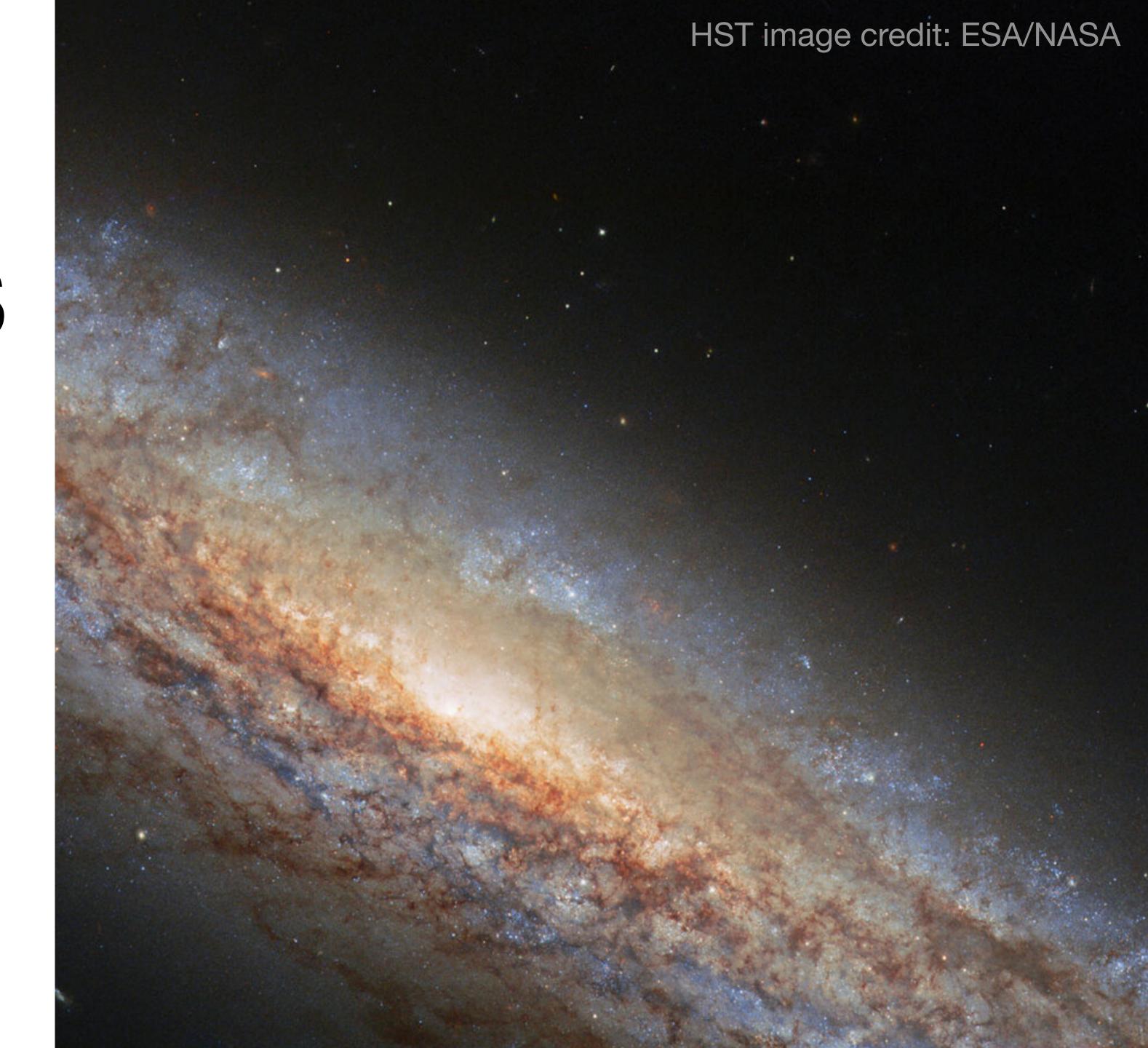
### Nuclear Transients in ZTF

O4 Workshop May 9, 2023





# It has been an exciting few years for the study of nuclear transients!

#### First ZTF TDE

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https://doi.org/10.3847/1538-4357/aafe0c



#### The First Tidal Disruption Flare in ZTF: From Photometric Selection to Multi-wavelength Characterization

```
Sjoert van Velzen<sup>1,2</sup>, Suvi Gezari<sup>1,3</sup>, S. Bradley Cenko<sup>3,4</sup>, Erin Kara<sup>1,3,5</sup>, James C. A. Miller-Jones<sup>6</sup>, Tiara Hung<sup>1</sup>,
     Joe Bright<sup>7</sup>, Nathaniel Roth<sup>1,3</sup>, Nadejda Blagorodnova<sup>8</sup>, Daniela Huppenkothen<sup>9</sup>, Lin Yan<sup>10</sup>, Eran Ofek<sup>11</sup>,
  Jesper Sollerman<sup>12</sup>, Sara Frederick<sup>1</sup>, Charlotte Ward<sup>1</sup>, Matthew J. Graham<sup>8</sup>, Rob Fender<sup>7</sup>, Mansi M. Kasliwal<sup>8</sup>,
 Chris Canella<sup>8</sup>, Robert Stein<sup>13</sup>, Matteo Giomi<sup>14</sup>, Valery Brinnel<sup>14</sup>, Jakob van Santen<sup>13</sup>, Jakob Nordin<sup>14</sup>, Eric C. Bellm<sup>9</sup>,
    Richard Dekany<sup>15</sup>, Christoffer Fremling<sup>8</sup>, V. Zach Golkhou<sup>9,16</sup>, Thomas Kupfer<sup>8,17,18</sup>, Shrinivas R. Kulkarni<sup>8</sup>, Russ R. Laher<sup>19</sup>, Ashish Mahabal<sup>8,20</sup>, Frank J. Masci<sup>21</sup>, Adam A. Miller<sup>22,23</sup>, James D. Neill<sup>8</sup>, Reed Riddle<sup>15</sup>, ,
           Mickael Rigault<sup>24</sup>, Ben Rusholme<sup>21</sup>, Maayane T. Soumagnac<sup>25</sup>, and Yutaro Tachibana (優太朗橘)<sup>26,27</sup>
                              Department of Astronomy, University of Maryland, College Park, MD 20742, USA; sjoert@astro.umd.edu
                                 <sup>2</sup>Center for Cosmology and Particle Physics, New York University, New York, NY 10003, USA
                                      <sup>3</sup> Joint Space-Science Institute, University of Maryland, College Park, MD 20742, USA
                          <sup>4</sup> Astrophysics Science Division, NASA Goddard Space Flight Center, MC 661, Greenbelt, MD 20771, USA
                               <sup>5</sup> X-ray Astrophysics Laboratory, NASA/Goddard Space Flight Center, Greenbelt, MD 20771, USA
                                              ICRAR—Curtin University, GPO Box U1987, Perth, WA 6845, Australia
                       <sup>7</sup> Department of Physics, University of Oxford, Denys Wilkinson Building, Keble Road, Oxford OX1 3RH, UK
                      <sup>8</sup> Division of Physics, Mathematics, and Astronomy, California Institute of Technology, Pasadena, CA 91125, USA
                  <sup>9</sup>DIRAC Institute, Department of Astronomy, University of Washington, 3910 15th Avenue NE, Seattle, WA 98195, USA
                               <sup>10</sup> The Caltech Optical Observatories, California Institute of Technology, Pasadena, CA 91125, USA
     <sup>11</sup> Benoziyo Center for Astrophysics and the Helen Kimmel Center for Planetary Science, Weizmann Institute of Science, 76100 Rehovot, Israel
                   <sup>2</sup> The Oskar Klein Centre & Department of Astronomy, Stockholm University, AlbaNova, SE-106 91 Stockholm, Sweden
                                         <sup>13</sup> Deutsches Elektronensynchrotron, Platanenallee 6, D-15738, Zeuthen, Germany
                                 <sup>14</sup> Institute of Physics, Humboldt-Universität zu Berlin, Newtonstr. 15, D-12489 Berlin, Germany
                                 <sup>15</sup> Caltech Optical Observatories, California Institute of Technology, Pasadena, CA 91125, USA
                                            The eScience Institute, University of Washington, Seattle, WA 98195, USA
                                <sup>17</sup> Kavli Institute for Theoretical Physics, University of California, Santa Barbara, CA 93106, USA
                                        <sup>18</sup> Department of Physics, University of California, Santa Barbara, CA 93106, USA
                           <sup>19</sup> Infrared Processing and Analysis Center, California Institute of Technology, Pasadena, CA 91125, USA
                                <sup>20</sup> Center for Data Driven Discovery, California Institute of Technology, Pasadena, CA 91125, USA
```

