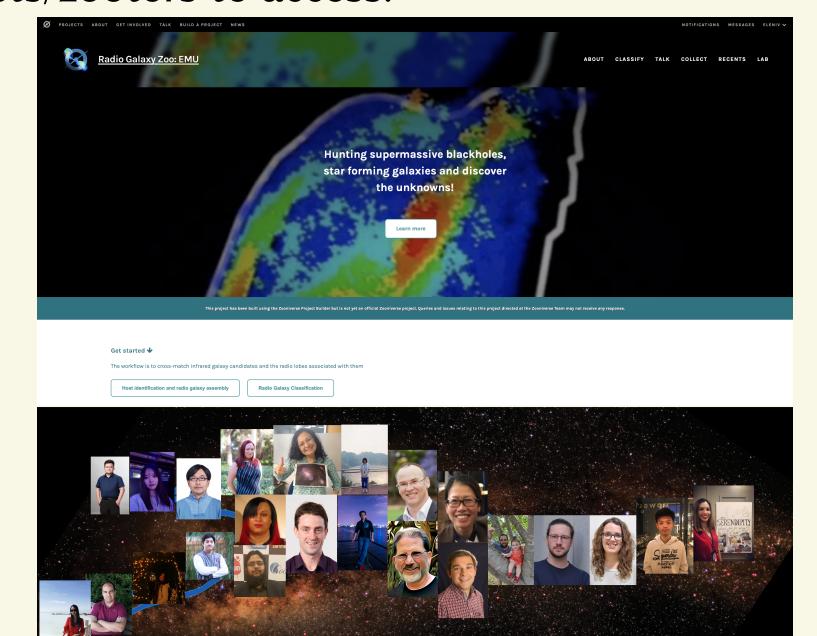


Fig. 1: (*Top*): **The RGZ-EMU Homepage** *It presents the two workflows: 1) source identification and assembly, and 2) source classification*, as well as details about the project, the team, and the education aspect of RGZ-EMU. (*Bottom*): Members of the RGZ-EMU team.

Novel methodologies

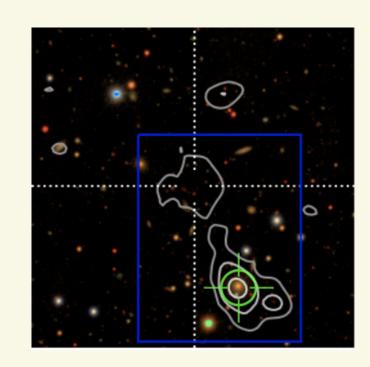
Anomaly detection to prioritise cutouts: Input data in the form of cutouts ($6' \times 6'$; Fig 2) are ranked using *complexity* [4, 5] to provide the most interesting objects to view and facilitate discoveries (Fig. 3).

A semantically meaningful taxonomy as a classification scheme: Semantically meaningful tags (Fig. 2–Right; [6] are derived (Fig. 4–Left), for the first time in astrophysics, using natural language processing (NLP). The aim is to provide a common language for classification between experts and to improve clarity in the communication between experts and citizen scientists.

The Science behind

RGZ-EMU has three science goals, presented via two workflows in the zoonivers.org platform:

- ▶ Identify the hosts of radio sources and assemble the different radio components into one parent source using overlays with radio, optical (DES; [7]), and infrared (WISE; [8]) images.
- Classify radio sources using descriptive tags [9] that were derived using results from early RGZ-EMU classification experiments and natural language processing [6].



