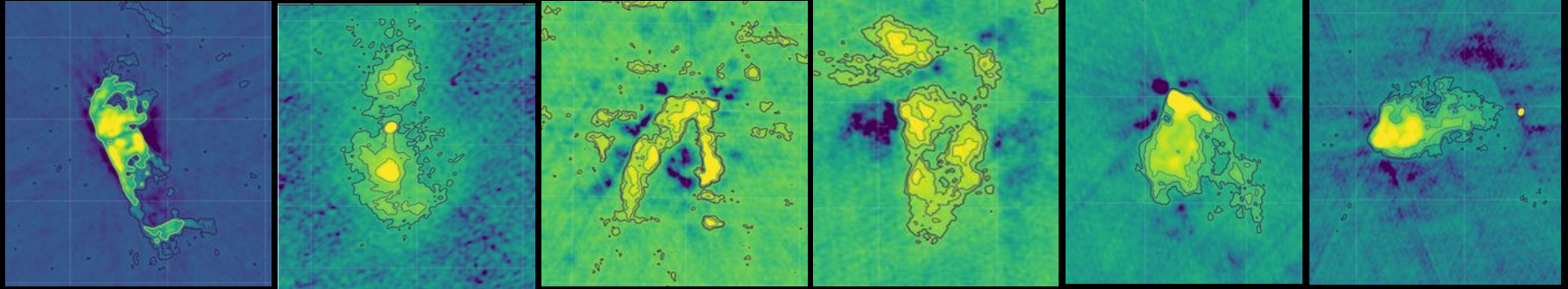


Anomaly Detection in the VLA Sky Survey (VLASS)



Viral Parekh

**National Radio Astronomy Observatory
Socorro, NM**

Acknowledgement to the VLASS team

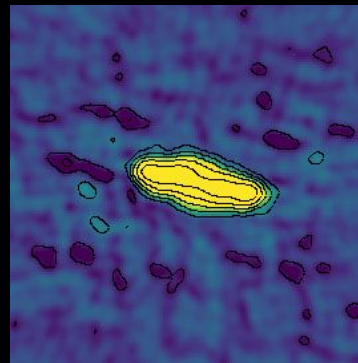
2nd Annual Cosmic Horizons Conference, July 15, 2026



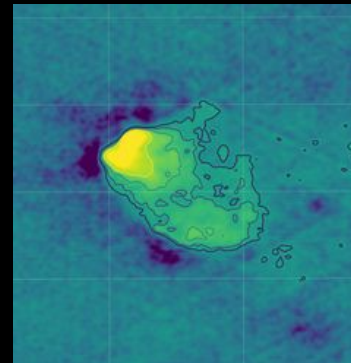
VLA Sky Survey (VLASS)



- ❑ VLA S-band (2-4 GHz) wide-area sky survey ($\delta > -40^\circ$, 33,885 deg^2 of sky)
- ❑ VLA B-configuration provides $\sim 2''$ resolution ($\sim 1'$ Largest angular scale)
- ❑ Multi-epoch observations (2017-2026)
- ❑ Single epoch rms $\sim 120 \mu\text{Jy}/\text{beam}$; combined images rms $\sim 70 \mu\text{Jy}/\text{beam}$
- ❑ Fully automated calibration and imaging pipeline
- ❑ Data products: Quick look, SE (self-cal) images, alpha (spx) maps, coarse cubes, catalogues, image combination, and image cutouts

