

# Anomaly Detection in Astronomy

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South African Radio Astronomy  
Observatory



UNIVERSITY *of the*  
WESTERN CAPE



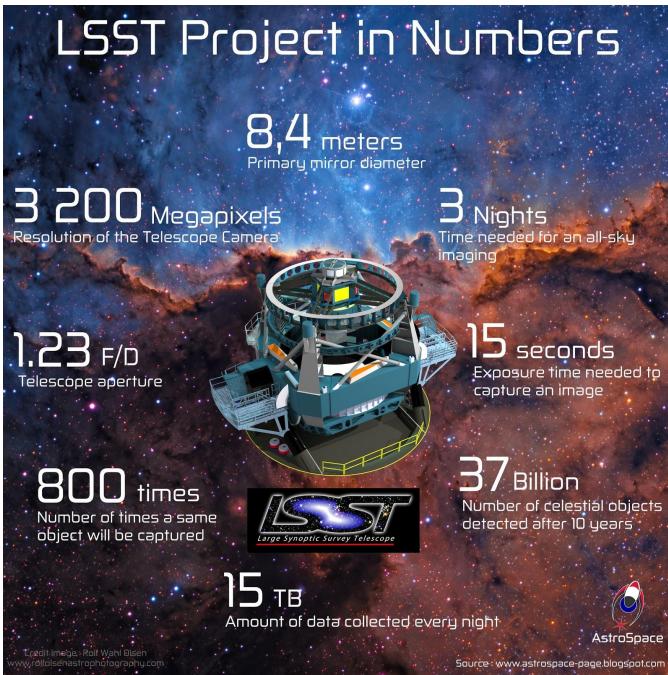
NRF  
National Research Foundation  
SARAO  
South African Radio Astronomy Observatory



science & innovation  
Department:  
Science and Innovation  
REPUBLIC OF SOUTH AFRICA

# The Vera C. Rubin Observatory

## Legacy Survey of Space and Time (LSST)



Rubin Obs/NSF/AURA & Bruno C. Quint

Day 000



Vera C. Rubin Observatory

# The Square Kilometre Array



## SKA1-mid

the SKA's mid-frequency instrument



Frequency range:  
**350 MHz to 15.3 GHz**  
with a goal of 24 GHz

197 dishes  
(including 84 MeerKAT dishes)

Maximum baseline:  
**150km**

## SKA1-low

the SKA's low-frequency instrument



Frequency range:  
**50 MHz to 350 MHz**

~131,000  
antennae pairs between  
512 stations

Maximum baseline:  
**~65km**

Square Kilometre Array Observatory

# MeerKAT



SARAO

# ASKAP



CSIRO

# LOFAR



LOFAR

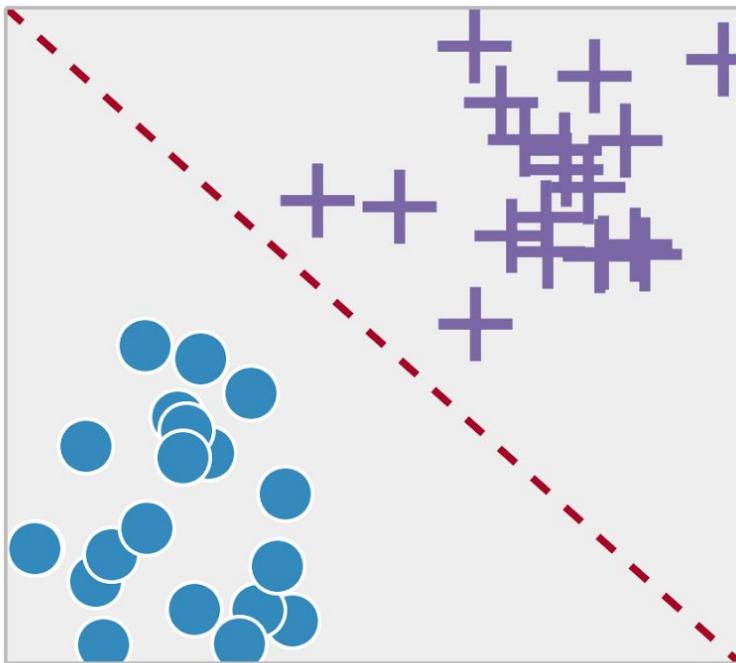
# We're facing a data explosion



Harvard Business Review

# Machine Learning

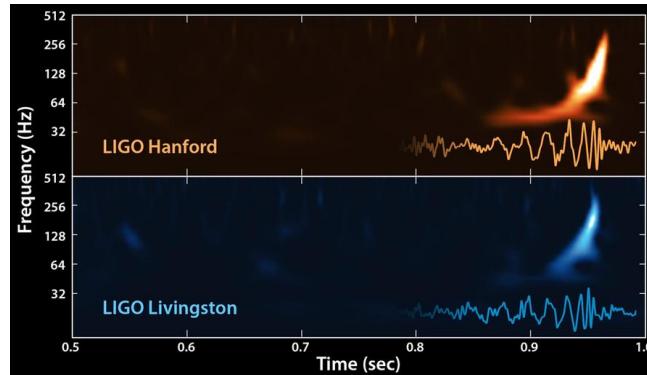
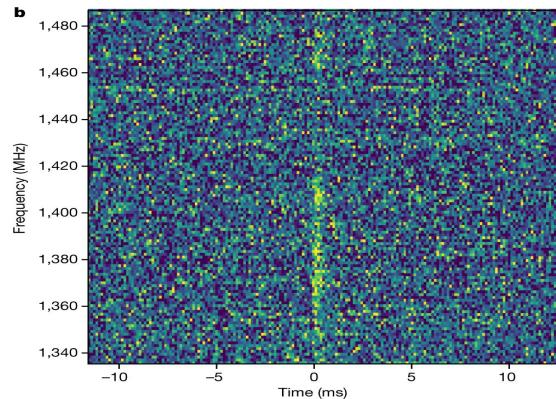
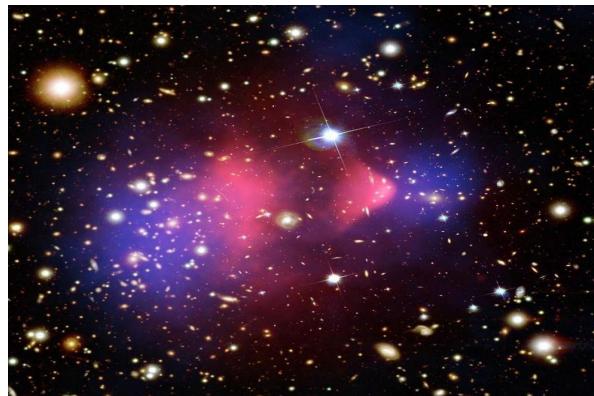
# Supervised Machine Learning



Automatically learns a model to map inputs to outputs, using a training set.

# Anomaly Detection

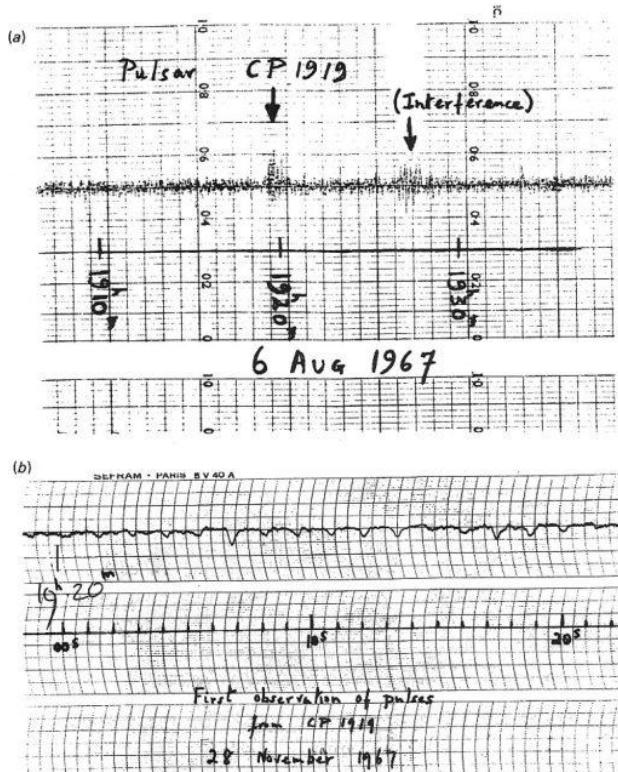
# Known Unknowns - rare events



Ravi et al. (2019)

