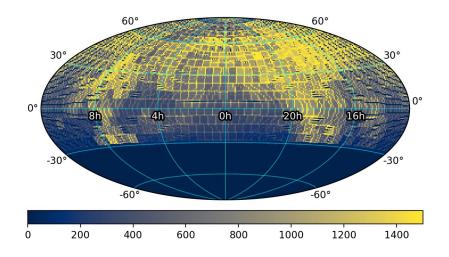
# Synergy of Gaia with ZTF and other variability surveys





Ashish Mahabal Center for Data Driven Discovery, Caltech MW-GAIA WG2, Sofia, 2023-06-08

## Outline

- Gaia observing card:
- Cadence, depth, filters, resolution, spectroscopy
  - Nominally 70 pointings on average in original
     5-year schedule
  - **BP/RP low res spectra**
- Strengths: Astrometry, spectra for "nearly" everything
- Limitations: 1D data sent down, not everything captured, epochs few for many purposes (which?)

# Synergy considerations

**Complementary depth** 

**Complementary filters/wavelengths** 

**Complementary cadence** 

Owing to its astrometry it has been incorporated into TESS pipelines for separating binary stars from exoplanet candidates (Nigraha)

**FOV - GW counterparts** 

Here we are more concerned about stellar variability

# Some other surveys

Pan-STARSS - deeper, many colors, few epochs

CSS/CRTS - no color info (but long time baseline)

ATLAS

ASAS-SN

Evryscope/Argus

TESS - transits; photometry for isolated stars. Some possibilities demonstrated in this meeting.

Plack Cor

**Black Gem** 

ZTF

LSST

# Summary of Gaia/ZTF seminar series

#### Preliminary talk schedule

All talks are on Tuesdays at 5 pm (CET)

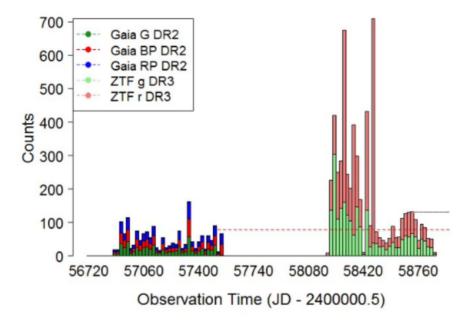
	start (CA)	start (Europe)		
March 16	9 am	5 pm	Laurent Eyer (University of Geneva)	General presentation on Gaia (recording)
March 30	8 am	5 pm	Matthew Graham (Caltech)	Overview of ZTF (capabilities, data products,)
April 13	8 am	5 pm	Laurent Eyer	Special focus on white dwarfs detected by Gaia
April 27	8 am	5 pm	Eric Bellm	ZTF surveys, past and present
May 11	8 am	5 pm	Nami Mowlawi	Large Amplitude variables with Gaia DR2 data (recording)
May 25	8 am	5 pm	Jan van Roestel, Ashish Mahabal	Classification of ZTF variable sources, current status and plans
June 8	8 am	5 pm	Marc Audard	Outliers in Gaia data and validation of the Gaia catalogue of variable sources
June 22	8 am	5 pm	Dan Perley	The ZTF Bright Transient Survey
			Yuhan Yao	Tidal disruption events with ZTF
September 7	8 am	5 pm	Krzysztof Nienartowicz	Data Handling of the Gaia data at the Geneva data processing center
September 21	8 am	5 pm	Kevin Burdge	White dwarf binaries from ZTF
October 5	8 am	5 pm	Panos Gavras (ESA)	Variability detection of the Gaia time series
October 19	8 am	5 pm	Ilaria Caiazzo	Massive, magnetized, fast rotating white dwarfs from ZTF
November 2	9 am	5 pm	Lorenzo Rimoldini	The classification of variable sources in the Gaia consortium
November 16	8 am	5 pm	Przemek Mróz, Antonio Rodriguez	Image difference photometry and the search for microlensing events
December 14	8 am	5 pm	Berry Holl	Features of the Gaia scanning law

# ZTF - Gaia synergy

From Eyer Gaia DR2 DR3 ZTF

With Gaia DR3 close to start of ZTF (2018), and with ZTF DR17, far ahead in time

In a way, ZTF is look-ahead for unreleased Gaia data!