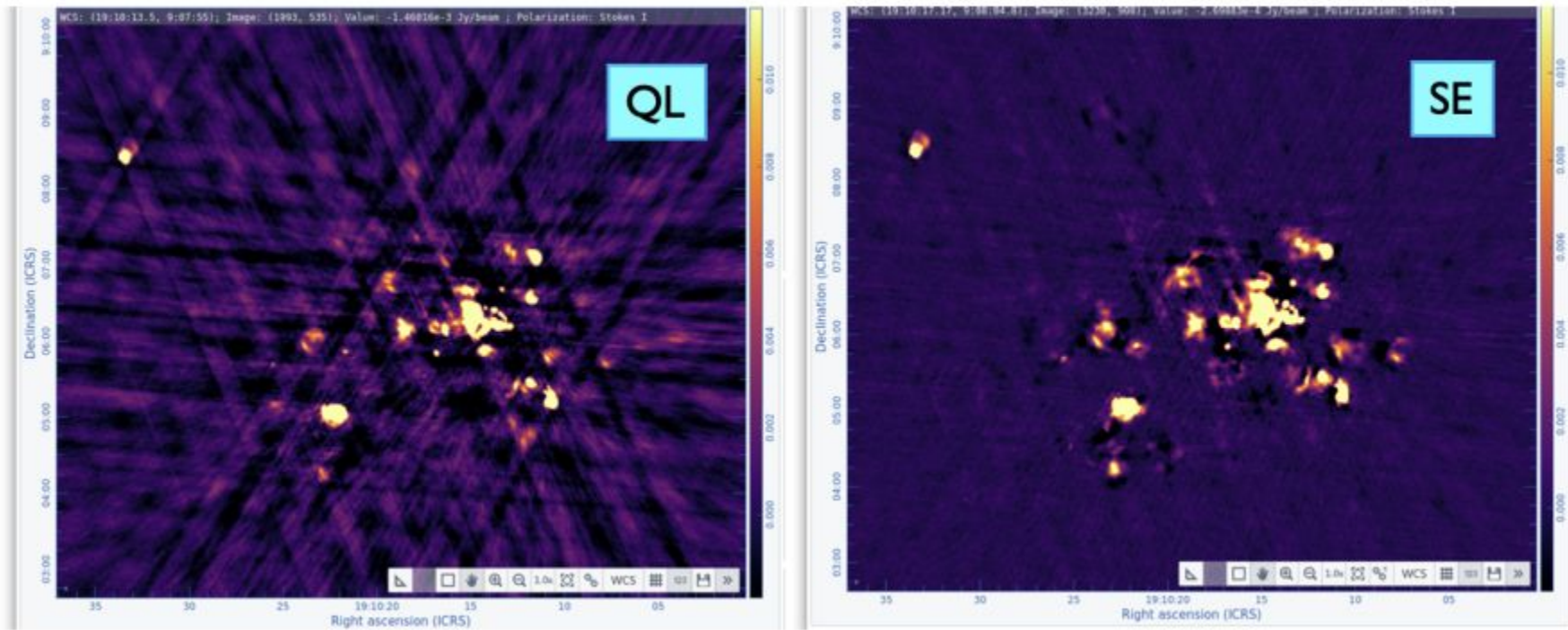


TGSS



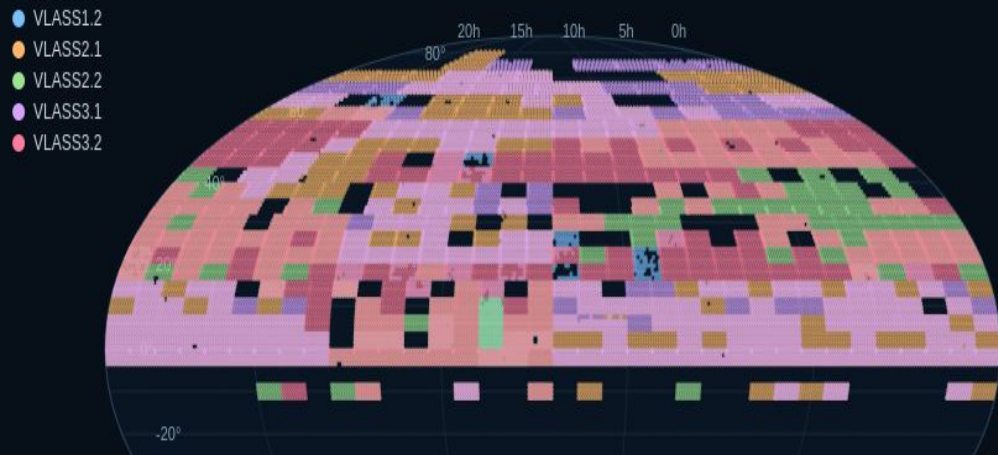
VLASS

QL vs SECI maps



(Amy Kimball et al. 2026)