Hubble/ NASA/ ESA

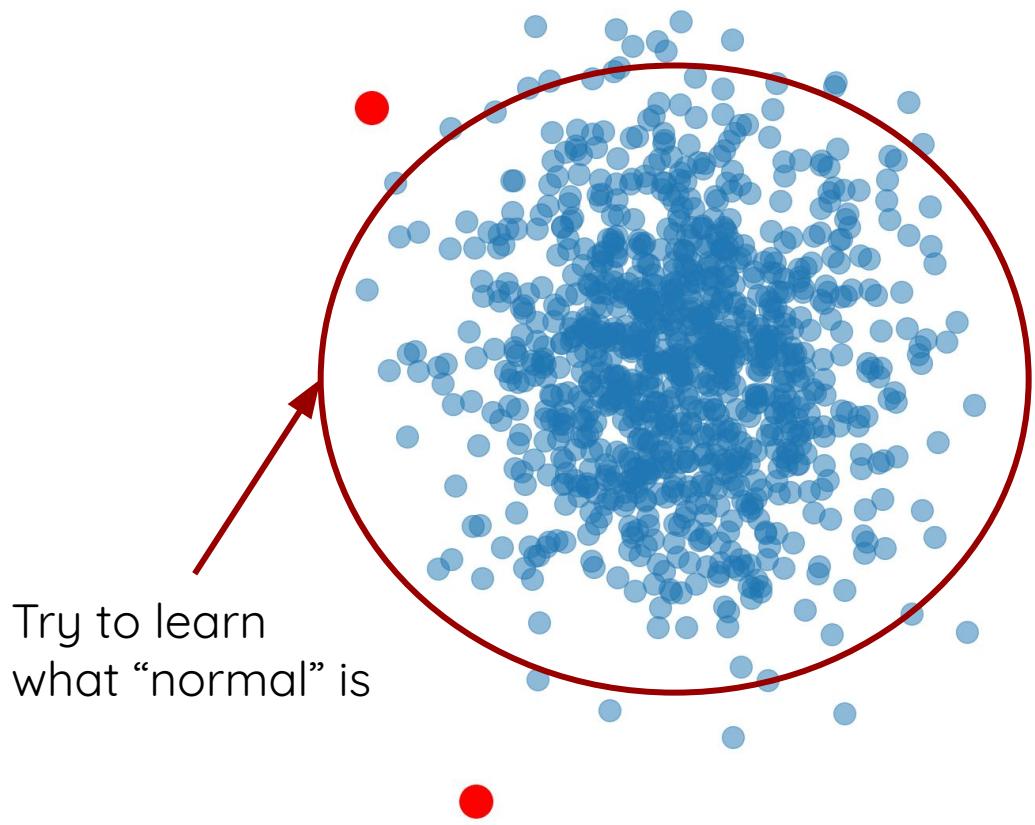
# Unknown Unknowns - new anomalies



How do we discover new phenomena...

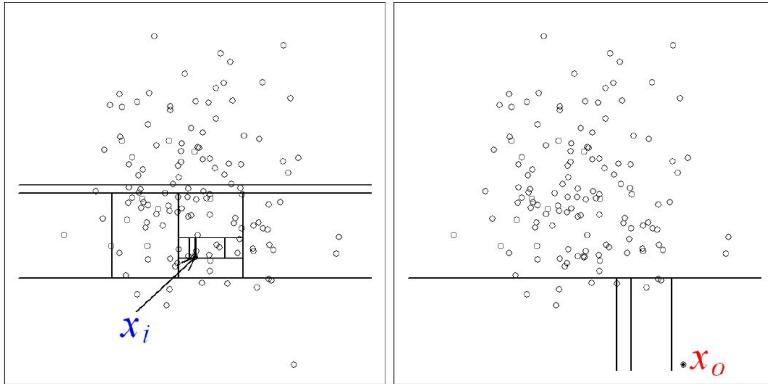
...among 10 million possibilities?

# Anomaly Detection



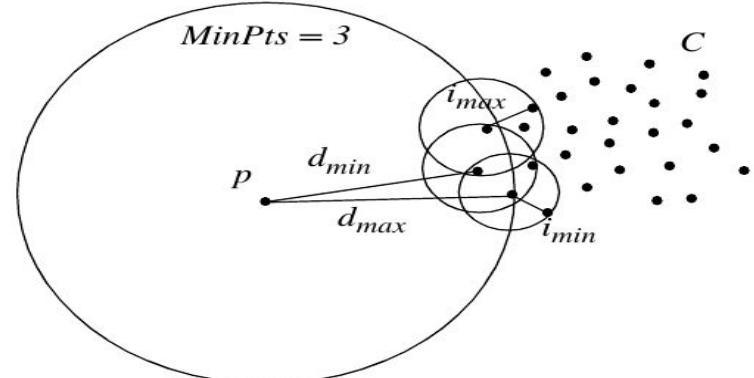
The farther away  
from normal the  
higher the  
anomaly score

# Anomaly Detection Algorithms

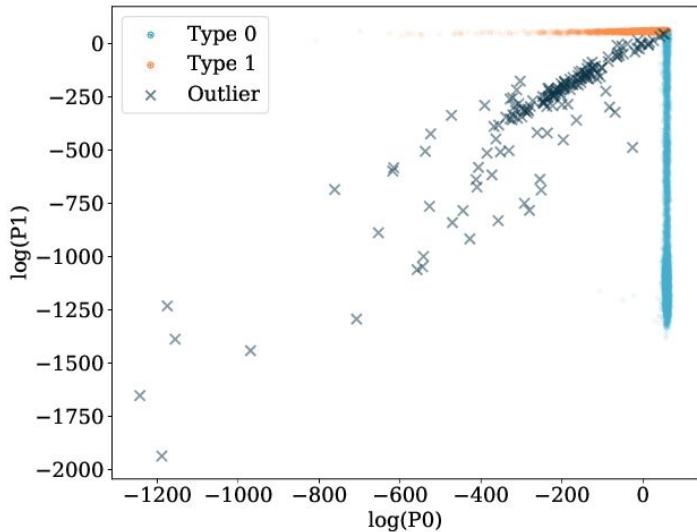


Isolation Forest (Liu,  
Ting & Zhou; 2008)

Local Outlier Factor  
(Breunig et al; 2008)

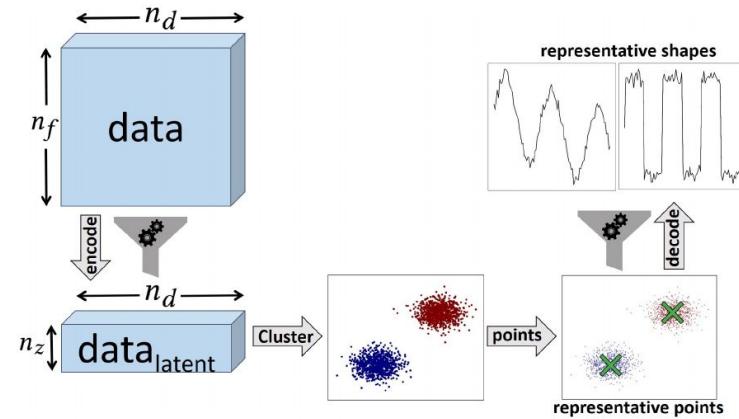


# Anomaly Detection Algorithms



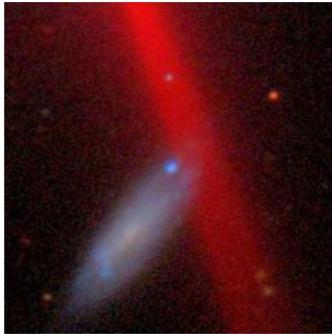
**BADAC** (Roberts, Bassett &  
Lochner - 1902.08627)