- 1.2m automated Telescope @ Palomar, CA
- 47 deg2 FOV
- mlim~20.5 in 30 sec exposures
- g, r, i filters
- 1.4 TB data nightly
- ~20000 sq deg every 2 nights in g and r







# **EXAMPLE 7 EXAMPLE 1 CONTRACT STREET FOR THE STREET FOR THE STREET FOR THE DESCRIPTION OF THE DYNAMIC SKY**

Shri Kulkarni

#### PI:

Co-PI: Project Scientist: Survey Scientist: Project Manager: Lead Camera Engineer: P48 Operations: Data Archive Director: Science Data System Lead: Machine Learning Lead: Data Quality Scientist:

Thomas Prince, Mansi Kasliwal, Matthew Graham, Richard Dekany Matthew Graham Eric Bellm Richard Dekany Roger Smith Tom Barlow George Helou Ben Rusholme Ashish Mahabal Andrew Drake

### + real stars

#### Ashish Mahabal



### **DR16:**

#### https://irsa.ipac.caltech.edu/data/ZTF/docs/releases/ztf\_release\_notes\_latest

Filter(s)	#PSFcat- <i>sci</i> sources	#Aperturecat- <i>sci</i> sources	#PSFcat- <i>ref</i> sources	#Aperturecat- <i>ref</i> sources
g	179,661,606,131	114,144,715,027	2,527,614,585	794,988,671
r	506,885,000,022	315,778,300,845	3,393,409,691	1,153,239,938
i	58,976,476,111	33,687,367,330	1,414,109,235	455,279,857
g + r + i	745,523,082,264	463,610,383,202	7,335,133,511	2,403,508,466

Ashish Mahabal

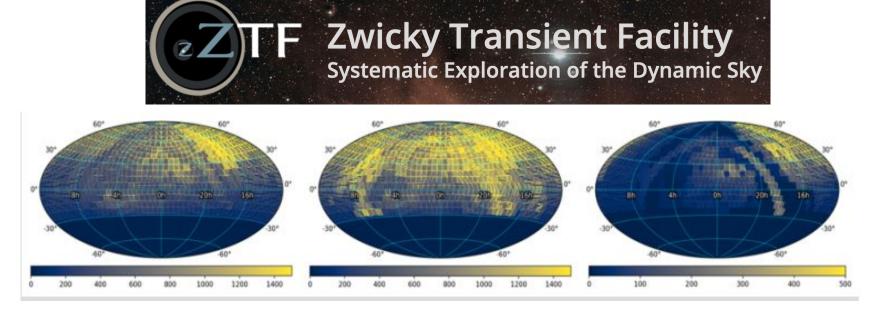


Figure 1 - Sky coverage and number of observation epochs in DR16 in g, r, i filters.

Example Query using the APIs

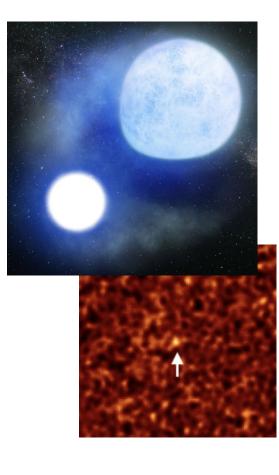
wget "https://irsa.ipac.caltech.edu/ibe/search/ztf/products/sci? POS=255.9302,11.8654&WHERE=obsjd>2458219.9678+AND+obsjd<2458228.8155+ AND+infobits<33554432" -O out.tbl

## Main science drivers of ZTF

A fast, wide-area time-domain survey:

- Fast, young, and rare flux transients
- Counterparts to gravitational wave sources
- Low-z Type Ia SNe for cosmology
- Variable stars & eclipsing binaries
- Solar System objects

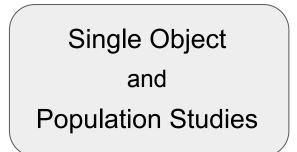
#### https://www.ztf.caltech.edu



Ashish Mahabal

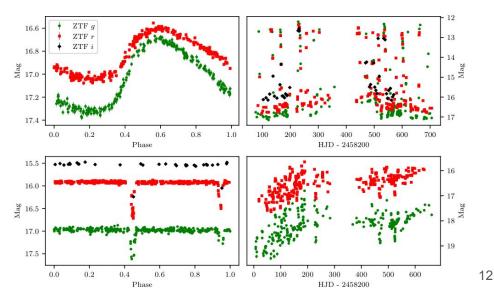
## Why and how do we use ZTF?