VLA Sky Survey (VLASS) SECI Images



VLASS Epoch	No. of images	Area (deg ²)	Area (sr)	% of survey area
1.2	3560	2795	0.85	8.24
2.1	9106	7151	2.17	21.10
2.2	6213	4879	1.48	14.39
3.1	7905	6208	1.89	18.31
3.2	8219	6455	1.96	19.04

Epoch	Observing dates
1.1	Sep 2017–Feb 2018
1.2	Mar 2019–Jul 2019
2.1	Jun 2020–Nov 2020
2.2	Sep 2021–Feb 2022
3.1	Jan 2023–Jun 2023
3.2	Apr 2024–Oct 2024
4.1	Aug 2025–Feb 2026

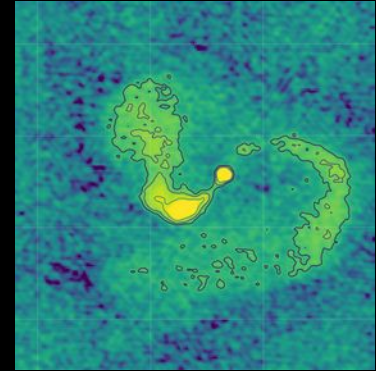
Extended Radio Sources in the VLASS

- ❑ Visual inspections of early QL maps (Hernandez and Andernach) of VLASS 1.1
 - ❑ Discovered 9 new giant radio galaxies of size > 1 Mpc
- ❑ DRAGNs in VLASS (Gordon et al. 2023): candidate double radio sources identified using the QL component catalogue
 - ❑ $> 17,000$ DRAGNs, 13,000 host galaxies, 31 new GRGs
- ❑ Morphology clustering of VLASS QL sources using Self-Organizing Maps (SOMs) by the CIRADA team

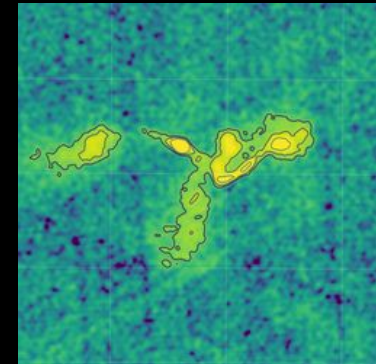
What is an anomaly?



- ❑ Non-Gaussian, elongated, complex, or irregular radio morphology
- ❑ Flexible across multiple source morphologies rather than targeting only one source class
- ❑ **Goal:** prioritize extended and morphologically unusual radio sources while suppressing compact sources, artifacts, and sidelobe structures
- ❑ **Benefit:** reduces the volume of data requiring expert visual inspection while preserving candidates with high scientific discovery potential