**DRAMA** (Vafaei Sadr, Bassett  
& Kunz - 1909.04060)

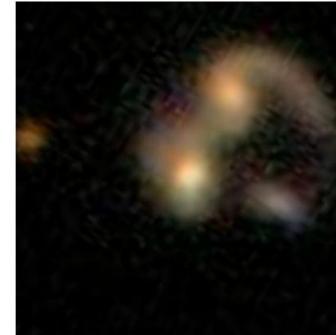


# Anomaly Detection Isn't Enough

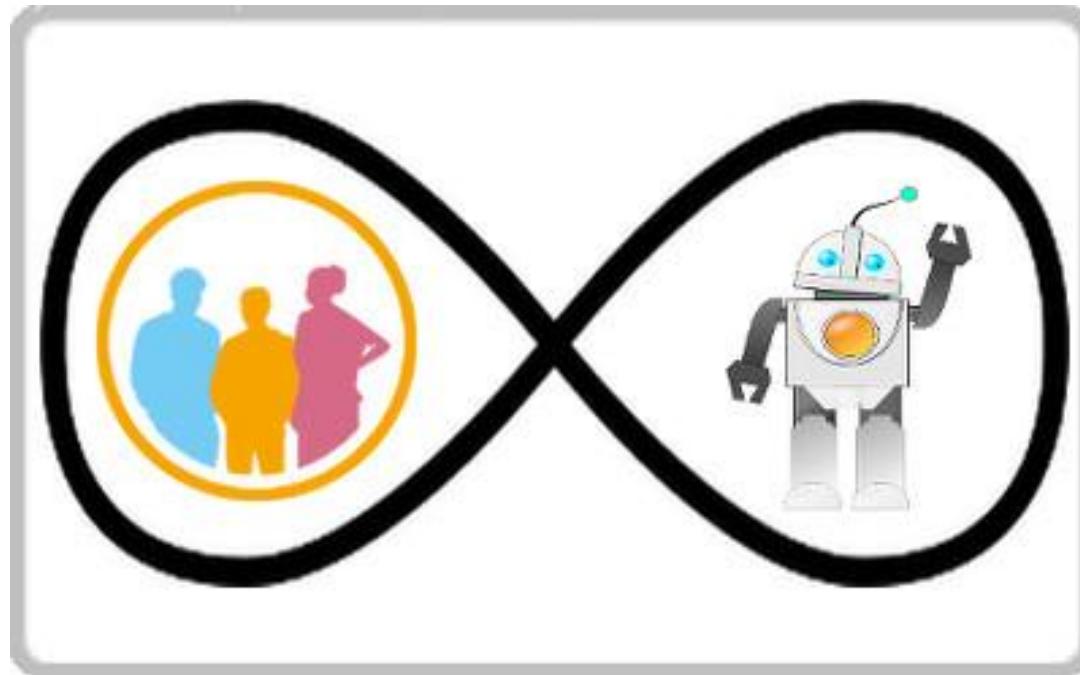
Artefacts



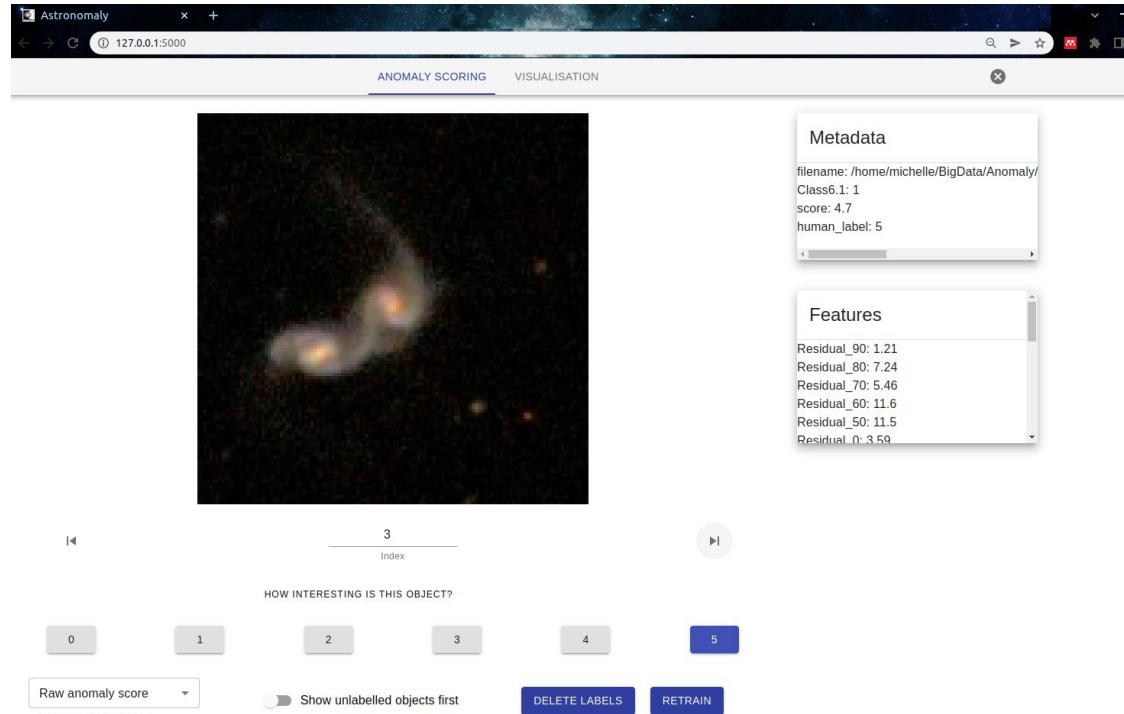
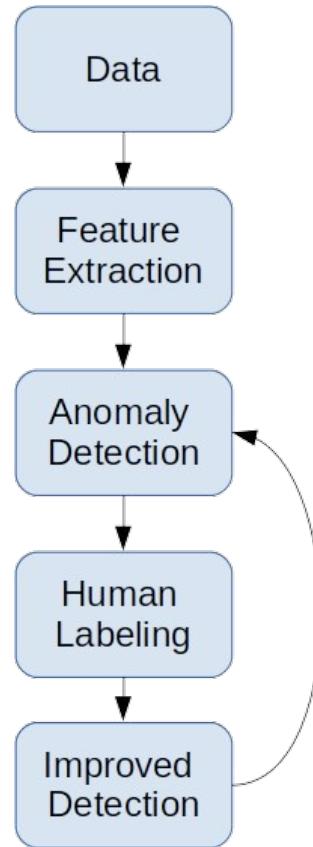
Real sources



# Active Learning



# Astronomaly



Lochner and Bassett (2020) - 2010.11202  
<https://github.com/MichelleLochner/astronomaly>

# Galaxy Zoo - Random Examples

The screenshot shows a web browser window titled "Astronomy" at the URL "127.0.0.1:5000". The main content area displays a large, elliptical galaxy with a bright central bulge and a surrounding disk. Below the image is a horizontal slider labeled "Index" with the value "0" in the center. At the bottom of the page, there is a section titled "HOW INTERESTING IS THIS OBJECT?" with numbered buttons from 0 to 5. A dropdown menu labeled "Random" is open, indicating the scoring method. To the right of the image, two panels provide detailed information: "Metadata" lists the filename and various scores, and "Features" lists several residual values.

ANOMALY SCORING    VISUALISATION

Metadata

filename: /home/michelle/BigData/Anomaly/  
Class6.1: 0.279  
score: 0.0847  
human\_label: -1

Features

Residual\_90: -0.364  
Residual\_80: -0.277  
Residual\_70: 0.299  
Residual\_60: -0.326  
Residual\_50: -0.529  
Residual\_0: -0.428

0  
Index

HOW INTERESTING IS THIS OBJECT?

0 1 2 3 4 5

Random

Show unlabelled objects first

DELETE LABELS RETRAIN

Scoring method to sort by

# Galaxy Zoo - No Active Learning

The screenshot shows a web browser window titled "Astronomy" with the URL "127.0.0.1:5000". The interface is divided into two main sections: "ANOMALY SCORING" (selected) and "VISUALISATION".

**Visualisation:** A large image of a spiral galaxy with a prominent central bulge and distinct spiral arms.

**Anomaly Scoring:** A horizontal slider labeled "Index" with a value of "0". Below the slider, a question asks "HOW INTERESTING IS THIS OBJECT?". Six buttons are shown: 0, 1, 2, 3, 4, and 5, with button 5 being highlighted in blue.

**Metadata:** A panel containing the following information:

- filename: /home/michelle/BigData/Anomaly/
- Class6.1: 0.916
- score: 5
- human\_label: 5

**Features:** A panel listing residual values:

- Residual\_90: 2.58
- Residual\_80: 23.2
- Residual\_70: 12.9
- Residual\_60: 8.1
- Residual\_50: 5.38
- Residual\_0: 3.19

**Buttons and Links:**

- Raw anomaly score ▾
- Show unlabelled objects first (radio button)
- DELETE LABELS
- RETRAIN

Scoring method to sort by

# Galaxy Zoo - Active Learning

Astronomy    x    +  
127.0.0.1:5000

ANOMALY SCORING    VISUALISATION



Metadata

```
filename: /home/michelle/BigData/Anoma
Class6.1: 0.916
score: 5
human_label: 5
predicted_user_score: 4.61
```

Features

```
Residual_90: 2.58
Residual_80: 23.2
Residual_70: 12.9
Residual_60: 8.1
Residual_50: 5.38
Residual_0: 3.19.
```

Index: 0

HOW INTERESTING IS THIS OBJECT?