- 1. Stellar remnants
- 2. Binary star physics and evolution
- 3. Accretion processes
- 4. High energy astrophysics
- 5. Stellar structure
- 6. Extrasolar asteroids/comets
- 7. Age/luminosity/period relations



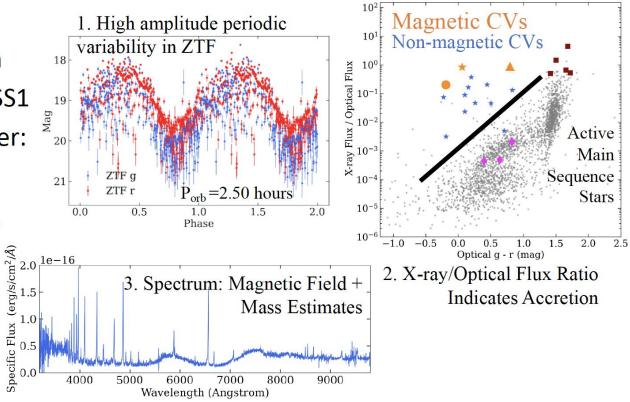
#### Zach Van der Bosch

- 1. Mostly **archival photometry** searches for (periodic) variability
  - a. Orbital periods
  - b. Rotations periods
  - c. Pulsations periods
  - d. Irregular dips/transits/eclipses
- 2. **Real-time (alert)** searches for outbursting stars



# Discovery of Polars from SRG/eROSITA + ZTF

- ~120 polars known
- ~200—300 in eRASS1
- Questions to answer:
  - X-ray luminosity function
  - Mean mass of WD
  - Orbital period distribution



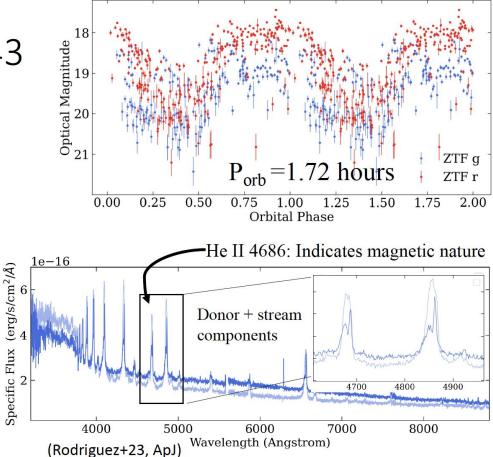
Tony Rodríguez (Caltech)

# eFEDS/ZTFJ0850+0443

- One of 8 polars showing pre-eclipse absorption from accretion stream.
- $M_{WD}$  = 0.81 ± 0.08  $M_{sun}$



Artist rendition, used with permission of M. Garlick



#### Spectroscopic Follow-up of NS and BH Candidates in Gaia DR3

Pranav Nagarajan, Kareem El-Badry, et al.

1.6

3456522854428709633

0.4

Primary Mass (Mo)

1.4

1.8

2.0

Period = 0.430924 days

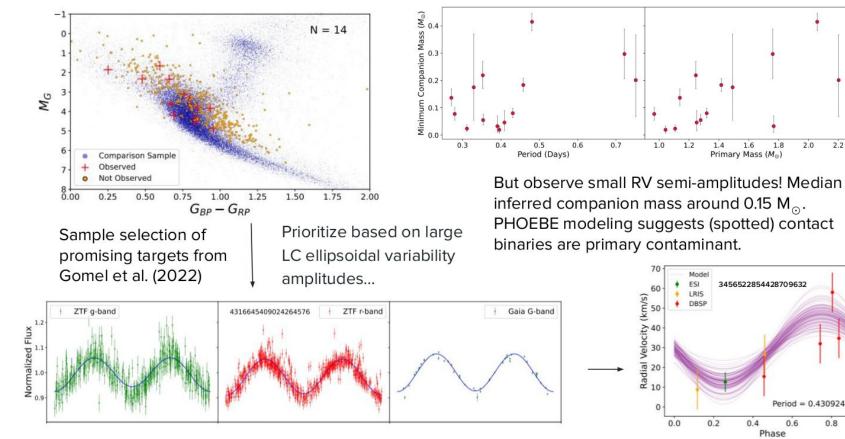
0.8

1.0

0.6

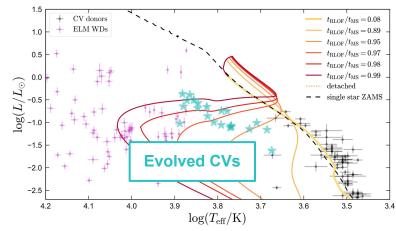
Phase

2.2

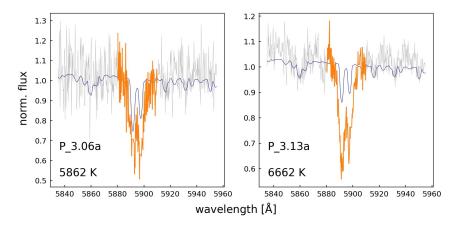


### **Sodium enhancement in Evolved Cataclysmic Variables**

Natsuko Yamaguchi, Kareem El-Badry, Antonio C. Rodriguez, Maude Gull, Benjamin R. Roulston