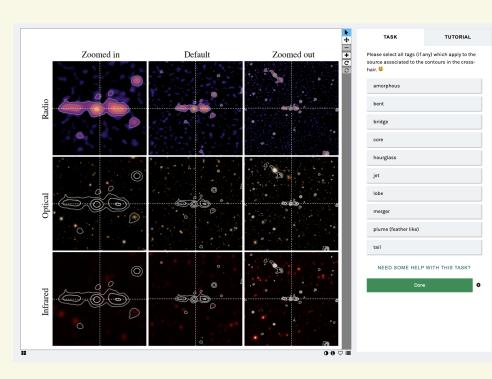
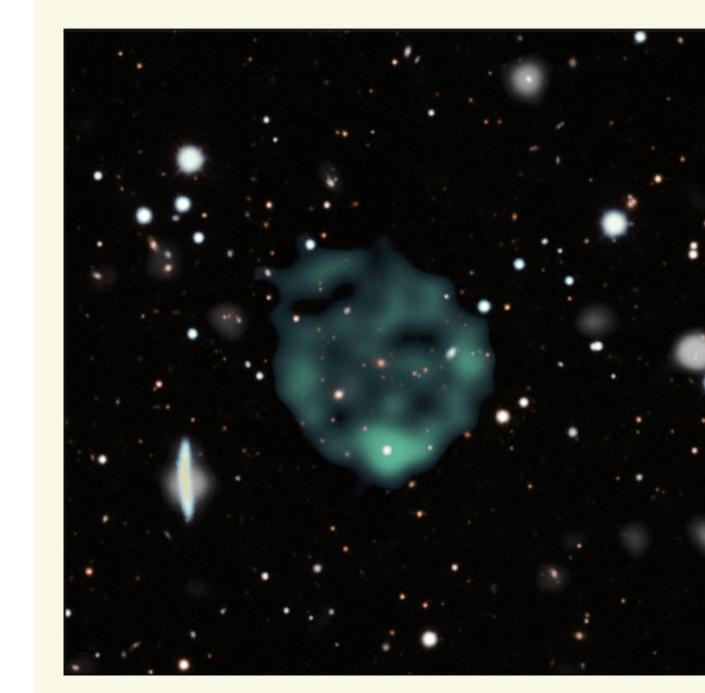


Fig. 2: Examples of the two workflows for the RGZ-EMU project release. (*Left:*) Zoom-in of the host identification (green circle) and source assembly (blue box) workflow. (*Right:*) Classification workflow based on tags. The setup of the cutouts in three different sizes $(3' \times 3', 6' \times 6', \text{ and } 12' \times 12')$ enhances workflow functionality and helps identify 95% of giant radio galaxies.

Hunting the unexpected

Radio surveys are filled with strange radio structures (Fig. 3). To facilitate discoveries, we applied the complexity methodology [4, 5] to provide the most interesting cutouts to view.



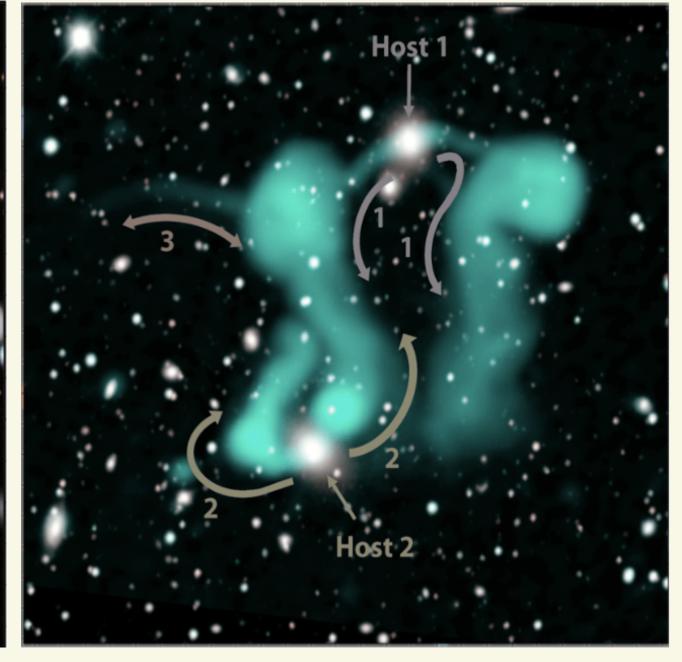
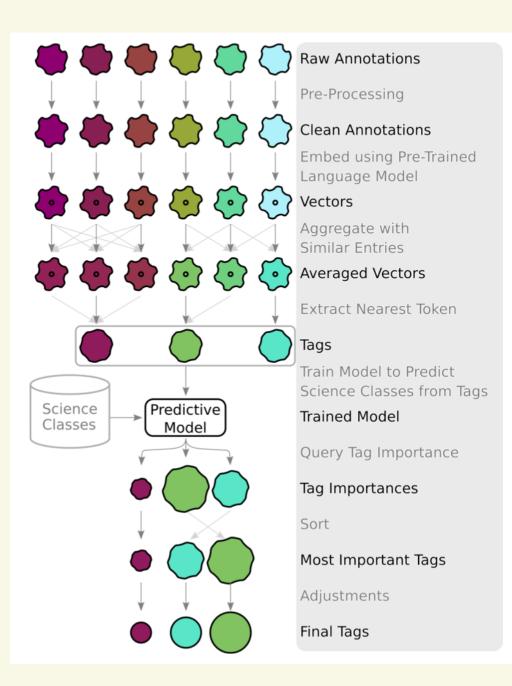