VLASSJ0615+5158

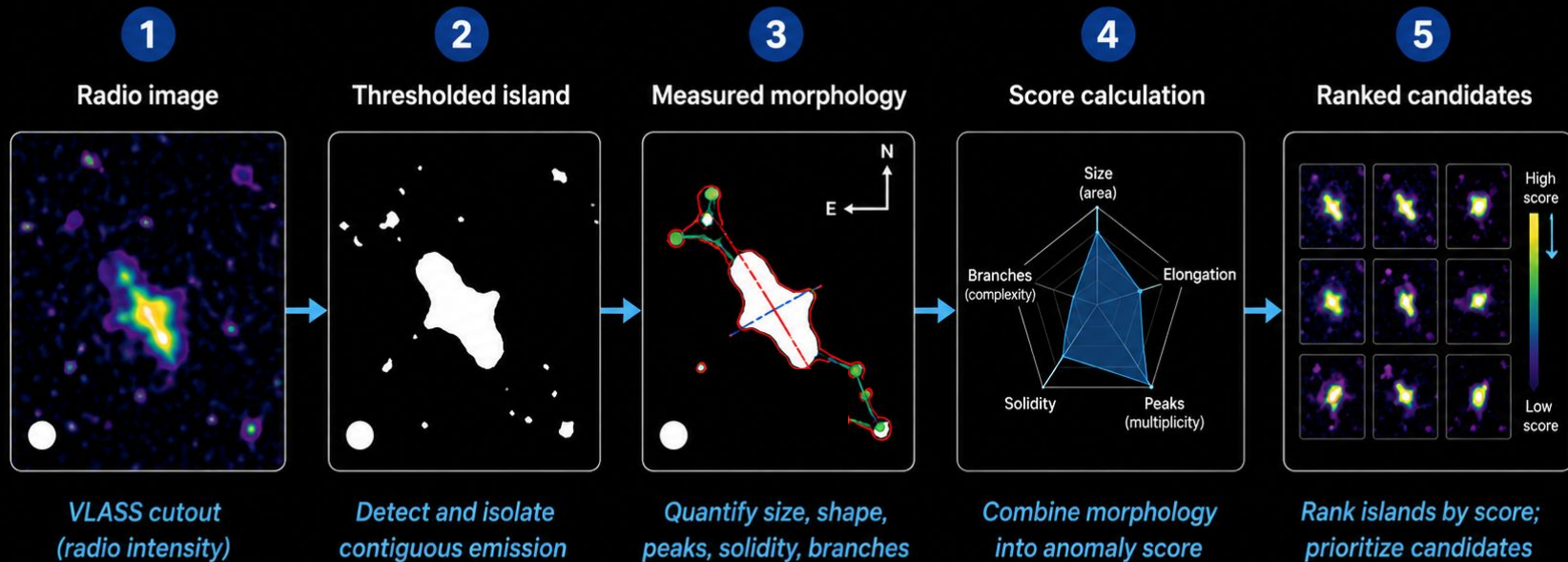


VLASSJ0707+3503

Anomaly detection in VLASS data

- ❑ Applied to VLASS 2.1 and 2.2 SECI images (9106 and 6213, respectively)
- ❑ **Radio-source morphology-based** anomaly detection
- ❑ Detect radio-source islands above a defined threshold
- ❑ Measure source morphology and structural complexity
- ❑ Assign an anomaly score to every detected candidate, including real sources and artifacts
- ❑ Use local background diagnostics to identify sidelobes and artifacts
- ❑ Build a sample of extended and unusual radio sources for future ML classification

Method

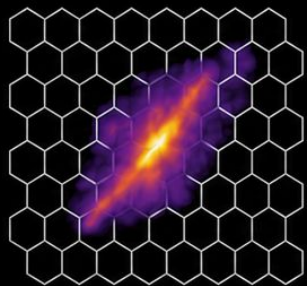


● = synthesized beam

Morphology parameters



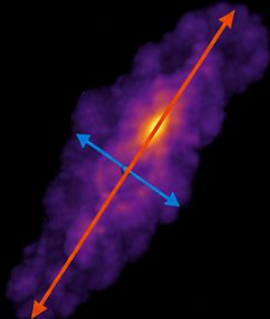
1 area_beams



○ = 1 beam

Size of the island in units of synthesized beams.

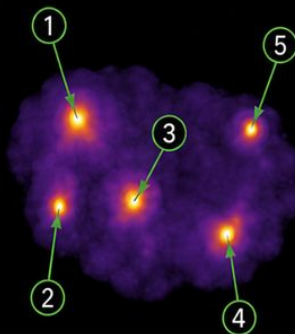
2 elongation



→ major axis
→ minor axis

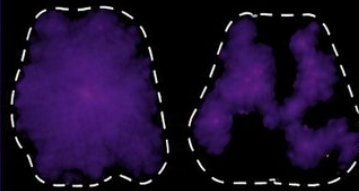
How stretched the source is.