van Velzen+ 2019

#### 29 more



van Velzen+ 2019

van Velzen+ 2021

#### 29 more



van Velzen+ 2019

van Velzen+ 2021

Hammerstein+ 2021

#### A jetted TDE



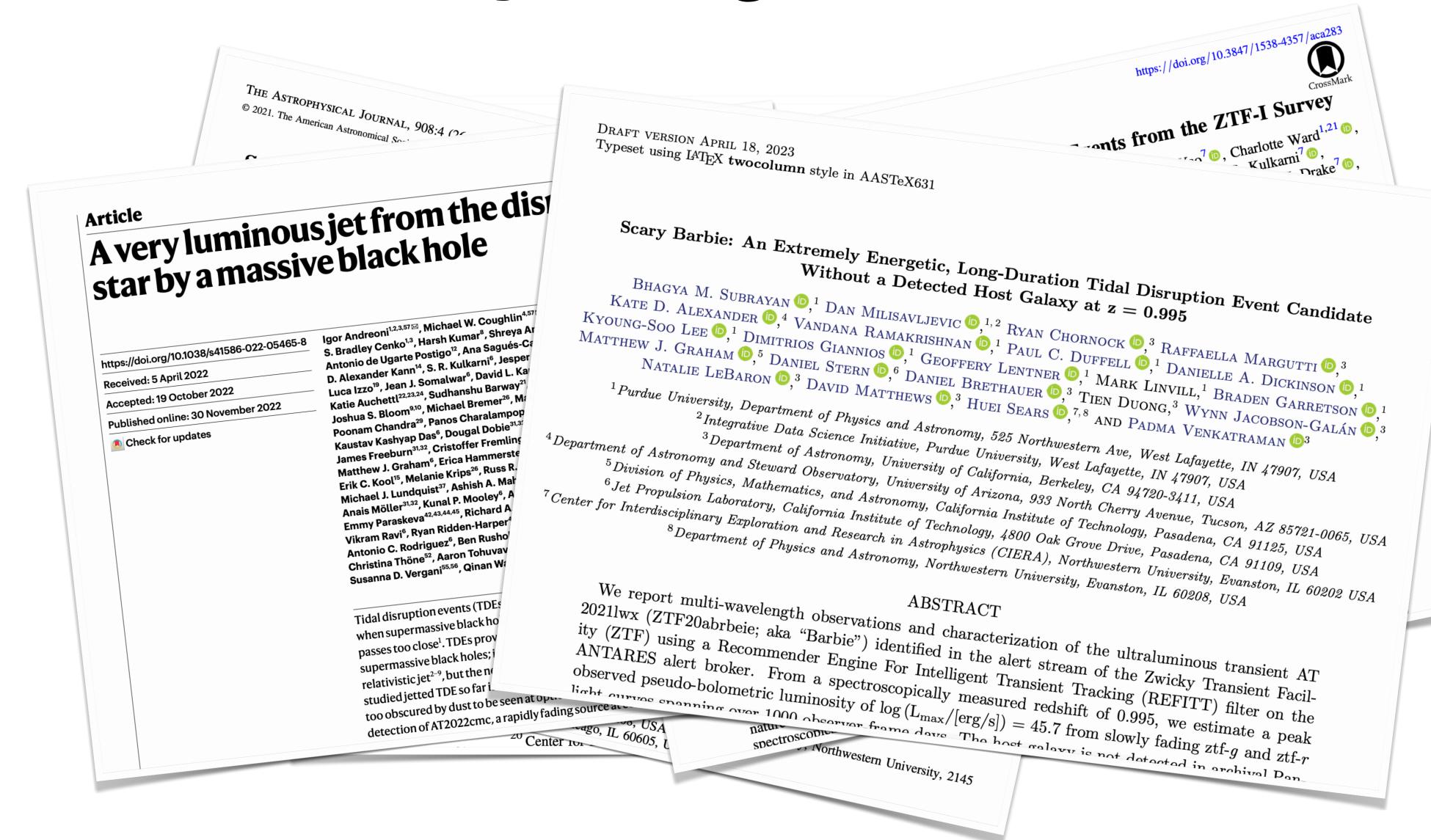
van Velzen+ 2019

van Velzen+ 2021

Hammerstein+ 2021

Andreoni+ 2022

#### An extremely energetic TDE



van Velzen+ 2019

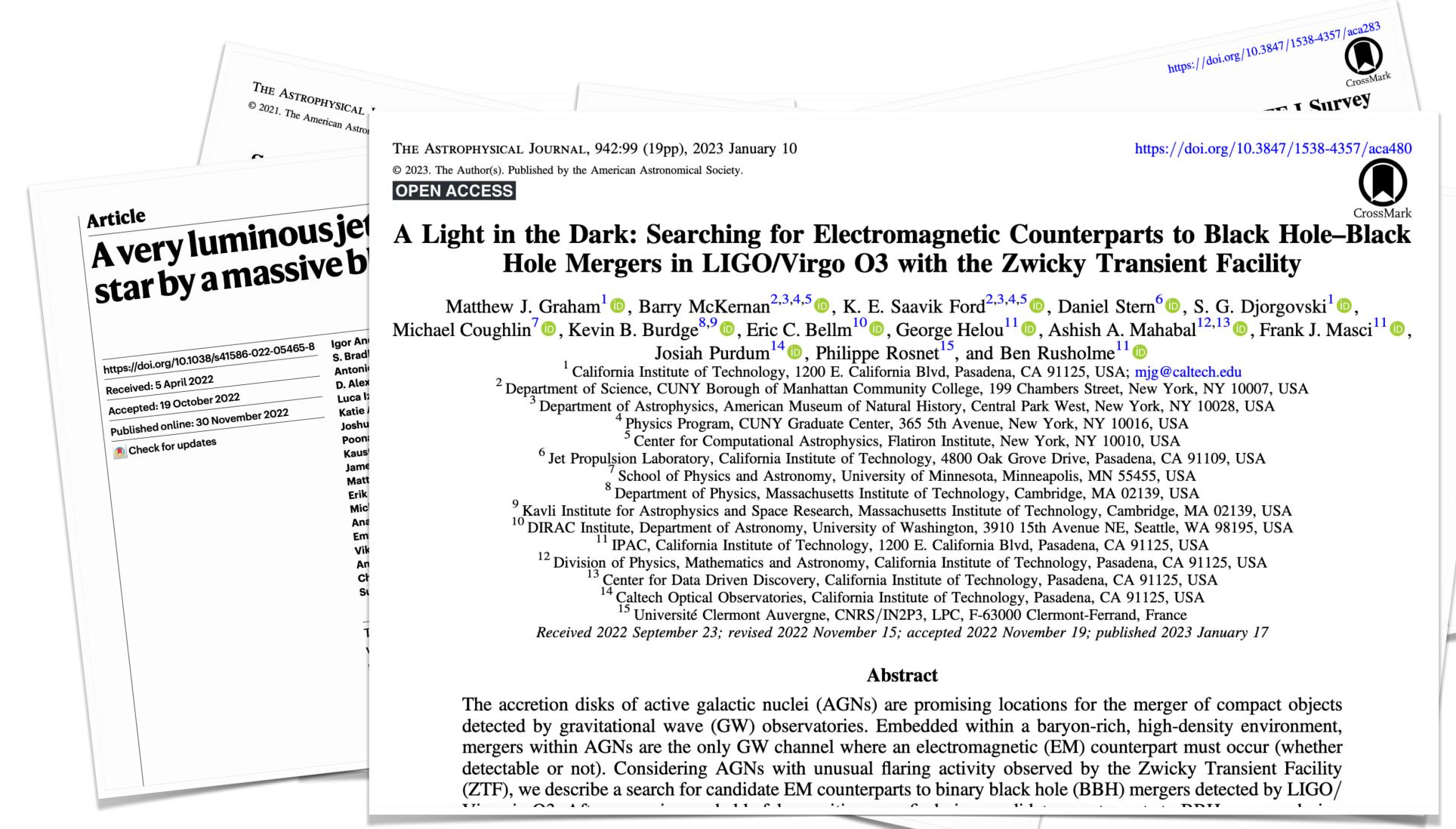
van Velzen+ 2021

Hammerstein+ 2021

Andreoni+ 2022

Subrayan+ 2023

#### Possible EM counterparts to BH-BH mergers



van Velzen+ 2019

van Velzen+ 2021

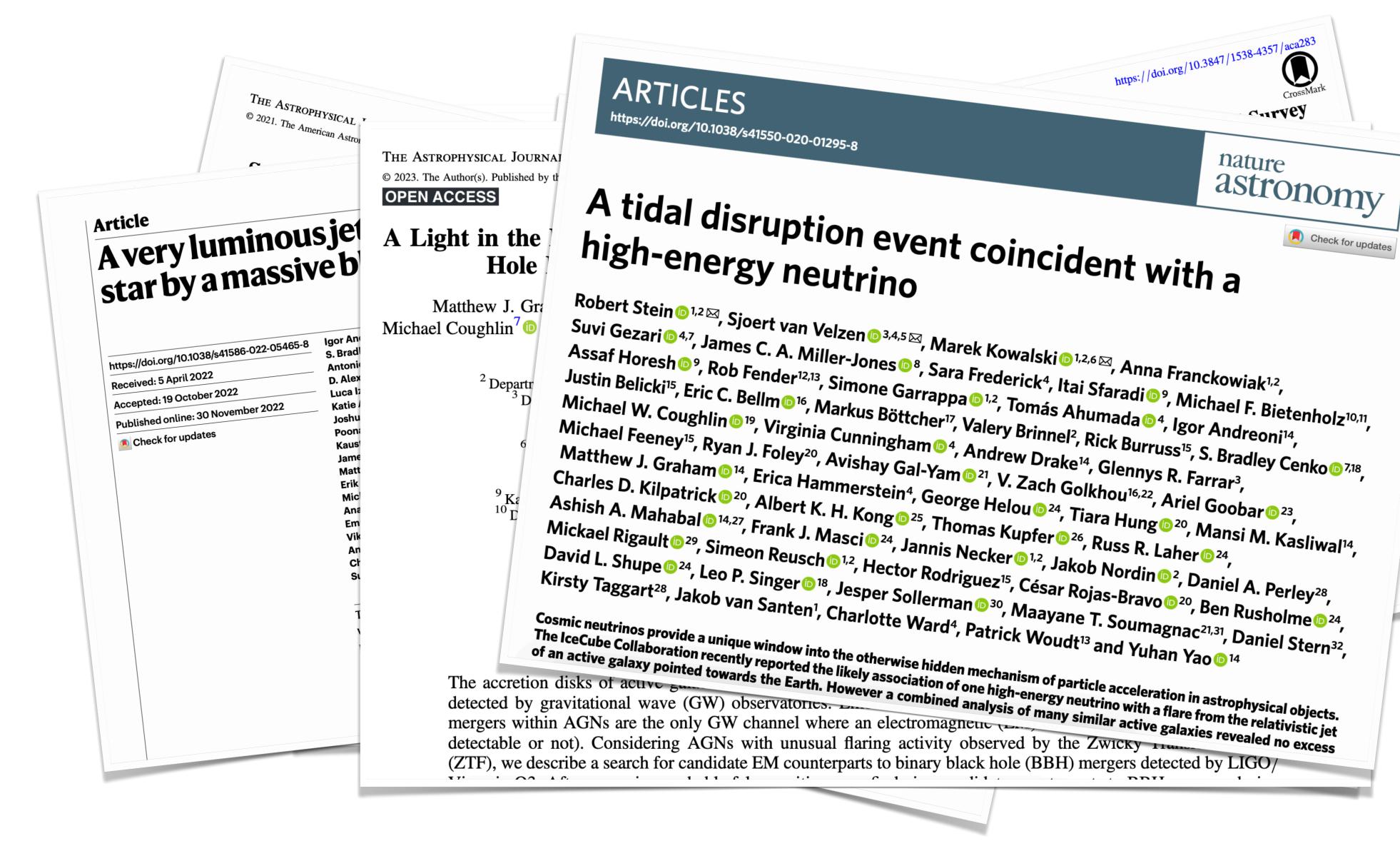
Hammerstein+ 2021

Andreoni+ 2022

Subrayan+ 2023

Graham+ 2023