**1)** 21 evolved CVs found by the Birth of the ELMs survey, with the use of ZTF light curves.

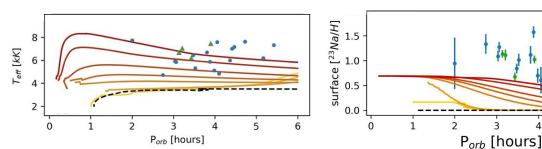


**2)** Carried out follow-up high resolution spectroscopy and measured Na abundances using the 5900 AA doublet.

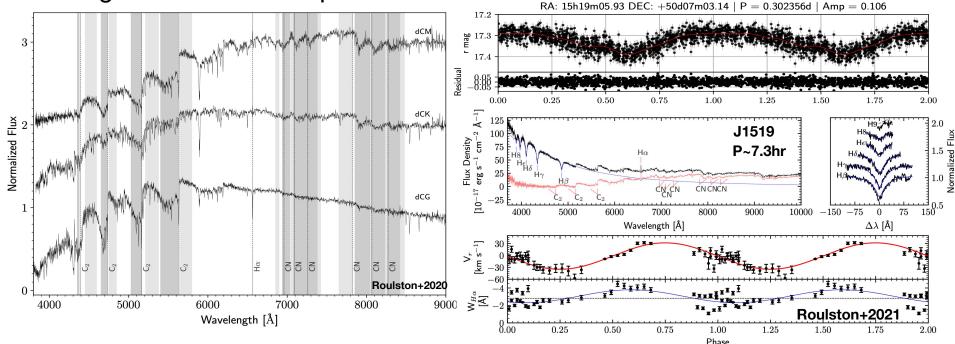
 $\rightarrow$  find significant enhancements:

[Na/H] = 0.3 - 1.5 dex, with a median of 0.956 dex

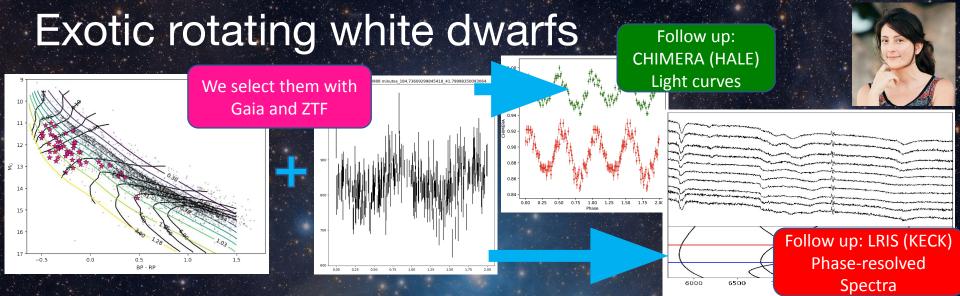
 3) Ran MESA models of evolved CVs
 → predict significant Na enhancement not seen in normal CVs but underpredict them compared to observations



#### Probing Common-Envelope Evolution with **Dwarf** Carbon stars Ben Roulston (Caltech)



- Main-Sequence stars with C/O>1, enhanced by a binary companion. Show strong carbon molecular bands in optical spectra
- 34 periodic dCs in ZTF with P<2d (down to P~2hrs) —> Post common-envelope binaries
- Would like to expand sample of known dCs using Gaia+? (SEDMv2?) then search for periods in ZTF



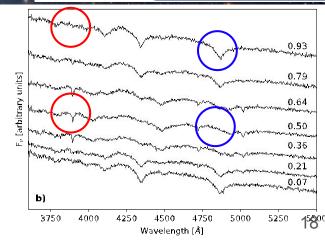
A highly magnetized and rapidly rotating white dwarf as small as the Moon

Finding a population of rapidly rotating and highly magnetized WDs

- Candidates double white dwarf mergers
- Reveal the characteristics of mergers and constrain merger rates
- Allow to study magnetic WDs

Finding exotic white dwarfs

Janus: double-faced white dwarf



#### **ZTF and Globular Clusters**

Chow-Choong Ngeow (NCU-Taiwan), et al.

- Goal: calibrate various old population distance indicators in gr(i)-band, most of them for the *first time*  $\rightarrow$  can be applied in, e.g. LSST, HSC-SSP, etc surveys observed with gri filters
- Why G.C.? Good → well-determined (and homogeneous) distance, most with low or vanished extinction, some rich in variable stars; Bad → blending (need PSF photometry + small pixel scale)
- Why ZTF? ZTF out-number PS1 in terms of number of observations!

 Table 1. Comparison of optical time-domain surveys in the northern sky.