3 n_peaks



Number of local brightness peaks in the local patch around the island.

4 solidity

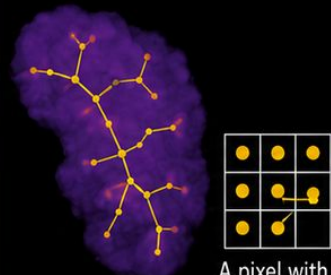


High solidity (convex)

Low solidity (concave)

How filled the island is compared to its convex hull.

5 branch_points



A pixel with ≥ 3 neighboring skeleton pixels is a branch point

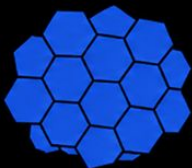
Number of skeleton branch points in the island.

How the score is calculated



$$\text{score} = S_{\text{area}} + S_{\text{elongation}} + S_{\text{peaks}} + S_{\text{solidity}} + S_{\text{branches}}$$

size



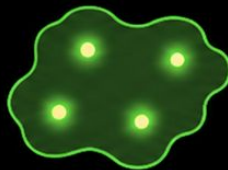
How large
the island is

shape



How stretched
or elongated
the island is

multiple peaks



How many local
brightness peaks
it contains

irregularity



How irregular or
non-convex the
island shape is

complexity



How complex
the skeleton
structure is



Weights set importance;
caps prevent one parameter from dominating.

Rejecting artifacts and sidelobes



VLASSJ1830+4831

RA,Dec: 277.709717, 48.517760 (18:30:50.33 +48:31:03.9)

Peak RA,Dec: 277.739458, 48.496685 (18:30:57.47 +48:29:48.1)

SIMBAD @ object: No match Type: NA Sep: nan arcsec z: nan Radio flux: nan

NED @ object: No match Type: NA Sep: nan arcsec z: nan Radio flux: nan

SIMBAD @ peak: No match Type: NA Sep: nan arcsec z: nan Radio flux: nan

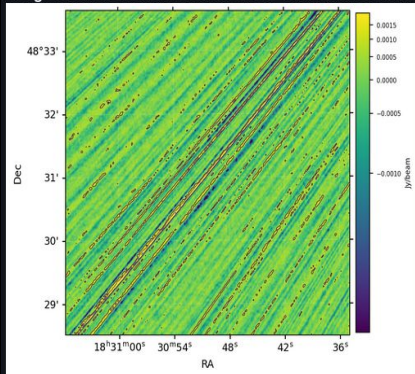
NED @ peak: WISEA J183057.89+482951.1 Type: UvS Sep: 5.25 arcsec z: nan Radio flux: 2.84e-08 Jy

Flux: Int=0.3281 ± 0.0166 Jy Peak=0.003229 Jy/beam Min=-0.001908 Std=0.0004545 RMS=0.0004545

Score: 6411.34 (orig 6411.34) peakSNR: 21.3 area(beams): 228.46 elong: 146.38 peaks: 1585 solidity: 0.66 branches: 57

bgAlign: 1.00 bgAniso: 0.92 negSide: 0.620 ridgeStraight: 1.00

Image: VLASS2.1.se.T23t19.J183252+483000.06.2048.v1.l.iter3.image.pbcor.tt0.subim.fits



VLASSJ0855+4911

RA,Dec: 133.983713, 49.187000 (08:55:56.09 +49:11:13.2)

Peak RA,Dec: 133.984886, 49.186395 (08:55:56.37 +49:11:11.0)

SIMBAD @ object: NVSS J085556+491112 Type: Radio Sep: 0.73 arcsec z: nan Radio flux: nan

NED @ object: No match Type: NA Sep: nan arcsec z: nan Radio flux: nan

SIMBAD @ peak: SDSS J085556.36+491110.6 Type: RadioG Sep: 0.32 arcsec z: nan Radio flux: nan