#### Three neutrino-associated TDE candidates



van Velzen+ 2019

van Velzen+ 2021

Hammerstein+ 2021

Andreoni+ 2022

Subrayan+ 2023

Graham+ 2023

Stein+ 2021

#### Three neutrino-associated TDE candidates



van Velzen+ 2019

van Velzen+ 2021

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Andreoni+ 2022

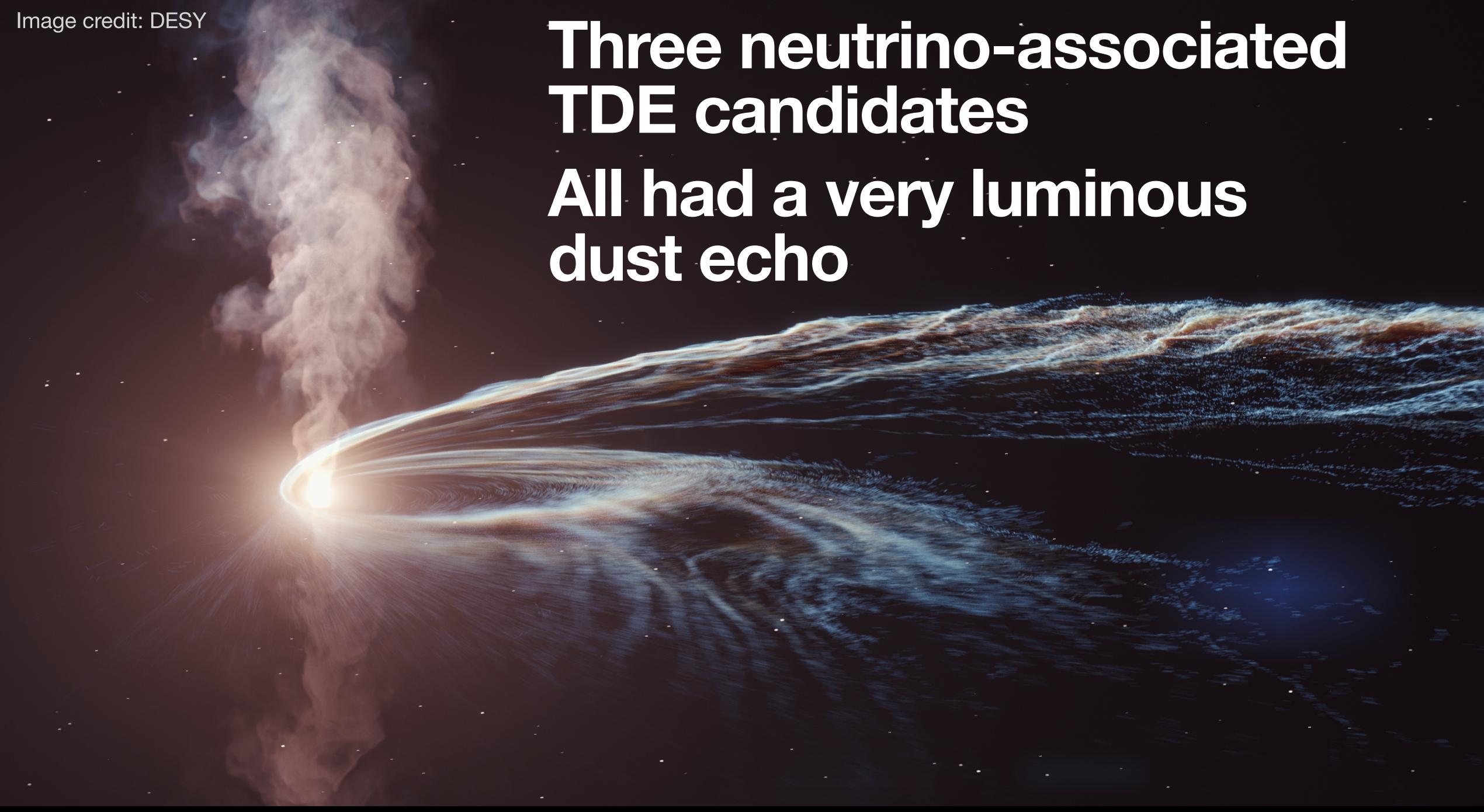
Subrayan+ 2023

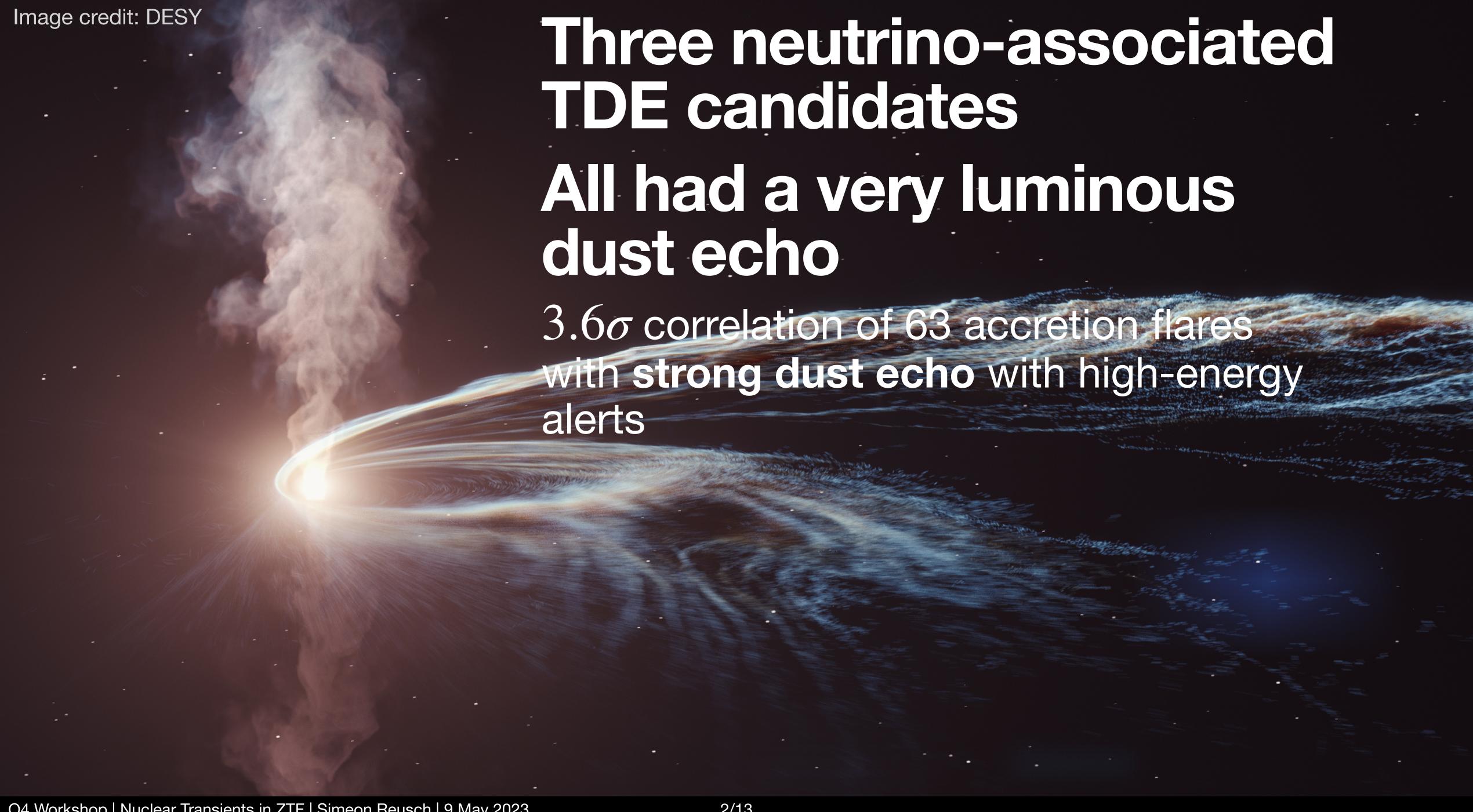
Graham+ 2023

Stein+ 2021

Reusch+ 2022





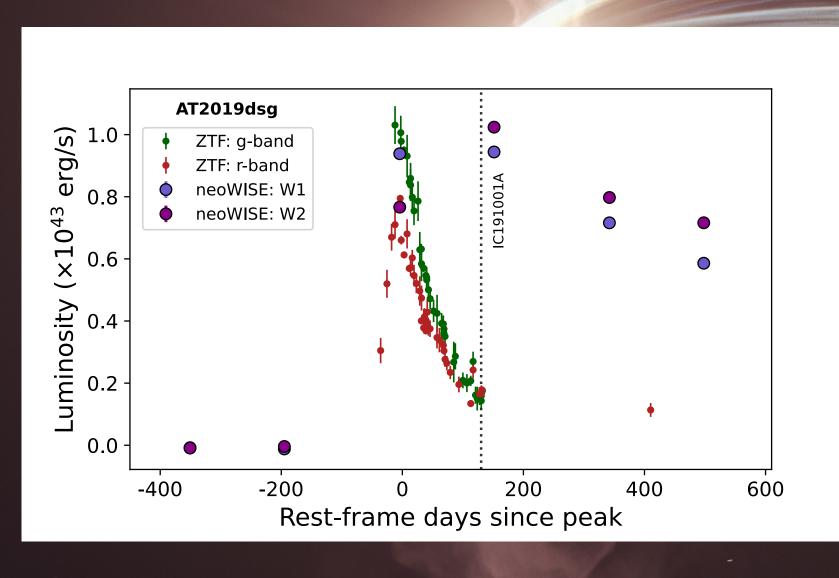


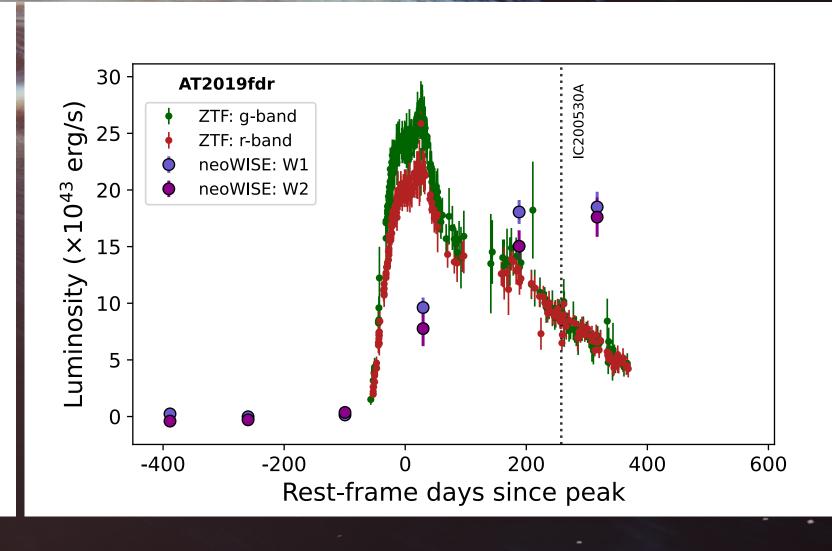


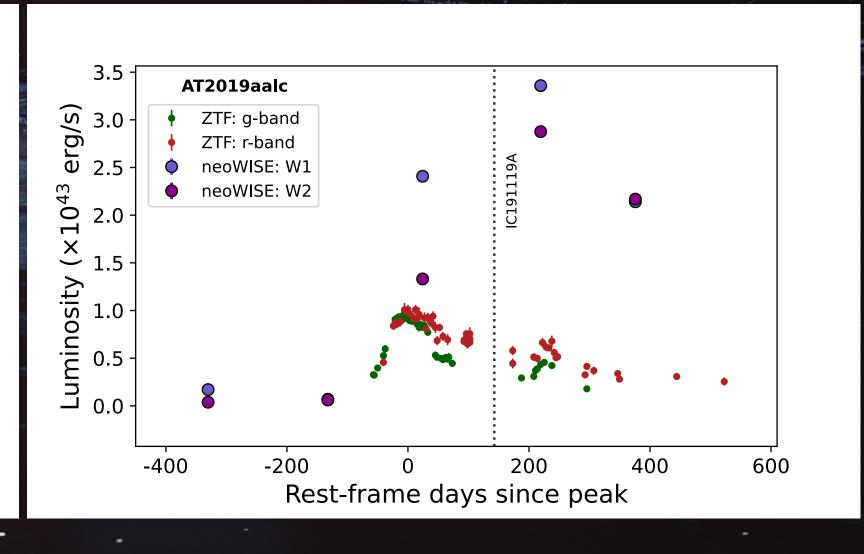
### Three neutrino-associated TDE candidates