	Surveya	$_{\rm Filters}b$	Pixel Scale <sup>C</sup>	$_{\rm Photometry}d$	Depth
Best 📫	ZTF	gri	1.01	PSF & AP	$r\sim 20.6$
	PS1 $3\pi$	grizy	0.258	PSF & AP	$r\sim21.8$
	ATLAS	oc	1.86	PSF	$m\sim 19.5$
	ASAS-SN	gV	8.0	AP	$V \sim 17$
	CSS	2 <u></u>	1.5	AP	$V\sim 19.5$
	LINEAR		2.25	AP	$m\sim 18$
	SuperWASP		13.7	AP	$V \sim 15$

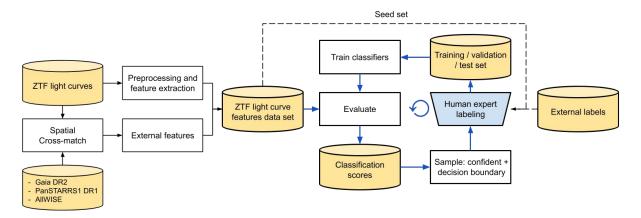
B

Distance Indicators	Publication	ZTF Data
Contact binaries	AJ 162:63 (2021)	DR 3 + private
RR Lyrae	AJ 163:239 (2022)	DR 7 + private
Type II Cepheids	AJ 164:154 (2022)	DR 10 + private
Yellow Post-AGB stars	AJ 164:166 (2022)	DR 10 + private
Anomalous Cepheids	AJ 164:191 (2022)	DR 11 + private
SX Phoenicis	AJ 165:190 (2023)	DR 13 + private
Miras	Work in-progress	DR16 + private
5 16 18 12 14 16 18 10 10	251.000000 days	ZTF II/III wish-list: ← more i-band data
K 12 14		← more i-band data
250 500 750 1000 <i>MJD</i> – 580	1250 1500 1750 20	

# Our big challenge:

# identify objects of interest from the 2 Billion sources

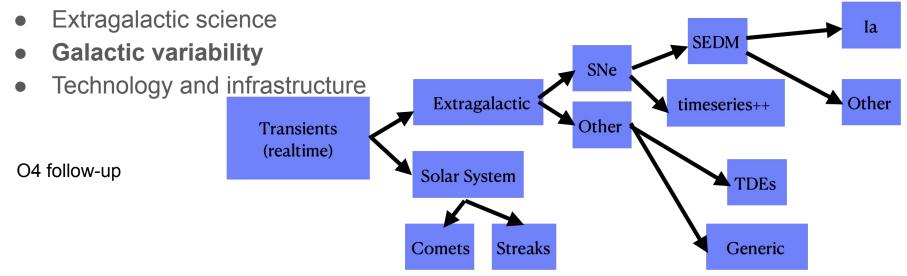
- External catalogues (Gaia, PS1, SRG, Fermi, SDSS-V, etc)
- ZTF alerts (positive and negative)
- Period searches & other **ZTF variability metrics**
- ZTF-SCoPe machine learning classification of persistent point sources



### Highlights - exemplars rather than exhaustive

Also likely biased towards areas I work in, or am better versed in

• Solar System



# Source Classification Project (SCoPe)

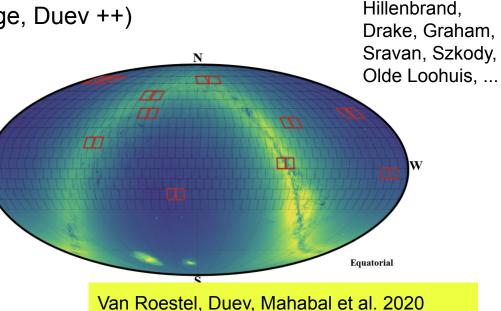


Van Roestel.

Duev, Coughlin,

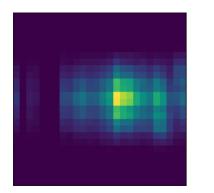
Mahabal, Mroz,

- Software/labeling set up based on DR2
- 20 Fields paper (Van Roestel, Duev, Mahabal ++)
- Periods paper (Coughlin, Burdge, Duev ++)
- 34M+ objects
- Features
  - variability characteristics
  - dmdt
  - period searches
  - external data