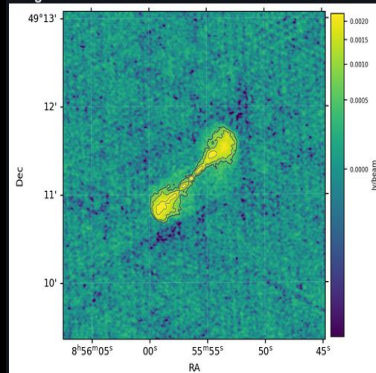
NED @ peak: No match Type: NA Sep: nan arcsec z: nan Radio flux: nan

Flux: Int=0.2304 ± 0.0116 Jy Peak=0.005139 Jy/beam Min=-0.0004731 Std=0.0001184 RMS=0.0001184

Score: 89.21 (orig 151.92) peakSNR: 46.5 area(beams): 169.18 elong: 5.53 peaks: 22 solidity: 0.59 branches: 38

bgAlign: 0.87 bgAniso: 0.20 negSide: 0.002 ridgeStraight: 0.49

Image: VLASS2.2.se.T23t09.J085726+493000.06.2048.v1.l.iter3.image.pbcor.tt0.subim.fits



Example candidate-retention cuts: $\text{neg_sidelobe} < 0.20$; $\text{elongation} < 14$; $\text{n_peaks} < 50$

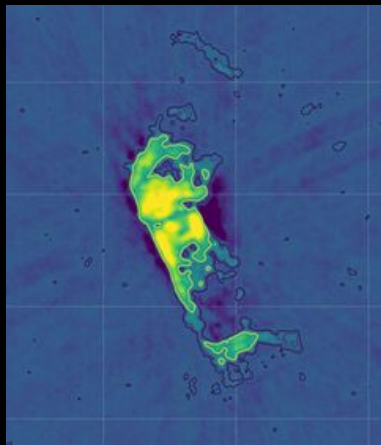
Anomaly data volume



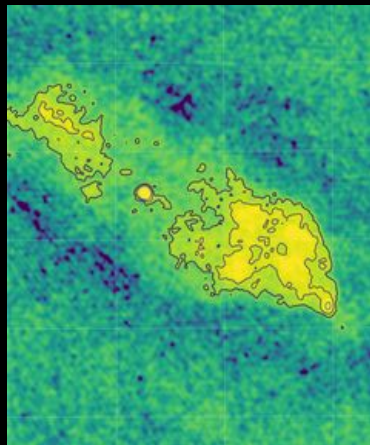
- ❑ VLASS 2.1: 26,158 initial cutouts → 8,413 candidates after filtering, corresponding to a **67.8% reduction** in visual-inspection volume.
- ❑ VLASS 2.2: 22,115 initial cutouts → 6,875 candidates after filtering, corresponding to a **68.9% reduction** in visual-inspection volume.



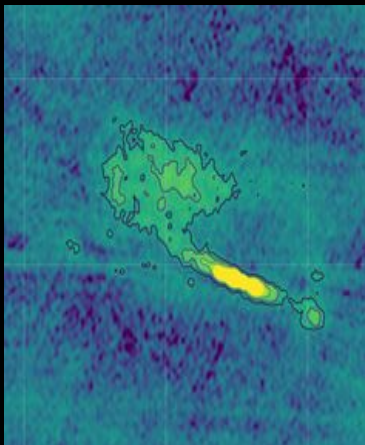
Results: Top-ranked complex radio sources