### All had a very luminous dust echo

3.6 $\sigma$  correlation of 63 accretion flares with strong dust echo with high-energy alerts







### Lots of interesting transients, but rather unstructured

### Lots of interesting transients, but rather unstructured

What if we had a systematic sample of optical and IR lightcurves of the ZTF nuclear transients?

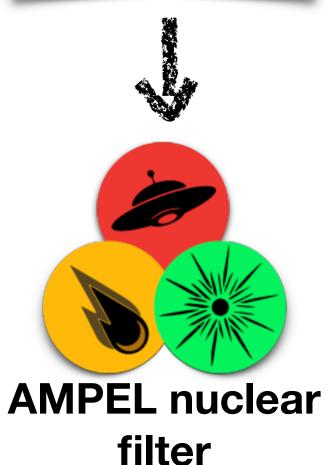
## II) A nuclear sample \*\*



[2018.5, 2021] → 3.5 years worth of data

(400 million alerts)





[2018.5, 2021]  $\rightarrow$  3.5 years worth of data

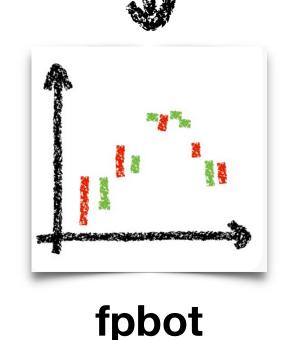
(400 million alerts)

Filtered by host distance, PS1 stargalaxy score, max. brightness and number of detections

11687 transients in final selection



AMPEL nuclear filter



[2018.5, 2021] → 3.5 years worth of data

(400 million alerts)

Filtered by host distance, PS1 stargalaxy score, max. brightness and number of detections

11687 transients in final selection

Obtained forced photometry with **fpbot**, applied baseline correction

Vast majority has a WISE counterpart



Image credit: NASA

Vast majority has a WISE counterpart

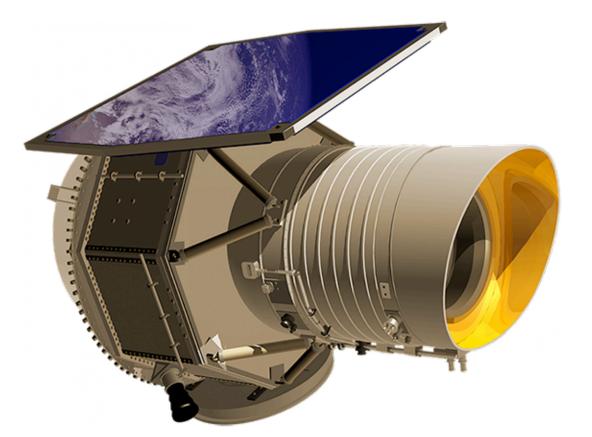
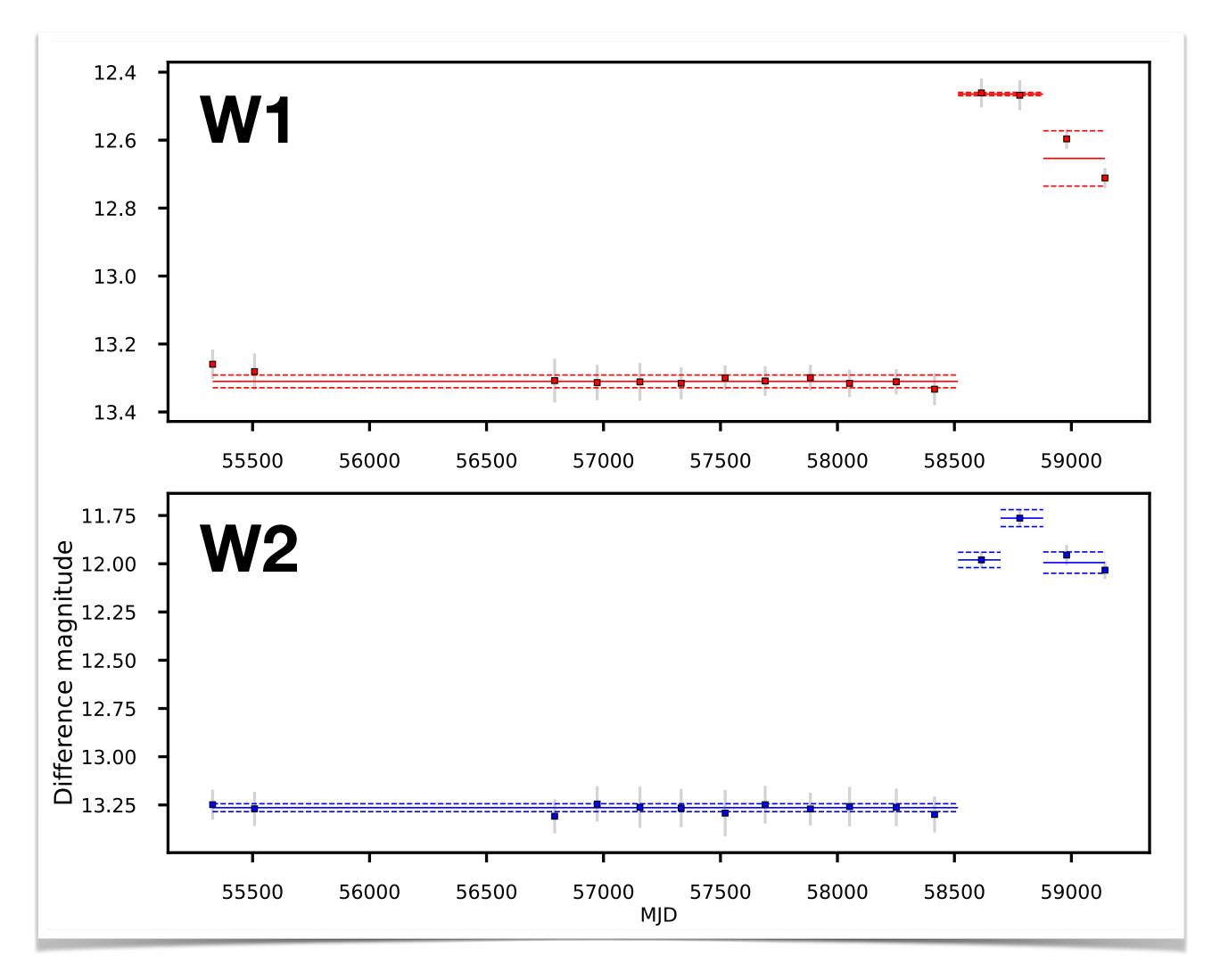


Image credit: NASA

Analyzed with a Bayesian block framework to find flares



Vast majority has a WISE counterpart

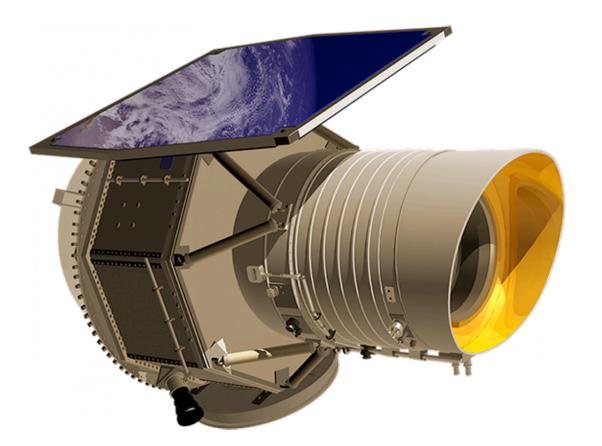
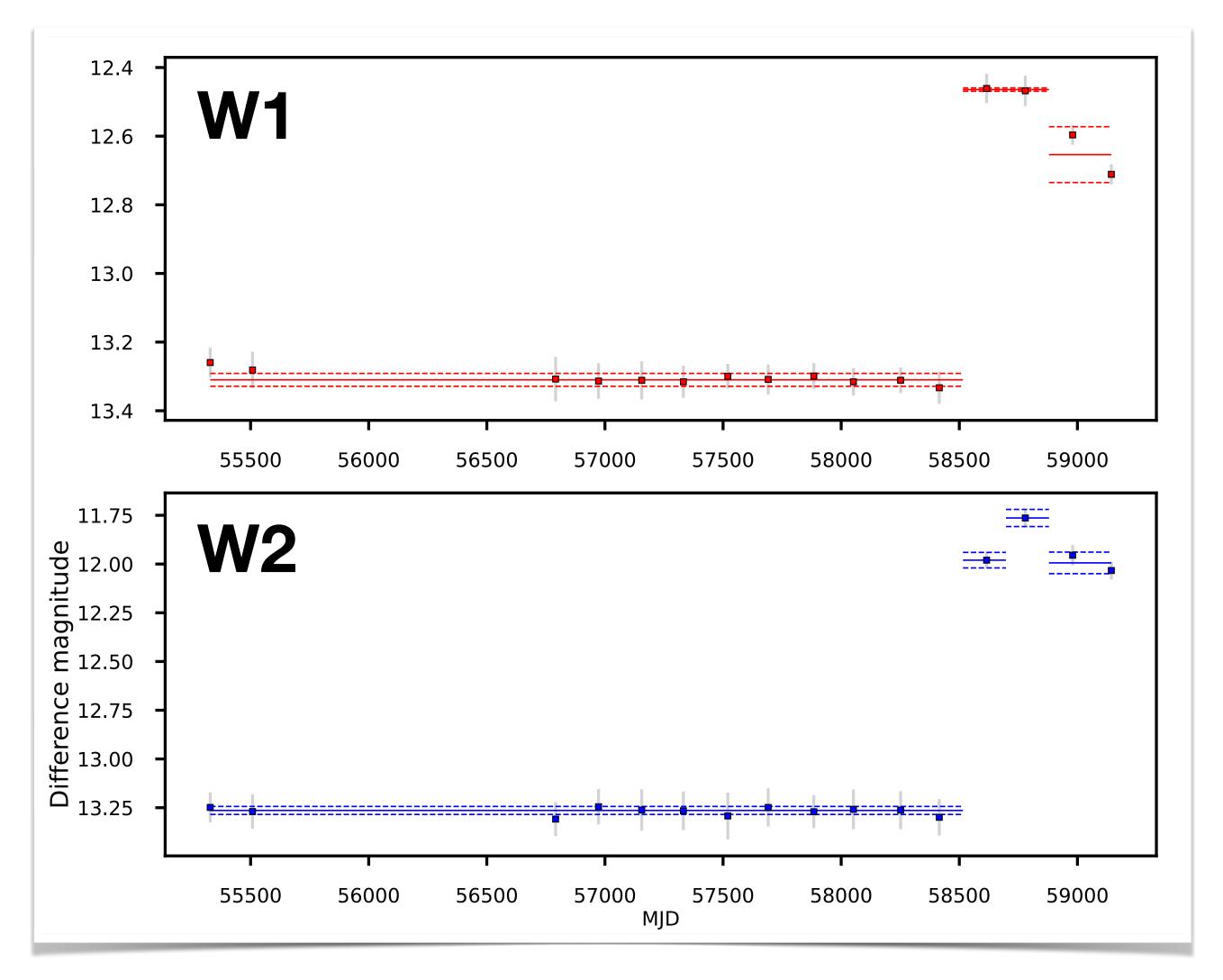