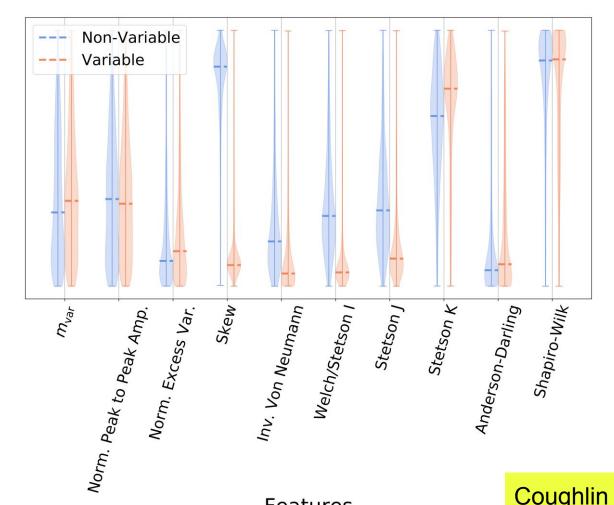
## Features

- Light-curves with 50+ epochs period-searched [GPUs + Kowalski]
  - Conditional Entropy (CE)
  - Lomb-Scargle (LS)
  - Analysis of Variance (AOV)
- Variability features (e.g.Sokolovsky+ 2016)
- Optimized dmdt's
- X-match: 10 catalogs



id: 10296001000057 ra: 19,556186750000002 dec: -19.040459849999998 period: 0.11717913349300235 significance: 9.32633052341019 pdot:0 n: 54 median: 15.3535 wmean: 15.353907407407405 chi2red: 1.7092470481841708 roms: 1.0754716981132069 wstd: 0.02201873656559632 norm peak to peak amp: 0.0023127789178800525 norm excess var: 8.306723761861014e-7 median\_abs\_dev: 0.014499999999999957 iar: 0.02875000000000497 f60: 0.034600000000001074 f70: 0.045099999999999696 f80: 0.0563000000000024 f90: 0.0680999999999976 skew: -0.24210490618175218 smallkurt: 5,6190133939968 inv vonneumannratio: 0.6095639426533307 welch i: 10.981297710846444 stetson j: 10.986989388536113

✤ SHOW 23 MORE FIELDS



Index	Statistic					
1	$N \over m_{ m median}$					
2						
3						
4	$m_{ m var}$					
5	$\chi^2$					
6	RoMS					
7	Median absolute deviation					
8	Normalized Peak to Peak Amplitude					
9	Normalized Excess Variance					
10-14	Ranges					
15	Skew					
16	Kurtosis Inverse Von Neumann Statistic Welch/Stetson I					
17						
18						
19	Stetson J					
20	Stetson K Anderson-Darling test					
21						
22	Shapiro-Wilk test					
23-35	Fourier Decomposition					

Features

Coughlin et al. arXiv:2009.14071

# Labels and classifiers

- Over 200,000 individual labeled light curves
  - Labels from CRTS, OGLE, ...
  - Unbalanced classes
  - Values quantized to [0, 0.25, 0.5, 0.75, 1] for label smoothing
- "Seed" classifiers, from subsets then active learning
  - Select and inspect <random | most confident | close to decision boundary | highest loss> predictions, label, add to training set, retrain, repeat

**ZTF** light curves

Spatial

Cross-match

Gaia DR2 PanSTARRS1 DR AllWISE Preprocessing and

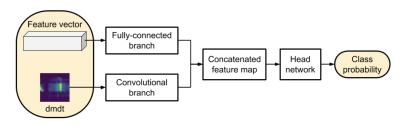
feature extraction

External features

ZTF light curve

features data set

- Marching down the nomenclature tree
- Classifiers
  - DNNs (MLP+CNN)
  - XGBoost
  - Hyperparameter tuning