Fig. 3: *Left:* An example of Odd Radio Circle [2] . (*Right:*) An example of Peculiar Radio Source [10] . The source consists of a group of distorted radio components, collectively known as PKS 2130–538, and nicknamed "the dancing ghosts".

Pre-launch Early Science results

The projects below were developed during internal testing and preparations for launching the project to the public.

- ▶ **Semantic Morphology Taxonomy** is derived via natural language processing (Fig. 4–*Left*) and provides a set of tags to the RGZ-EMU workflow (Fig. 2–*Right*; [6]). These are meant to **replace astro jargon** (e.g. FRI, FRII, WAT). By using plain English terms, we can recover known scientific as well as rare sources with abnormal radio structures (Bowles et al. in prep.).
- ▶ **Self organising maps** (SOM; Vardoulaki, Tang, et al. (in prep.) are produced using the Parallelized rotation and flipping INvariant Kohonen (PINK; [11]) algorithm, with the purpose of radio-source classification and the discovery of peculiar radio sources (Fig. 3).



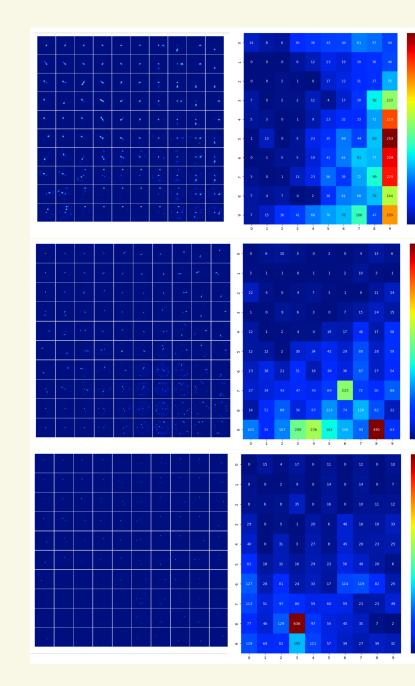


Fig. 4: (*Left:*) Proposed workflow to computationally derive a semantically meaningful plain English taxonomy from a set of annotations (Bowles et al. in prep.). (*Right:*) Testing the effect of different cutout sizes $(3' \times 3' - \text{top})$; $6' \times 6' - \text{middle}$; and $12' \times 12' - \text{bottom}$) as input to self organising maps using PINK [11] . The SOM was run by students (Anees, Dinh, Medina, Meyounyo, Mirshanova, Pöppelmann) at the Darmstadt University of Applied Sciences.

References

[1] R. Norris et al., PASA, **28**, 215, 2011[2] R. Norris et al., PASA, **38**, 3, 2021[3] M. Whiting & B. Humphreys, PASA, **29**, 371, 2012[4] G. Segal et al., PASP, **131**, 8007, 2019[5] G. Segal et al., arXiv:2206.14677, MNRAS, 2023[6] M. Bowles, et al., arXiv:2210.14760, NeurIPS, 2022; https://github.com/mb010/Text2Tag[7] T. M. C. Abbott et al., ApJS, **239**, 18, 2018[8] E. L. Wright et al., AJ, **140**, 1868, 2010[9] L. Rudnick et al., Galaxies, **9**, 85, 2021[10] R. Norris et al., PASA, **38**, 46, 2021[11] L. K. Polsterer et al., ASPC, **495**, 81, 2015