Image credit: NASA

Analyzed with a Bayesian block framework to find flares

15% have a prominent IR flare

2% flare after the optical peak



4/13

### Trained BDT classifier with noisified lightcurves from Bright Transient Survey



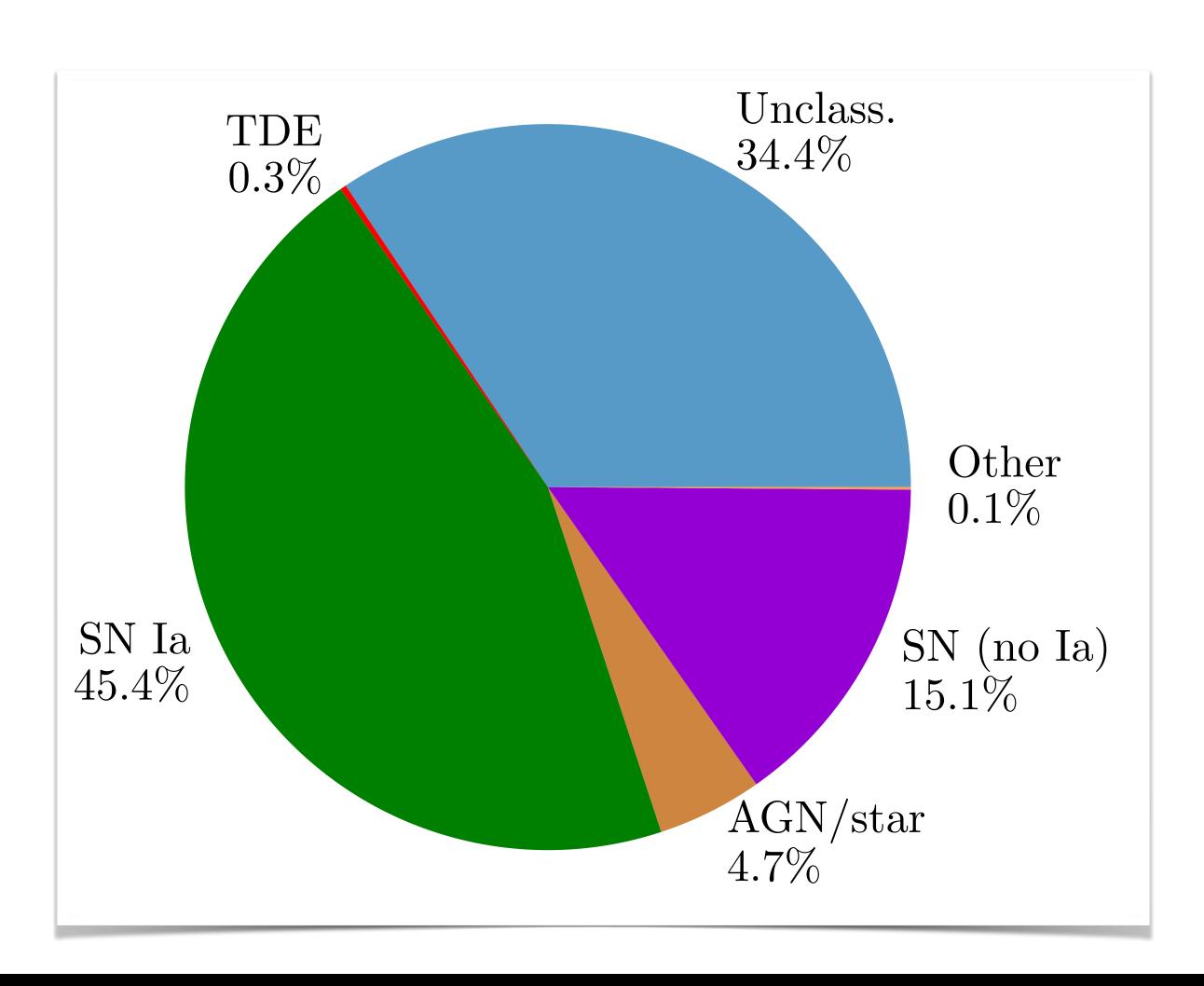
7284 ZTF transients

### Trained BDT classifier with noisified lightcurves from Bright Transient Survey

see Dan Perley's talk

7284 ZTF transients

5315 of them classified (brighter than 18.5 mag)



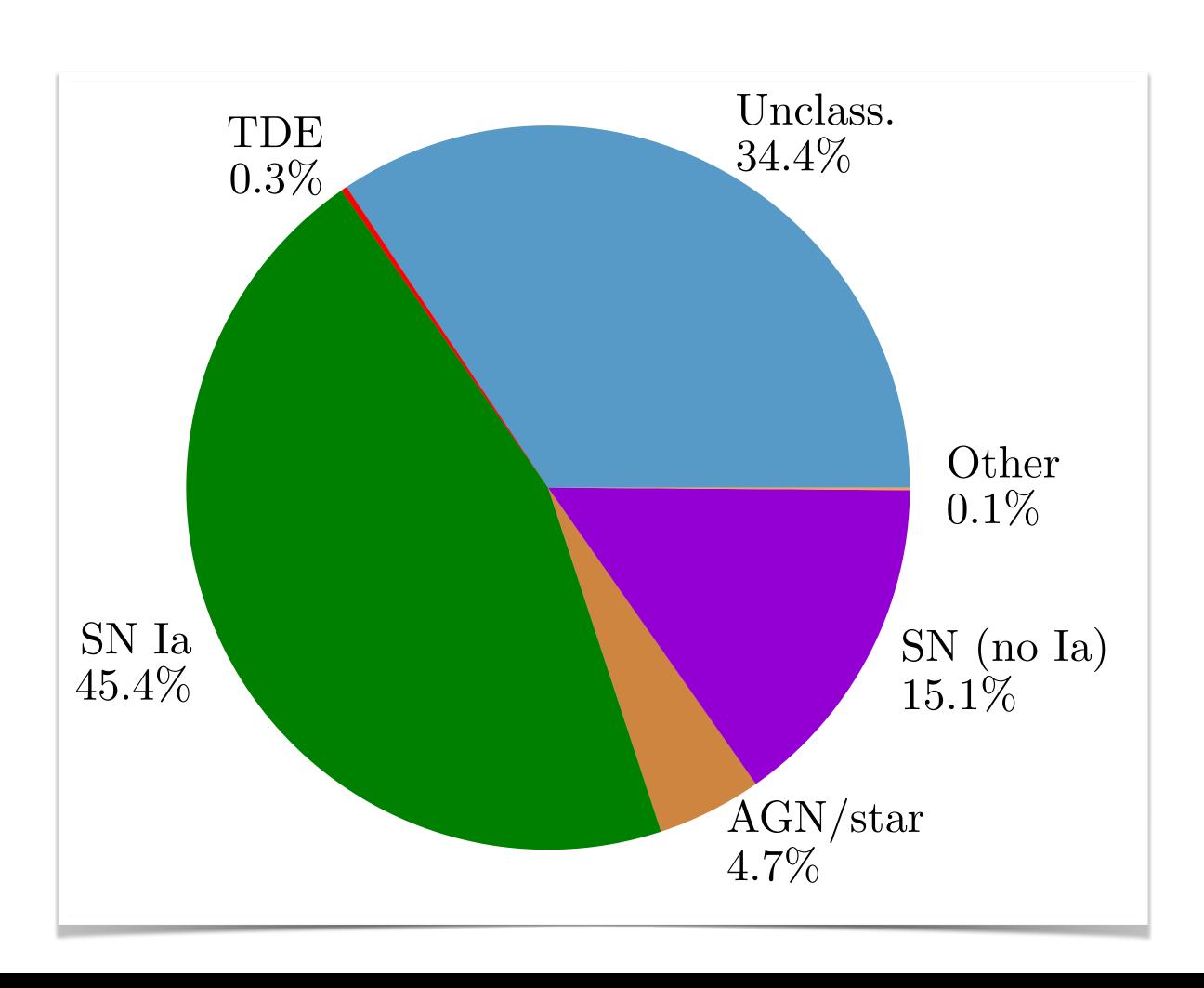
### Trained BDT classifier with noisified lightcurves from Bright Transient Survey

see Dan Perley's talk

7284 ZTF transients

5315 of them classified (brighter than 18.5 mag)