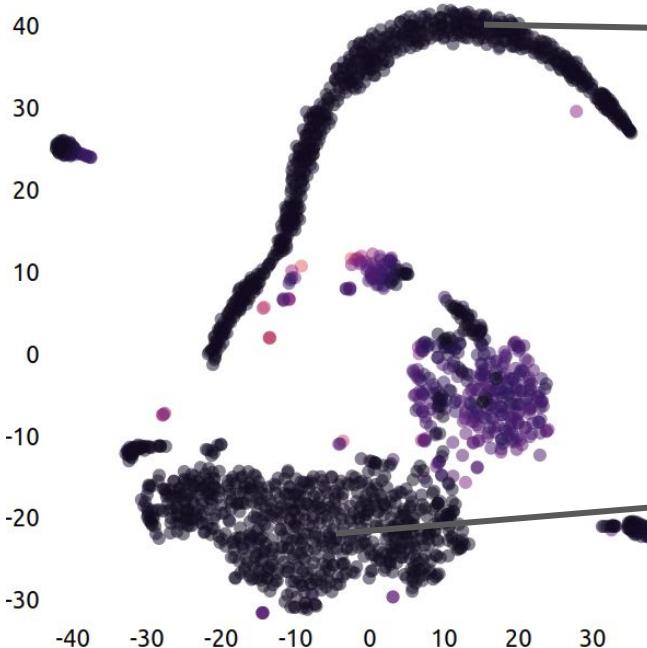
0    1    2    3    4    5

Active learning score ▾  
Scoring method to sort by

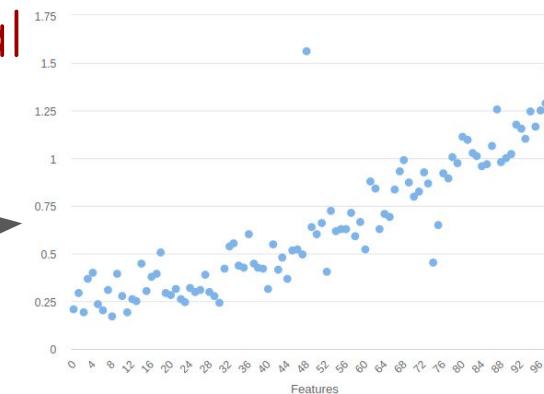
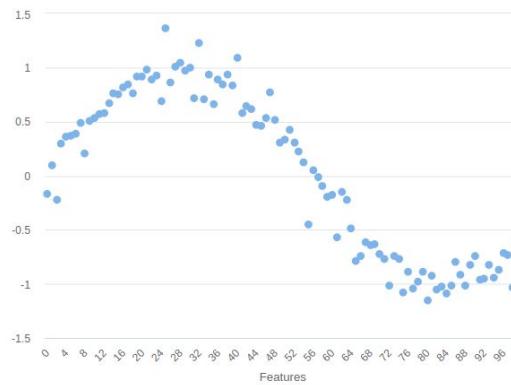
Show unlabelled objects first