Training / validation

/ test set

Human expert

labeling

Sample: confident +

decision boundary

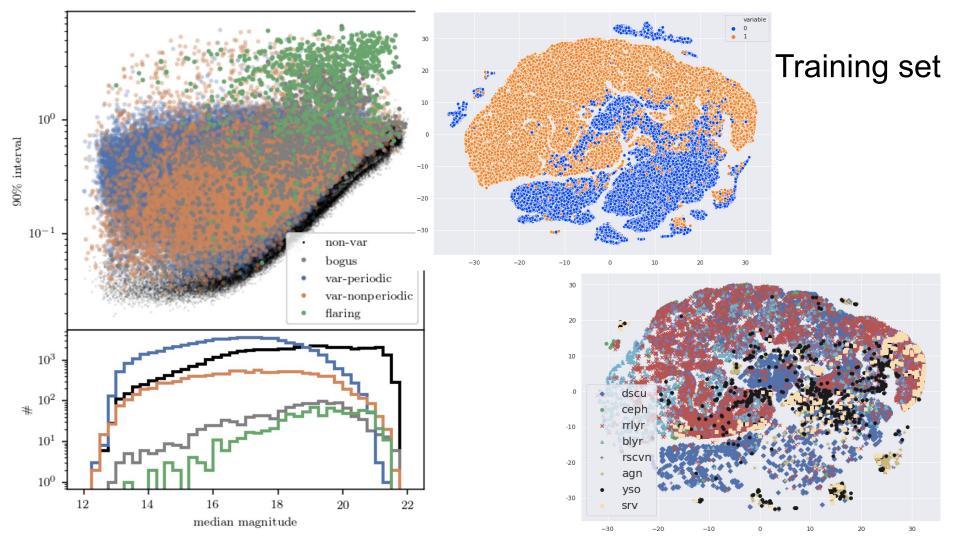
External labels

Train classifiers

Evaluate

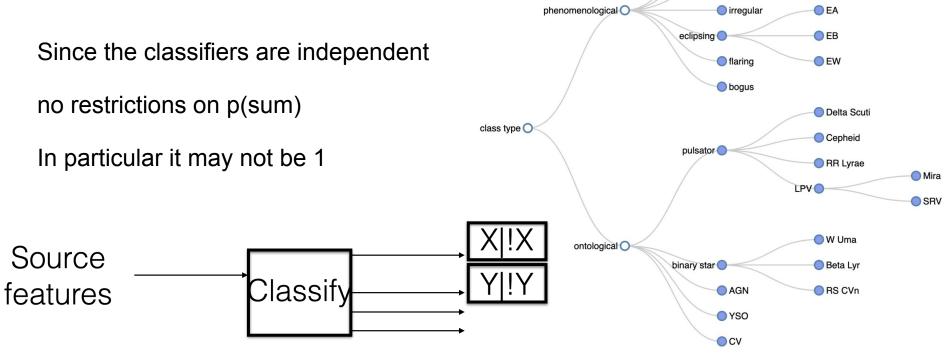
Classification

scores



# Hierarchical/stackable Classification Through Independent Binary Classifiers

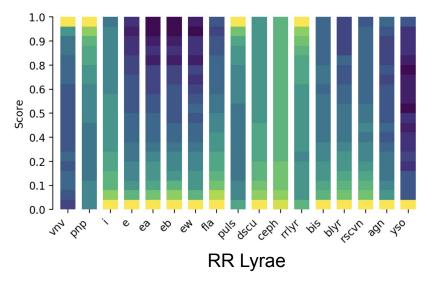
- Phenomenological: based on just the ZTF data
- Ontological: based on not just the ZTF data



periodic

Iong timescale

## Performance



Mahabal et al. 2017 http://arxiv.org/abs/1709.06257v1 Duev et al 2019 http://arxiv.org/abs/1904.05920v2 Duev et al. 2021 https://arxiv.org/abs/2102.13352 Coughlin et al. 2020 https://arxiv.org/abs/2009.14071 Van Roestel et al. 2021 https://arxiv.org/abs/2102.11304 Fremling 2021 https://arxiv.org/abs/2104.12980

Class	#	Accu	iracy	Prec	ision	Ree	call	F1 S	core
		DNN	XGB	DNN	XGB	DNN	XGB	DNN	XGB
е	44721	0.94	0.95	0.9	0.92	0.93	0.95	0.92	0.93
ea	819	0.94	1	0.91	1	0.87	0.02	0.89	0.03
eb	950	0.88	0.99	0.86		0.74	0	0.8	
ew	39079	0.94	0.95	0.91	0.92	0.89	0.93	0.9	0.92
fla	829	0.97	1	1	0.84	0.87	0.82	0.93	0.83
i	1842	0.93	0.99	0.92	0.79	0.84	0.28	0.88	0.42
longt	968	0.95	1	0.93	0.87	0.93	0.38	0.93	0.53
pnp	64910	0.95	0.95	0.95	0.95	0.96	0.96	0.96	0.95
vnv	78083	0.97	0.98	0.99	0.98	0.97	0.98	0.98	0.98
agn	608	0.98	1	0.94	0.94	0.98	0.71	0.96	0.81
bis	44532	0.95	0.96	0.92	0.93	0.93	0.96	0.93	0.94
blyr	836	0.89	0.99	0.8	0.46	0.81	0.9	0.81	0.61
$\operatorname{ceph}$	1075	0.93	1	0.88	0.76	0.89	0.92	0.89	0.83
dscu	6118	0.96	1	0.92	0.96	0.93	0.97	0.93	0.96
puls	18664	0.96	0.99	0.94	0.94	0.93	0.98	0.94	0.96
lpv	968	0.99	1	0.97	0.88	0.99	0.79	0.98	0.84
rrlyr	10866	0.95	0.99	0.93	0.95	0.89	0.95	0.91	0.95
rscvn	1210	0.85	1	0.83	0.77	0.68	0.82	0.75	0.8
$\mathbf{srv}$	420	0.95	1	0.88	0.81	0.98	0.69	0.93	0.74
yso	849	0.99	1	0.99	0.92	0.99	0.99	0.99	0.95

# Now classifying entire DR

- More classes
- Anomalies (hdbscan)