Problem: Brighter than nuclear selection, high class imbalance



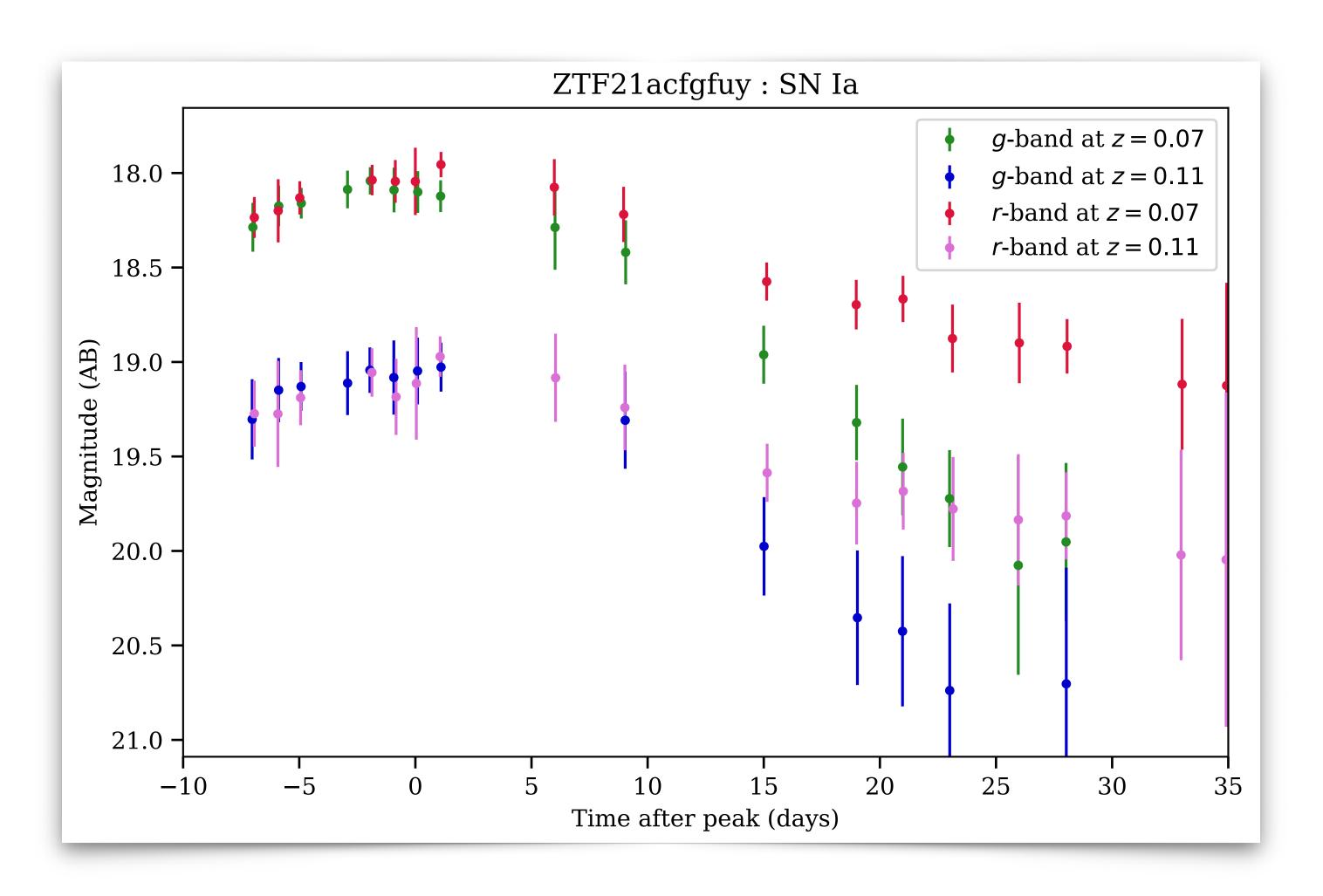
### Trained BDT classifier with noisified lightcurves from the Bright Transient Survey

Noisification: Physically motivated remedy

### Trained BDT classifier with noisified lightcurves from the Bright Transient Survey

Noisification: Physically motivated remedy

- → redshift
- → noisify
- → K-correct

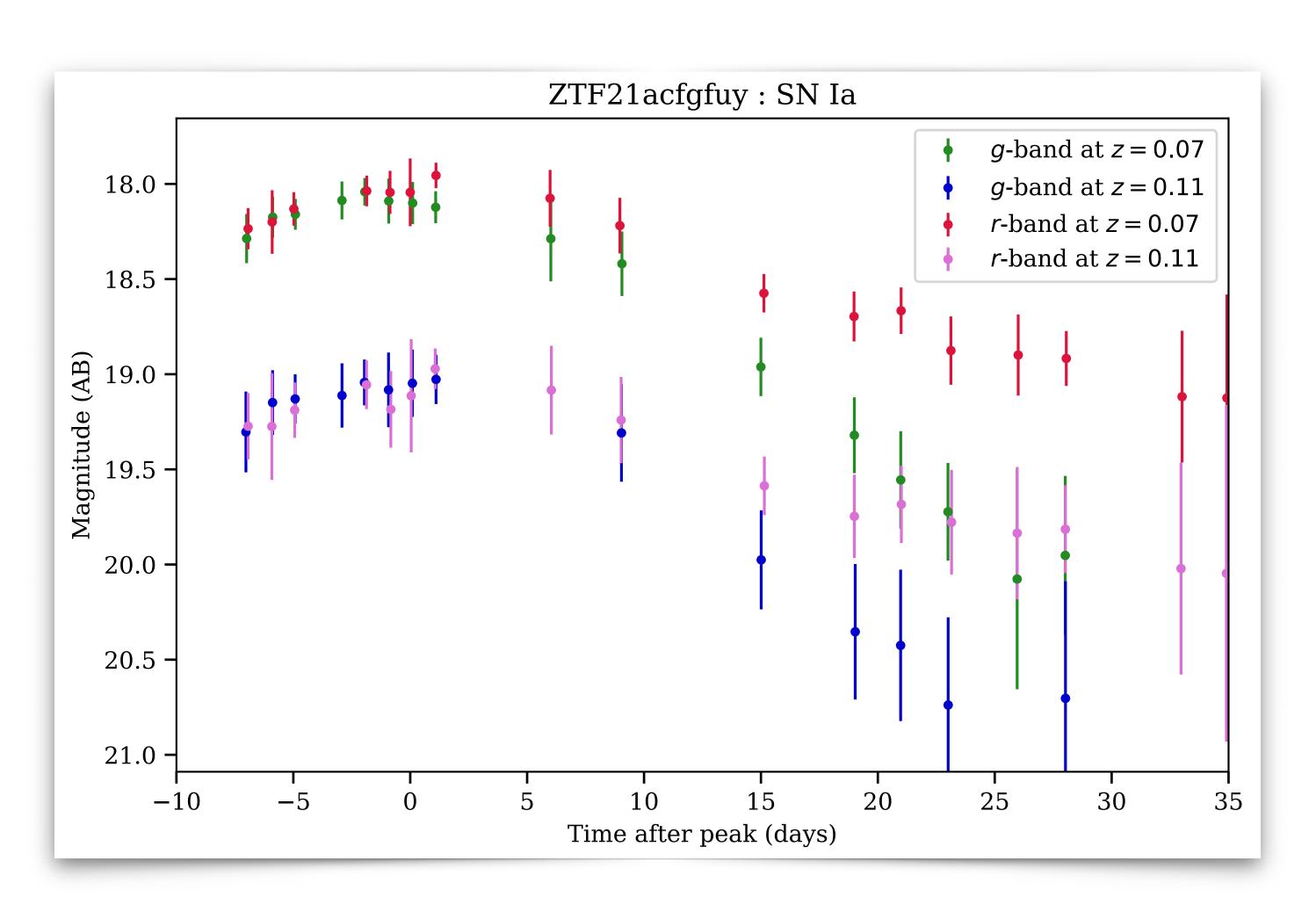


### Trained BDT classifier with noisified lightcurves from the Bright Transient Survey

Noisification: Physically motivated remedy

- → redshift
- → noisify
- → K-correct

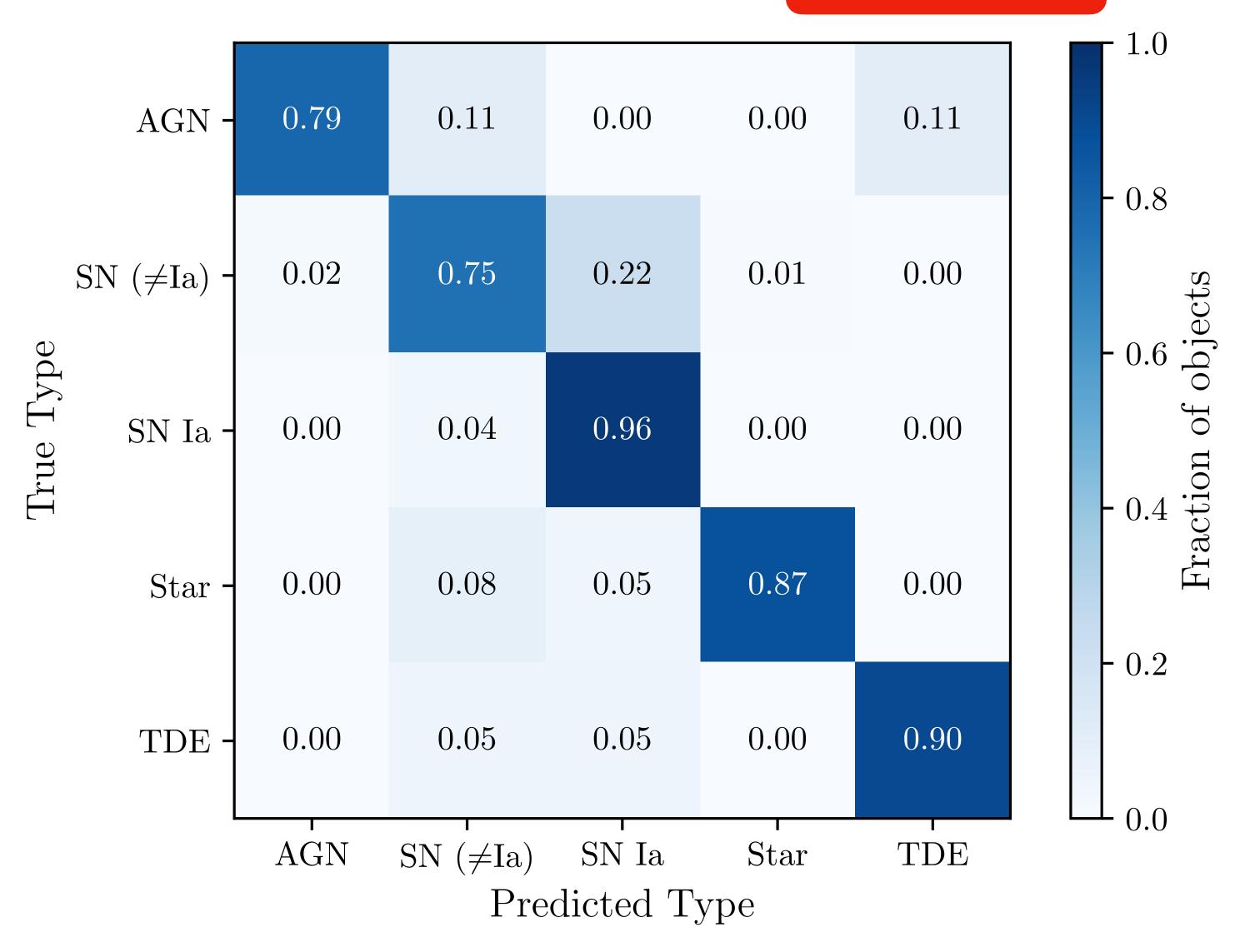
Feature extraction, currently only based on optical data