DELETE LABELS    RETRAIN

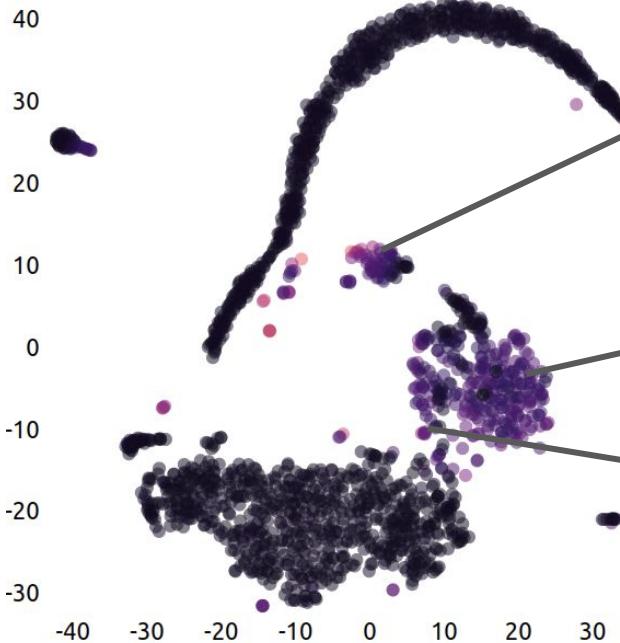
# Synthetic data



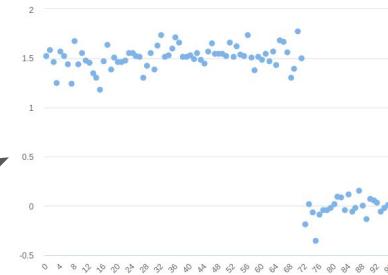
Normal



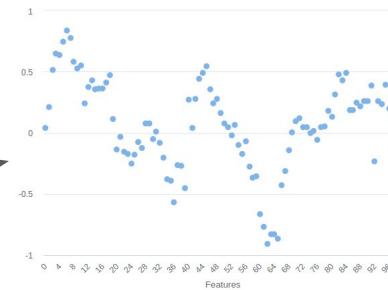
# Synthetic data



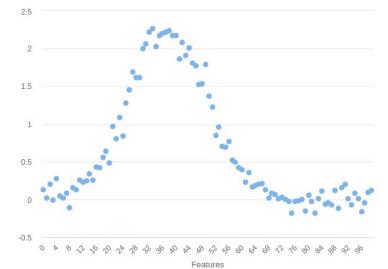
Anomalous



Anomaly Score: 3.112012

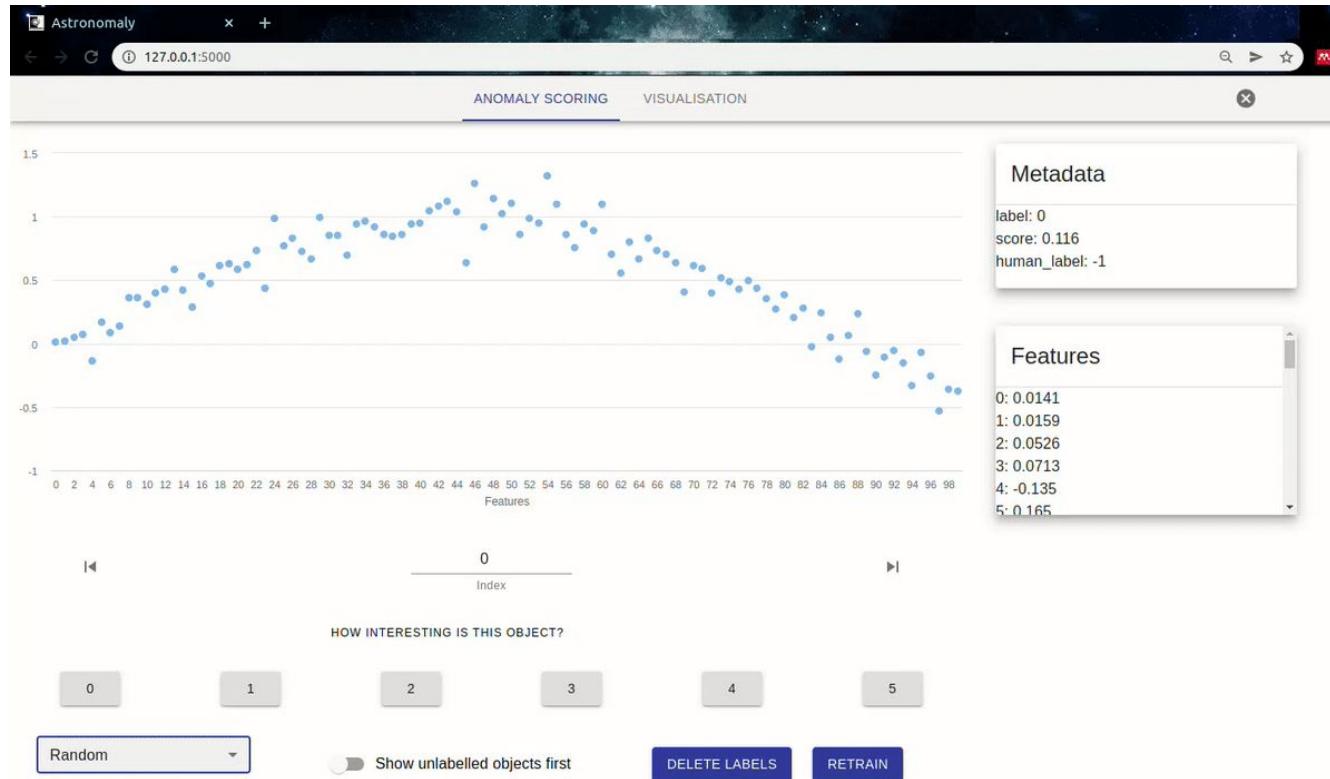


Anomaly Score: 1.303772



Anomaly Score: 3.92992

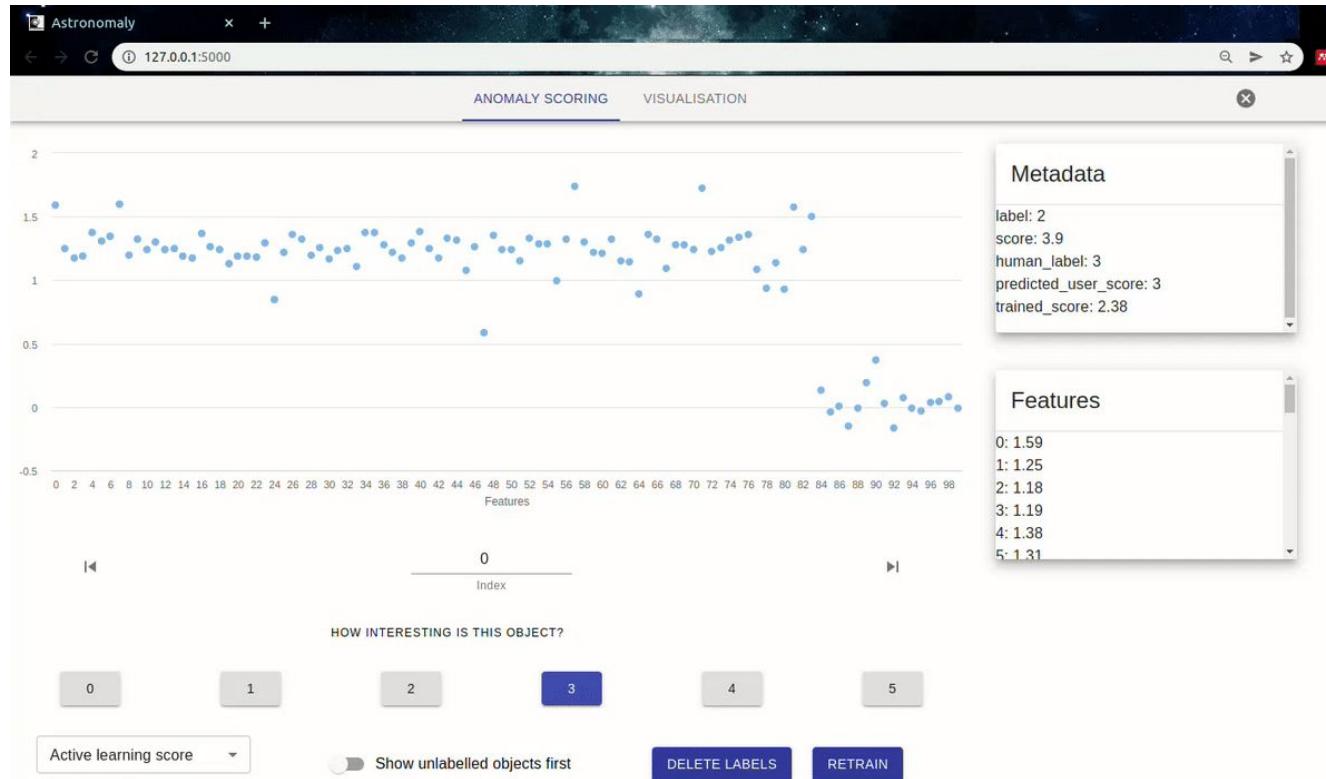
# Synthetic data - Random



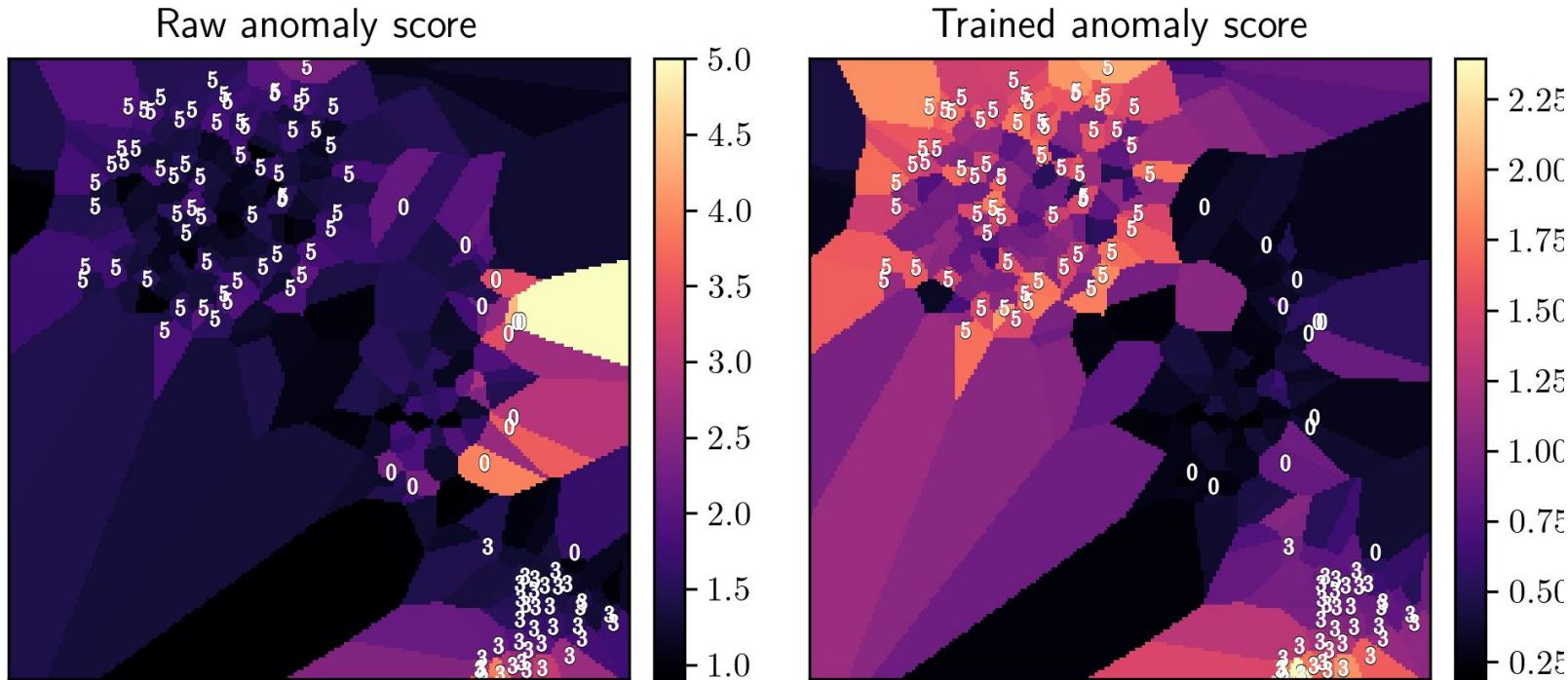
# Synthetic data - No active Learning



# Synthetic data - Active Learning

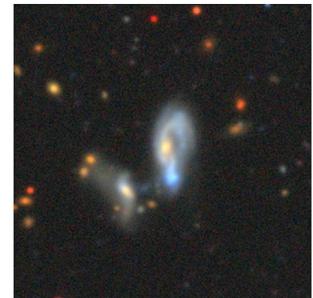
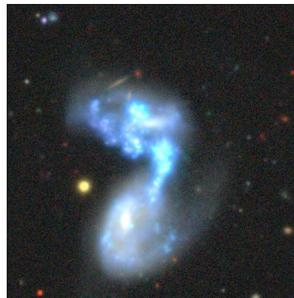
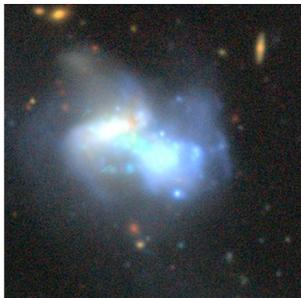
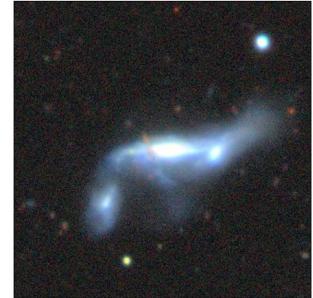
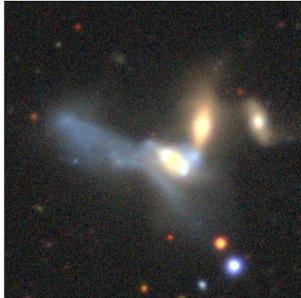


