## Machine Learning & Anomaly Detection

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# Supervised Machine Learning



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

https://towardsdatascience.com

# Transient Classification



Lochner et al. (2016) - 1603.00882

# Transient Classification



#### **Feature Extraction**

Lochner et al. (2016) - 1603.00882

# Machine Learning



 $\tanh\left[w_{5} \tanh(w_{1}i_{1} + w_{2}i_{2}) + w_{6} \tanh(w_{3}i_{1} + w_{4}i_{2}))\right]$ 

# Transient Classification



# Multiwavelength Transient Classification

(a) Confusion matrix without optical feature



(b) Confusion matrix showing the difference when optical feature is added

## Online Learning with Brokers



E.g: ANTARES (Saha et al. - 1409.0056,

Narayan et al. - 1801.07323)

#### Science with Imperfect Classification



Bayesian Estimation Applied to Multiple Species (BEAMS) Kunz, Bassett & Hlozek - 0611004 Newling et al. - 1110.6178 Hlozek et al. - 1111.5328 Lochner et al. - 1205.3493 Roberts et al. - 1704.07830

# Anomaly Detection

#### Known Unknowns - rare events



Caltech/MIT/LIGO Lab

#### Unknown Unknowns - new anomalies



Daily Herald Archive / SSPL / Getty Images

#### How do we discover new phenomena...

### ...among 10 million possibilities?

# Unsupervised Learning



The farther away from normal the higher the anomaly score

### Anomaly Detection Algorithms



representative points

### Anomaly Detection Isn't Enough



# Human-in-the-loop Learning



www.clickworker.com



ANOMALY SCORING CLUSTERING



#### Object ID: 6812



Anomaly Score: 4.766405

Lochner and Bassett (in prep)

ANOMALY SCORING CI

CLUSTERING



#### **Object ID: 653**



Anomaly Score: 0.884448

Lochner and Bassett (in prep)



Lochner and Bassett (in prep)



Lochner and Bassett (in prep)

Anomaly Score: 1.303772







## Astronomaly Applied to DWF



Webb et al. (in prep)

## Real-time anomaly detection

Muthukrishna et al. (in prep)

#### Real-time anomaly detection



Muthukrishna et al. (in prep)

### Real-time anomaly detection



Muthukrishna et al. (in prep)

# Talking Points

- No ML is perfect how do we as a community move towards dealing with probabilistic catalogues for things like SN cosmology?
- Non-representivity remains a huge problem for ML algorithms how do we build better training sets?
- How do we avoid a high false positive rate of anomaly detection in the low SNR regime?
- When 10 000 anomalous objects are detected each night, how do we coordinate the community?
- How do we use ML to automatically follow-up targets intelligently?