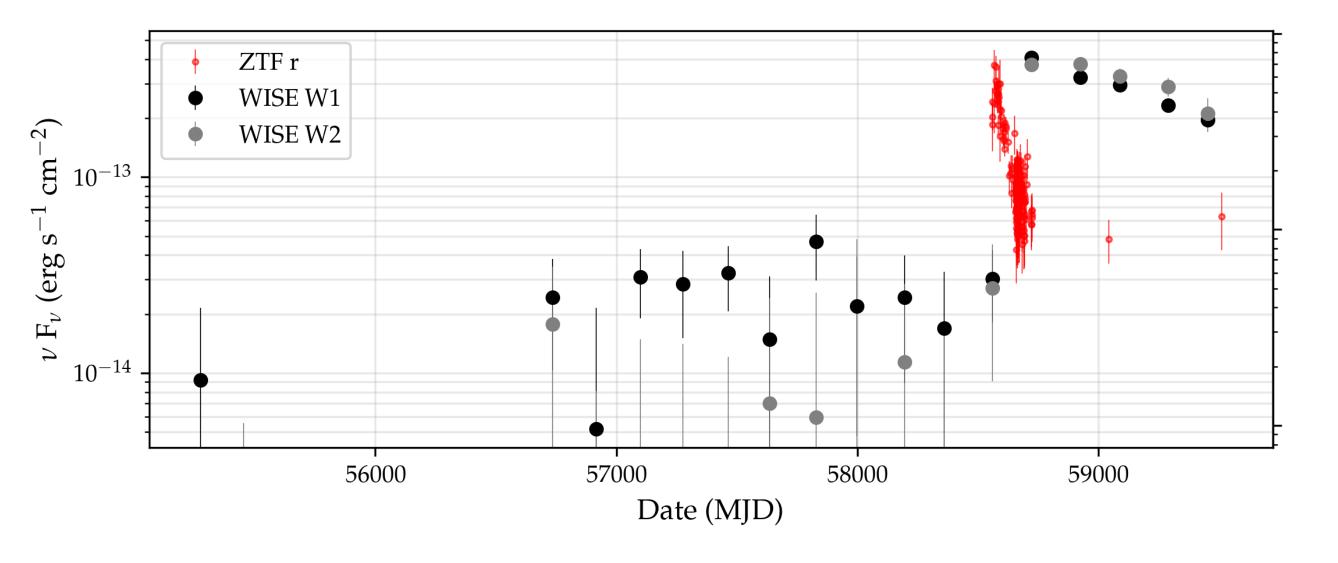
#### Evaluation with test sample

preliminary

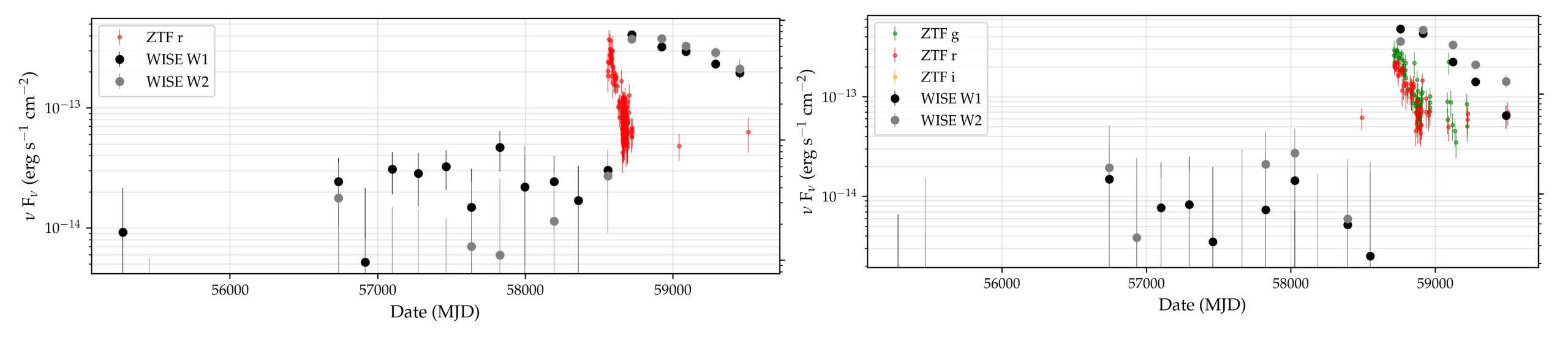
Not including noisified lightcurves in the test sample



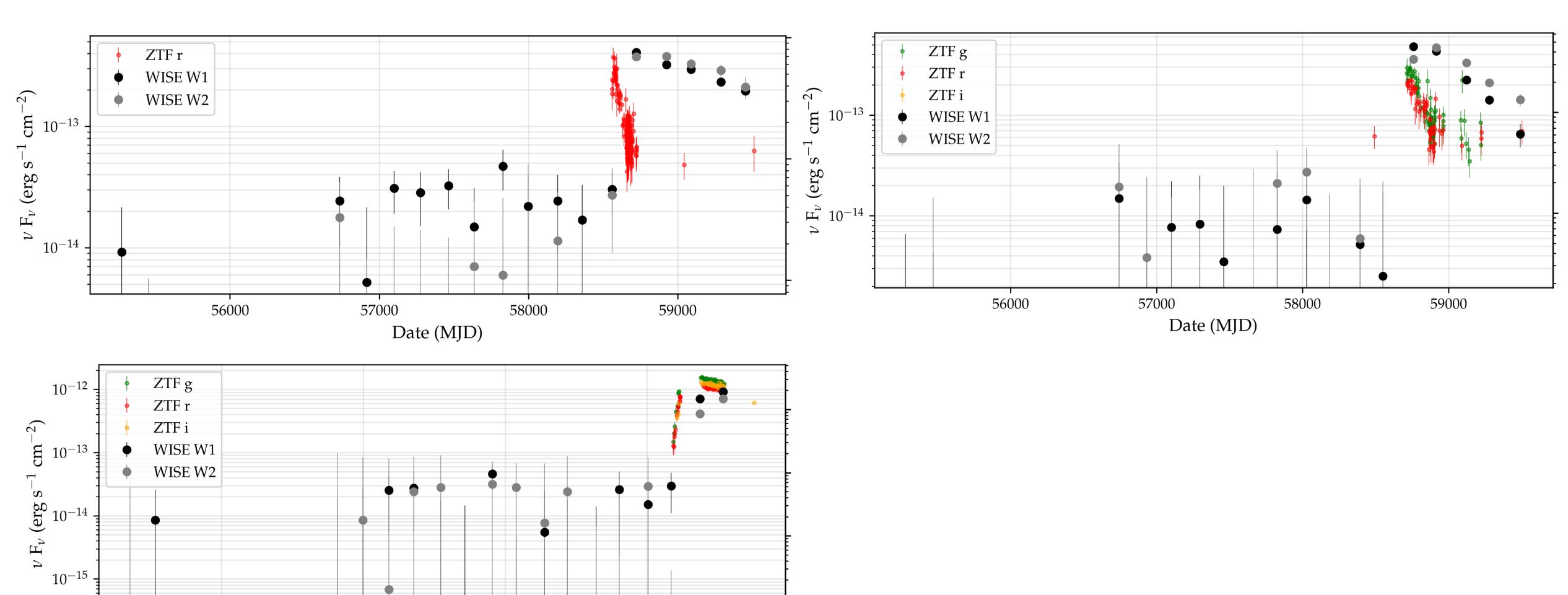
### Identified some strong TDE candidates previously missed



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### Identified some strong TDE candidates previously missed



57000

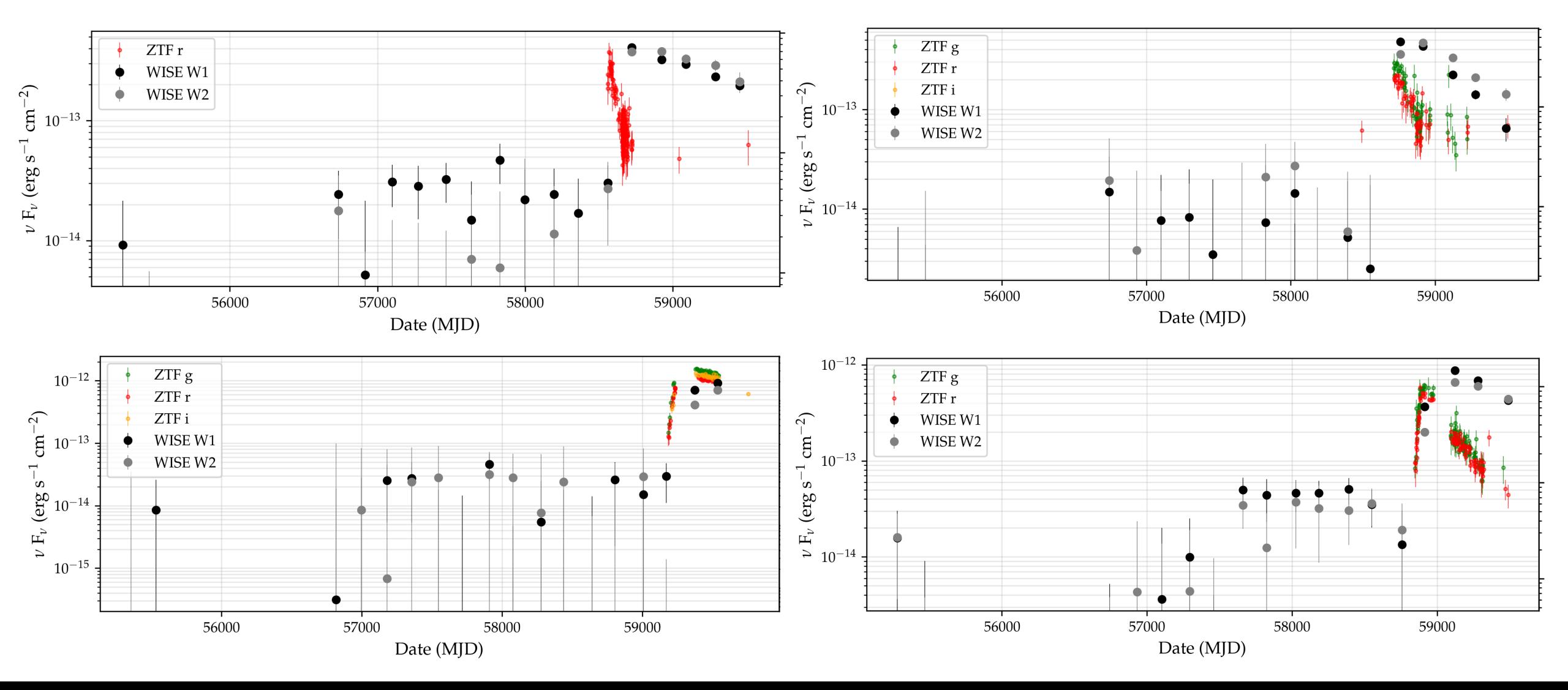
58000

Date (MJD)

59000

56000

# Identified some strong TDE candidates previously missed



# Identified some strong TDE candidates previously missed



# III) What does this tell us about GW follow up?

### What can we learn from such a sample?

Photometric identification is still a challenge!