# Visualisation with Synthetic Data

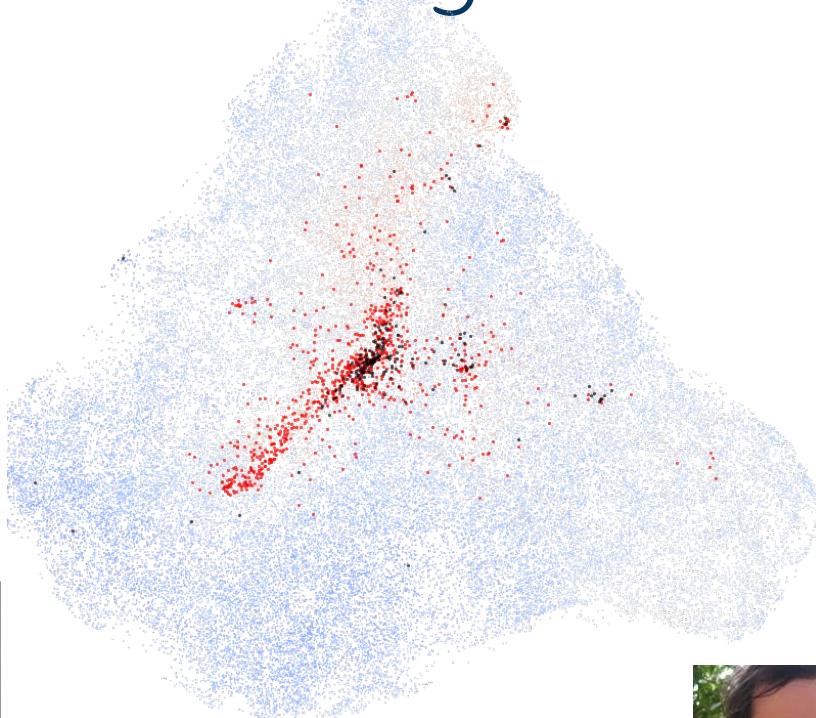
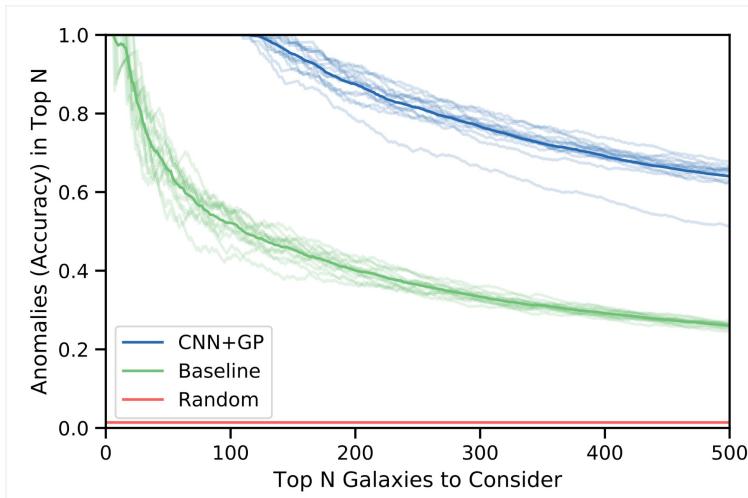


# Anomalies in DECaLS

Verlon Etsebeth (MSc student)



# Deep Representation Learning



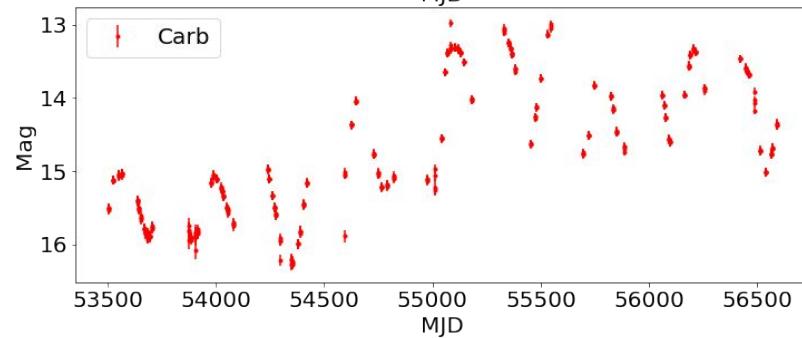
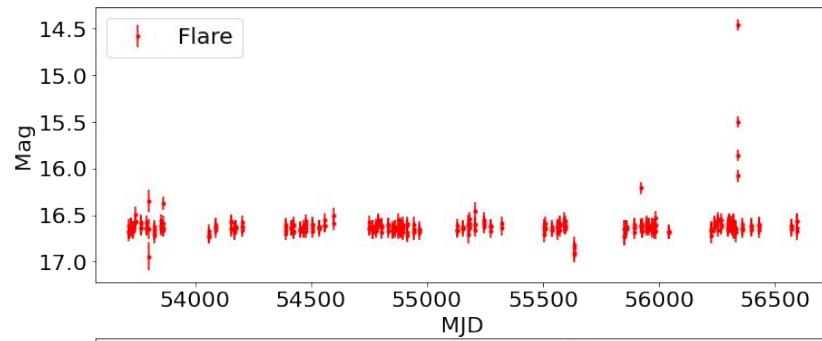
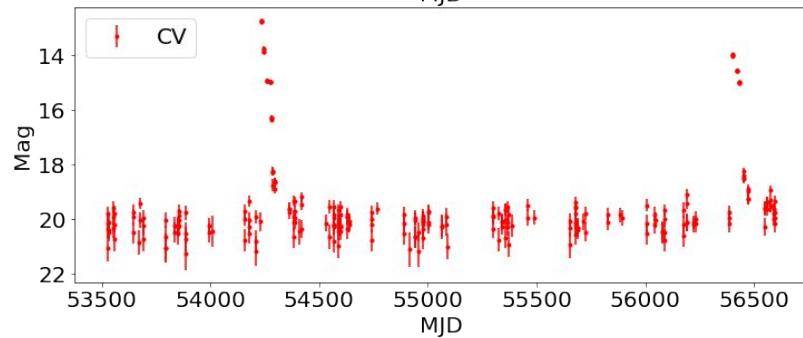
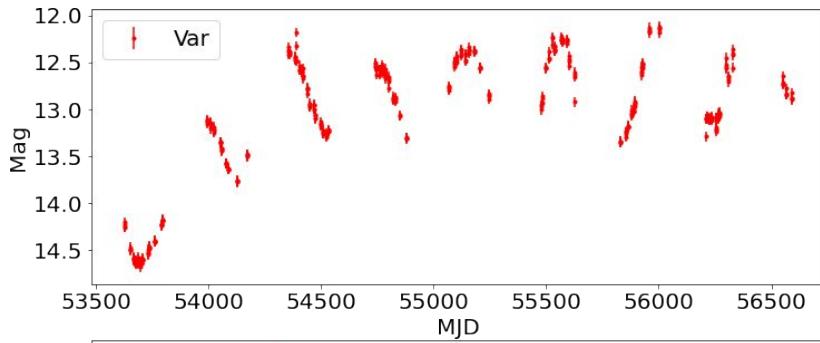
Walmsley et al. (2021) [2110.12735](https://arxiv.org/abs/2110.12735)