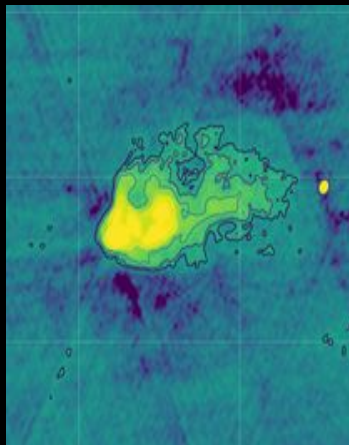
VLAJ2027+3722
Score 424



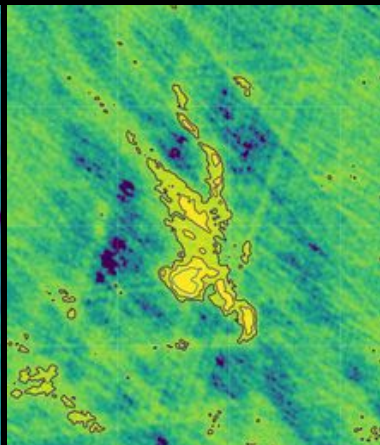
VLAJ0207+2930
Score 419



VLAJ1724-0242
Score 404



VLAJ1602+0157
Score 395

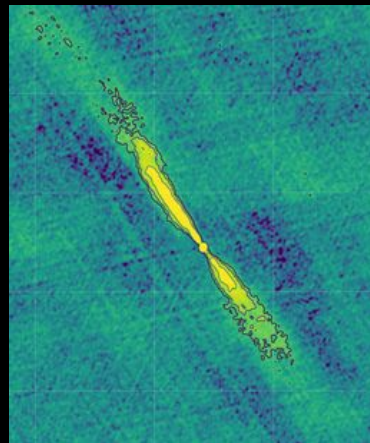


VLAJ2032+4016
Score 390

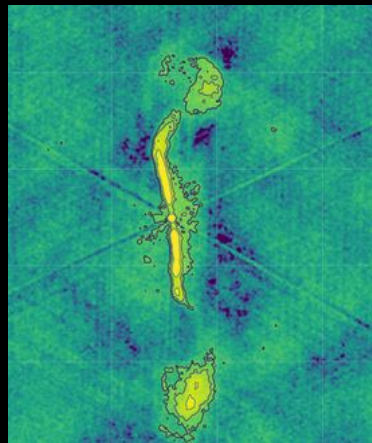
Results: Top-ranked extended radio sources



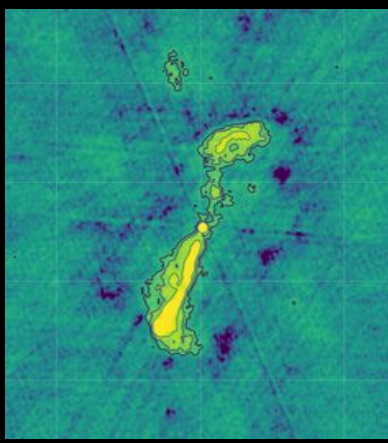
Each candidate is cross-matched with the SIMBAD and NED databases