0

0

auc

group: r2g-500

150

200

250

Improvements (active learning)

37 classes

Step

300

- Metaclassification
- Interpretability

0.8

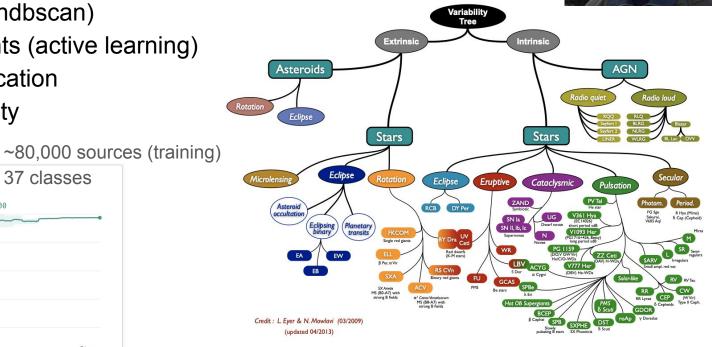
0.6

0.4

0.2

50

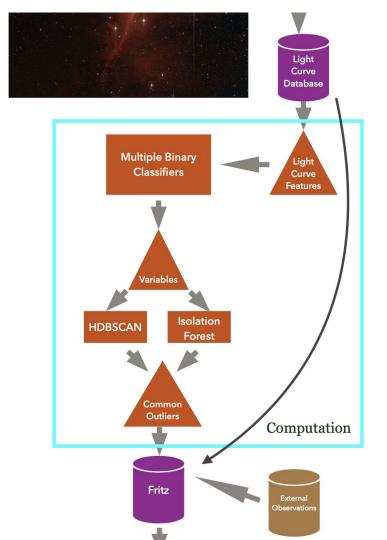
100



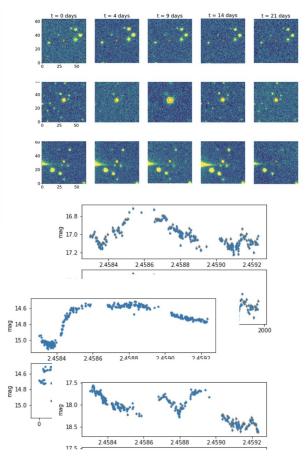
Brian Healy

UoMinnesota





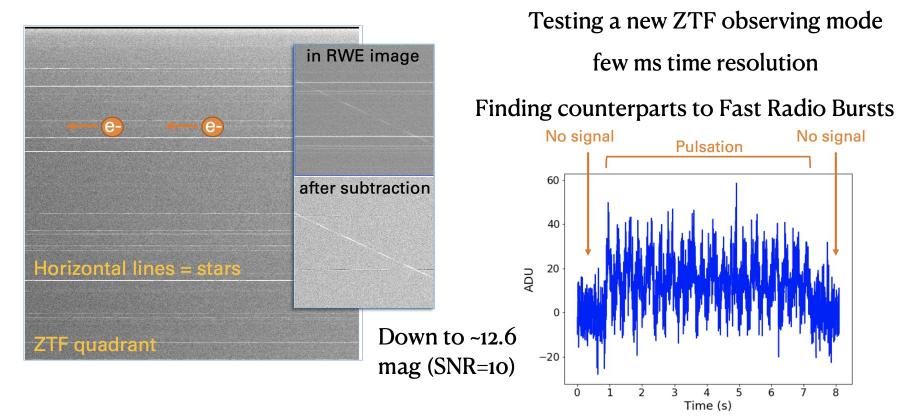
# No Anomaly Left Behind



With PPurohit, SParikh, YHassan, T Jegou Du Laz, ...

## **ZTF** news

- Public part funded until end of O4 (Dec 2024)
- CMOS possibilities (one problem non-buttable)
- RWE mode



#### **Read While Expose**

Rapidly spinning space debris The pulsating object was bright for ~6s

With Igor Andreoni, Roger Smith, ...

# ZARTH - Pokemon GO for ZTF transients

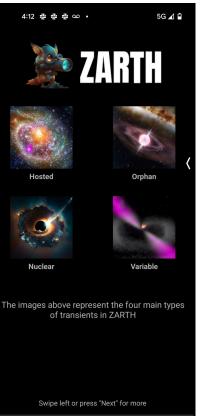
Coming this month to androids

With D Pindawala, A Arora, D Thummar, A Bhavsar, I Kostadinova, ...



Made with midjourney

Ashish Mahabal



NEXT



# Summary

- Combining archives generally underexplored
- Combined ZTF Gaia data have fantastic possibilities
- More ML and population level studies
- Move towards Rubin/LSST

Ashish Mahabal ashish at caltech.edu