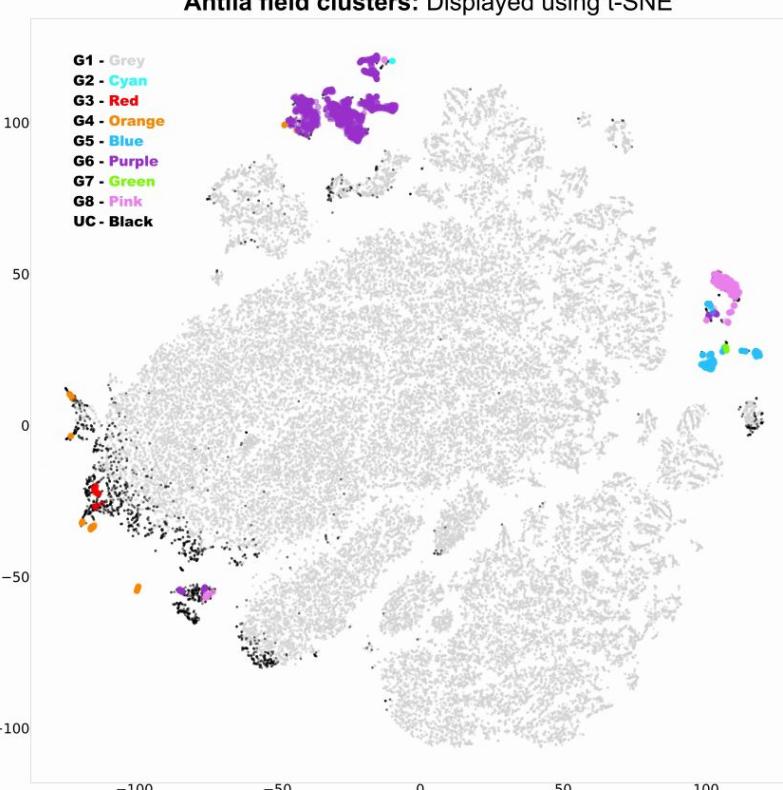
# Anomalous Transients



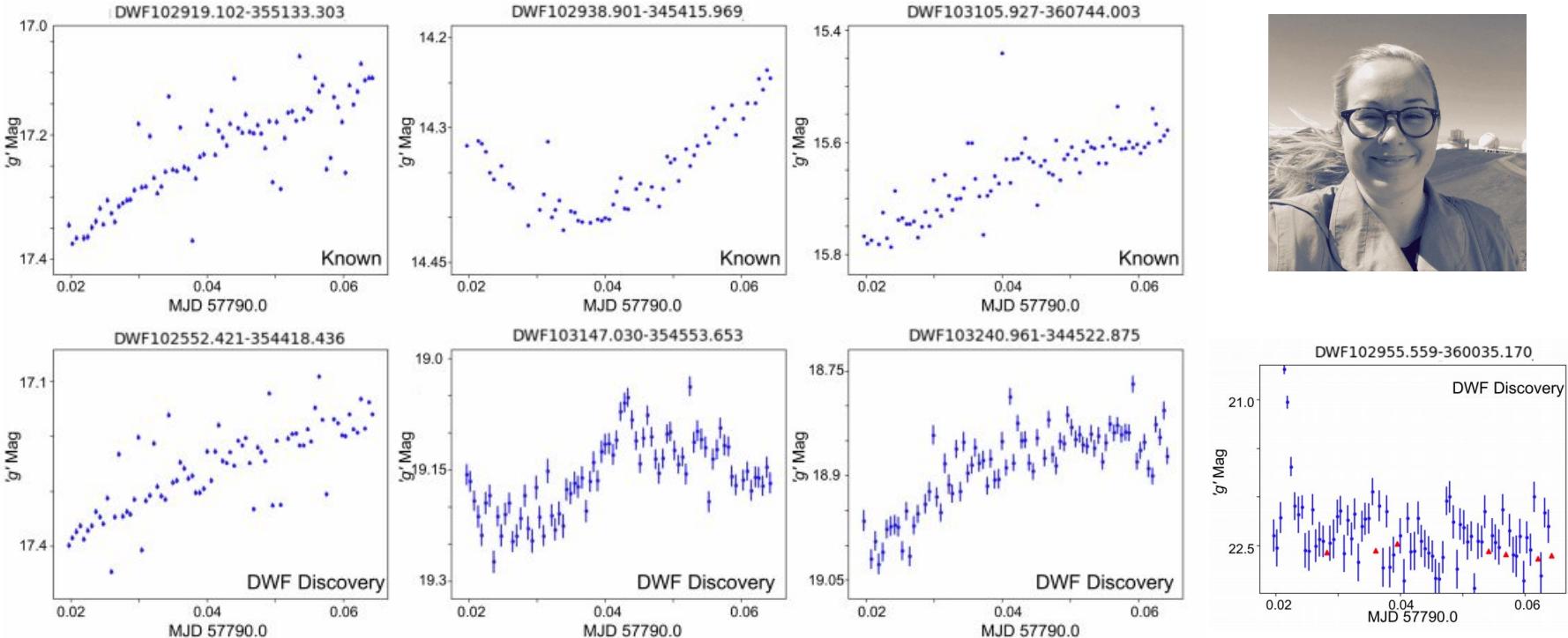
Malema Ramonyai (MSc student)