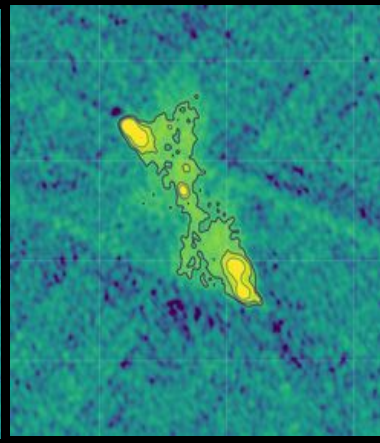
VLAASSJ1416+1048
Score 232



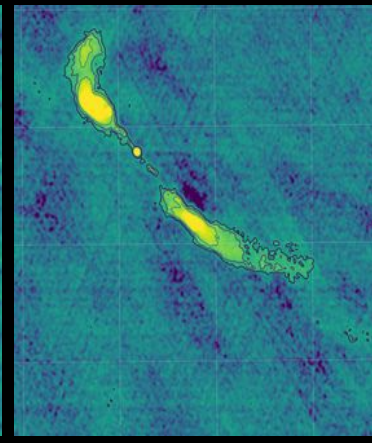
VLAASSJ2231+3921
Score 167



VLAASSJ0057-0123
Score 162



VLAASSJ0849+0949
Score 152

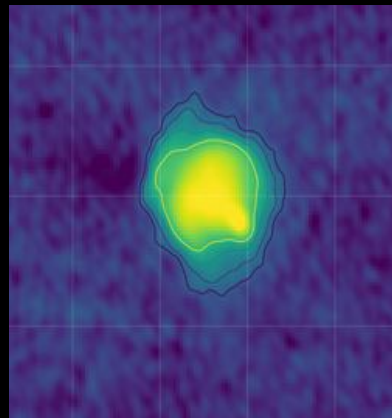


VLAASSJ0452+5204
Score 142

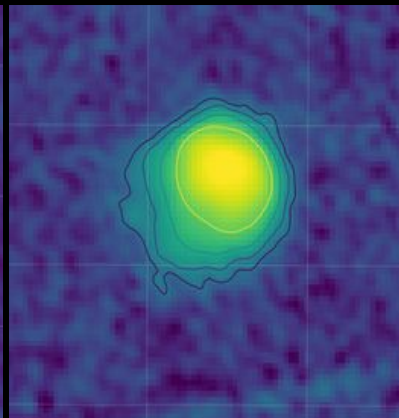
Results: Top-ranked roundish radio sources



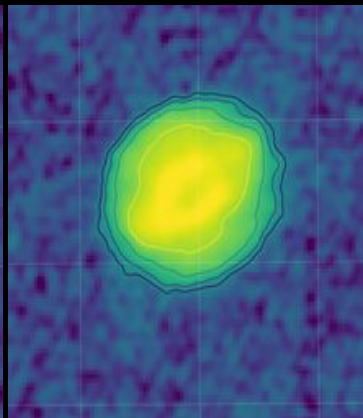
Science interest: Searching for radio halos in galaxy clusters, Odd Radio Circles (ORCs), and high-redshift objects.



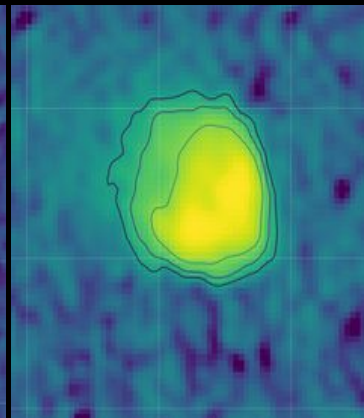
VLASSJ0643-0108
Score 190



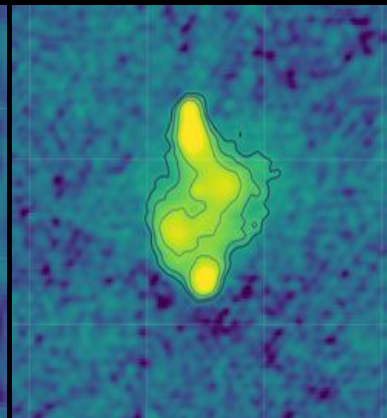
VLASSJ0436+5112
Score 180



VLASSJ1944+5031
Score 170



VLASSJ1901+0504
Score 160



VLASSJ0752+3252
Score 120

<https://github.com/viralp/VLASS-anomaly>

Future ML-based classification of radio sources



Use the current anomaly catalogue and cutouts as a seed dataset for image-similarity search, self-supervised learning, and active learning.

1 Current outputs

candidates.csv
FITS/PNG cutouts
Morphology scores

2 Embedding search

Convert each cutout
into an image vector
“find more like this”

FiftyOne + FAISS

3 Self-supervised ML

Train on unlabeled
VLASS cutouts so model
learns radio morphology

PyTorch + Lightly/BYOL

4 Active learning

Human labels only the
most useful examples:
real / artifact / type

modAL + scikit-learn

5 ML catalogue

Class labels
probabilities
review notes +
exports

Immediate next step

Load a small VLASS anomaly subset into FiftyOne → compute pretrained embeddings → select seed examples → retrieve nearest neighbours → export a ranked review set.

Then train BYOL/Lightly on clean radio cutouts and use active learning to improve labels with minimal human review.