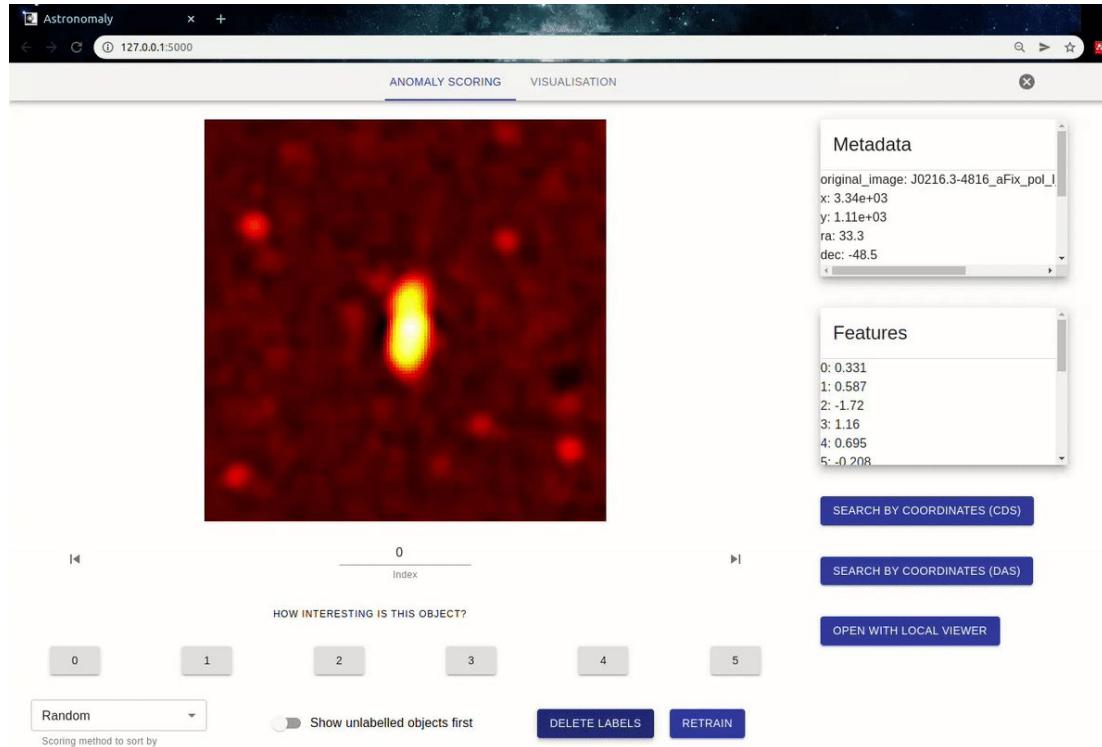
# Astronomaly Applied to DWF



# Astronomaly Applied to DWF

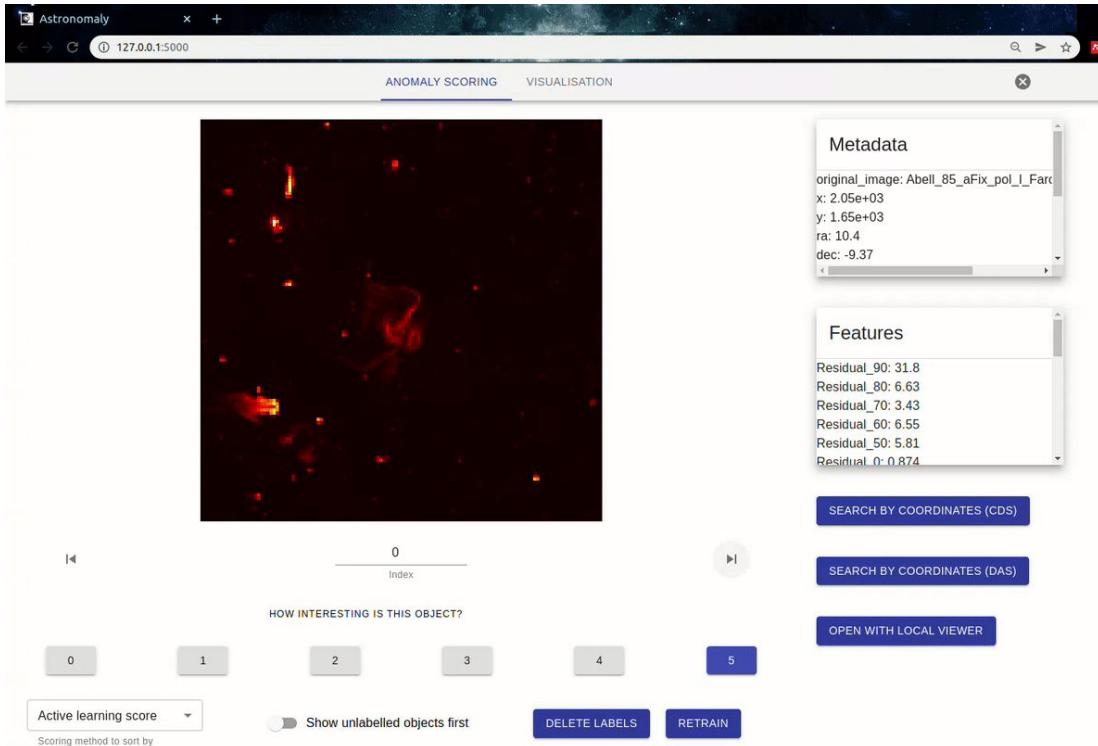


# MeerKAT Data - Random Examples



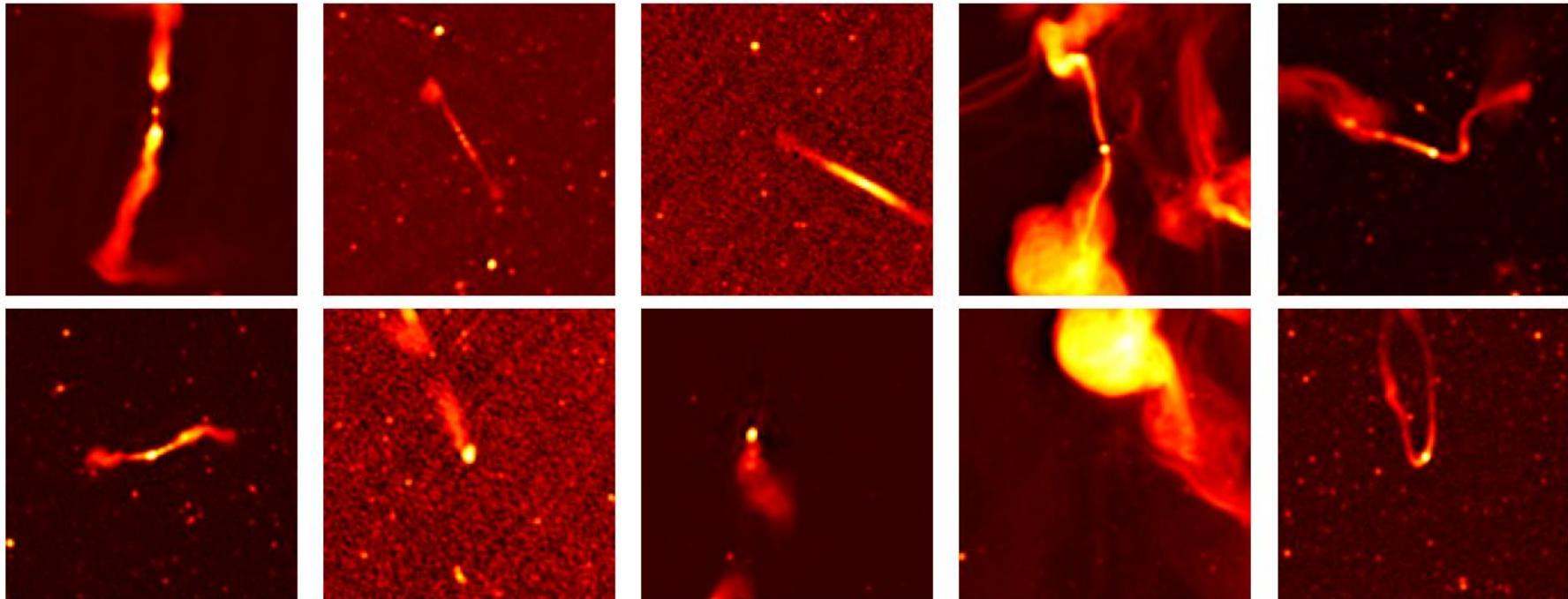
MGCLS data from Knowles, K et al. (2021) [2111.05673](#)

# MeerKAT Data - Anomalies

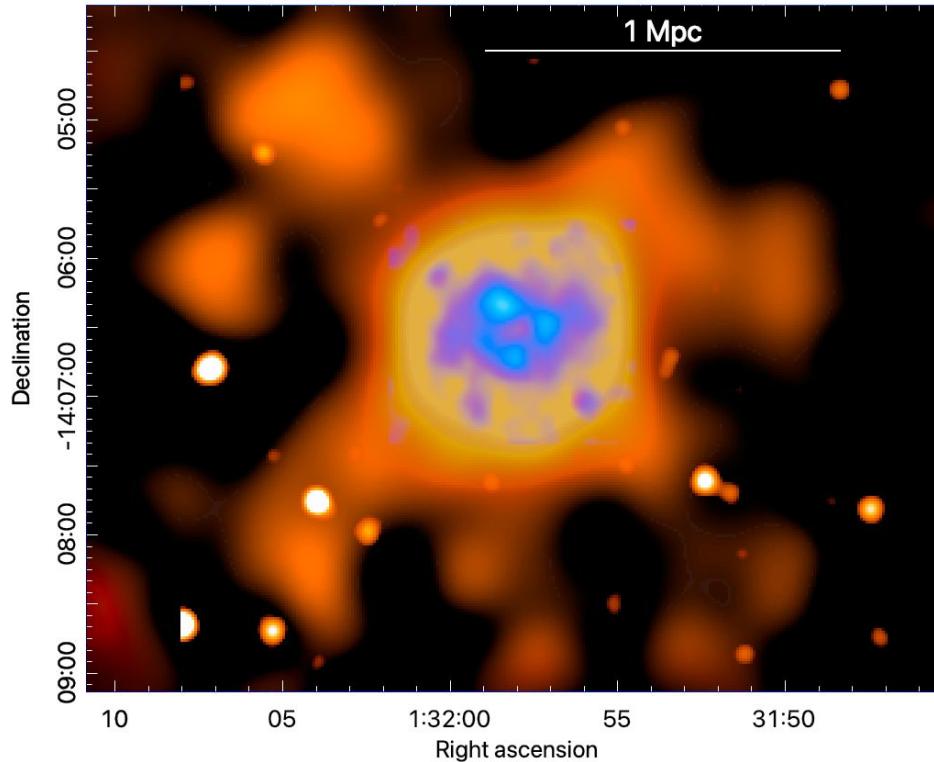


MGCLS data from Knowles, K et al. (2021) [2111.05673](#)

# MeerKAT Galaxy Cluster Legacy Survey



# A genuine machine learning discovery



# THE SUPERNOVA FOUNDATION

Mentoring For Women In Physics

SANGEETA UJJWAL



Country: India  
Research Field: Physics  
Institute: University of Delhi  
Position: Postdoc

SATYA GONTCHO A GONTCHO



Country: United States of America  
Research Field: Astrophysics and Cosmology  
Institute: Lawrence Berkeley National Laboratory  
Position: Project Scientist

SHAZRENE S. MOHAMED



Country: South Africa  
Research Field: Astronomy  
Institute: South African Astronomical Observatory and University of Cape Town  
Position: Faculty

SIPHEPHILE NCUBE



Country: South Africa  
Research Field: Physics  
Institute: University of the Witwatersrand  
Position: Postdoc

SIYI ZHOU



Country: Sweden  
Research Field: Theoretical Physics  
Institute: Stockholm University  
Position: Postdoc

SUDESHNA BORO SAIKIA



Country: Austria  
Research Field: Astrophysics and Cosmology  
Institute: University of Vienna  
Position: Post doctoral researcher

SWARNAMALA SIRSI



Country: India  
Research Field: Theoretical Physics  
Institute: University of Mysore  
Position: Associate Professor (retired)

VALERIA PETTORINO



Country: France  
Research Field: Astrophysics and Cosmology  
Institute: CEA  
Position: STAFF Scientist

# Conclusions

- Machine learning is critical in facing the data deluge
- We need automated anomaly detection to ensure scientific discoveries in datasets aren't missed
- Check out Astronomaly:
  - <https://arxiv.org/abs/2010.11202>
  - <https://github.com/MichelleLochner/astronomaly>
- And the Supernova Foundation:
  - <https://www.supernovafoundation.org/>