There might be types of events in the sample we missed so far

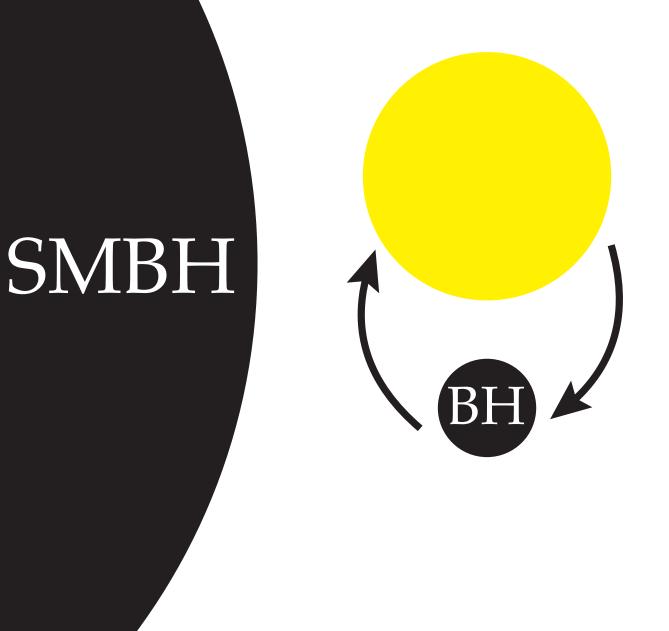
We know about SNe, stochastic AGN variability, AGN flares, CVs and TDEs. But what else?

Exclusively looking for KN signatures poses the danger of missing something

BBH counterparts (slowly evolving!)

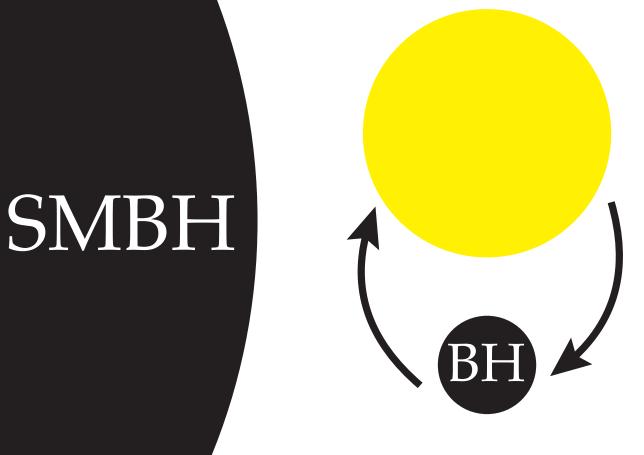
See Matthew Graham's talk

stellar mass BH tidally disrupting a star



Y. Yang+ 2022

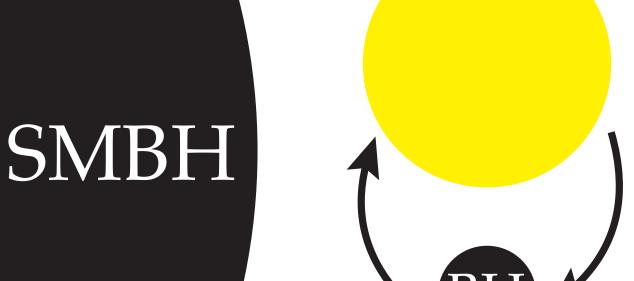
stellar mass BH tidally disrupting a star



Maybe enhanced rate close to the SMBH due to pre-existing accretion disk

Y. Yang+ 2022

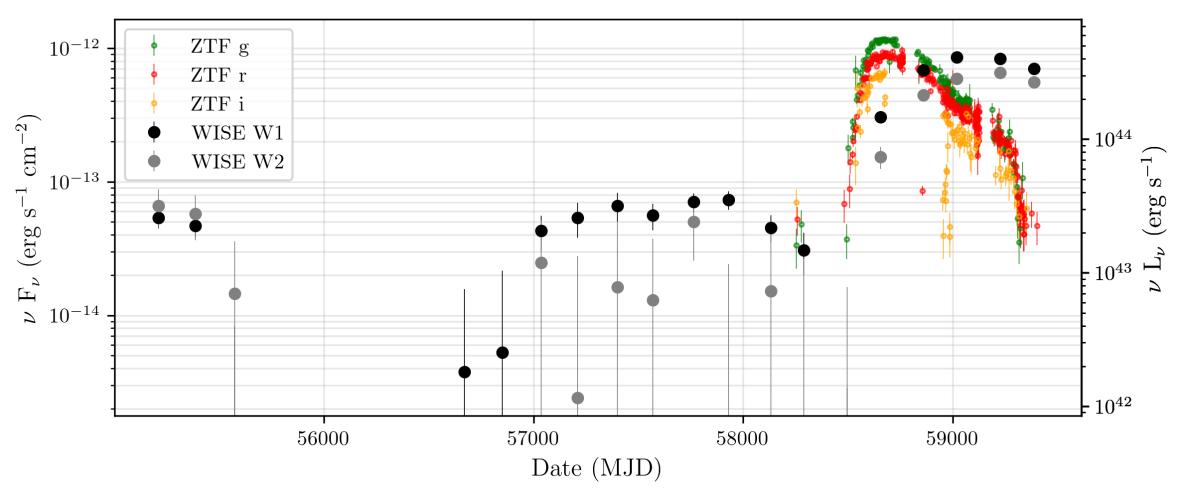
stellar mass BH tidally disrupting a star



Maybe enhanced rate close to the SMBH due to pre-existing accretion disk

EM signature: Similar to TDEs, but SMBH mass above Hills mass for **bright**  $\mu$ TDEs

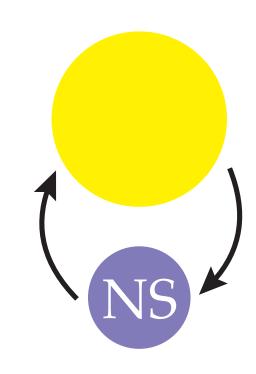
Faint  $\mu$ TDEs: unusual flaring in the AGN light curve

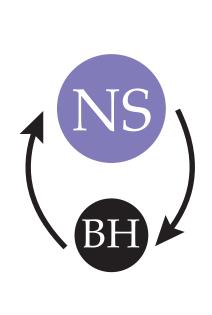


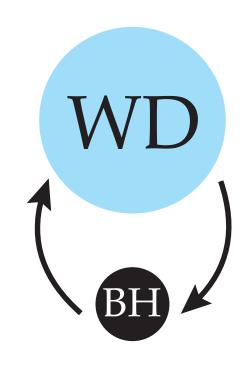
Literature candidate  $\mu$ TDE **ZTF19aailpwl** (part of the sample)

Y. Yang+ 2022

SMBH

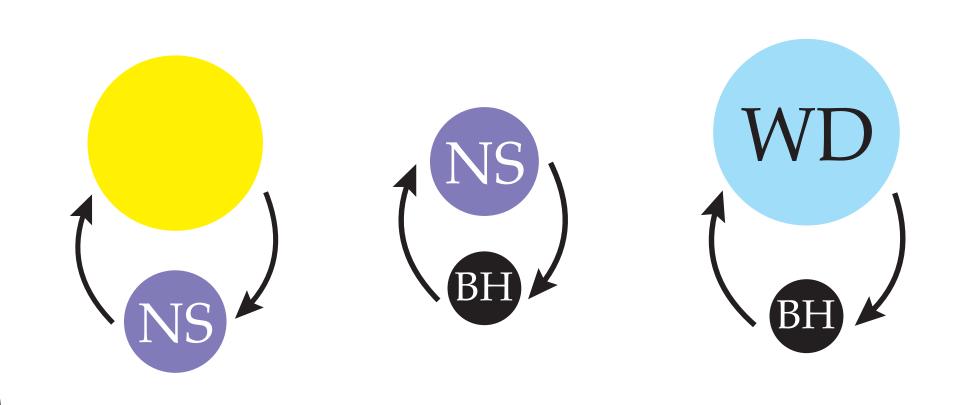






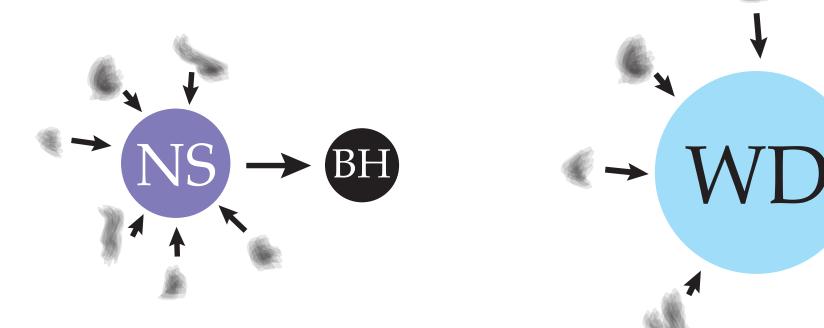
Other  $\mu$ TDE channels

SMBH

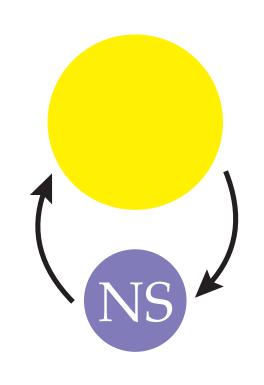


Other  $\mu$ TDE channels

SMBH



Accretion induced collapse



There should be some of these contained in the nuclear sample

S

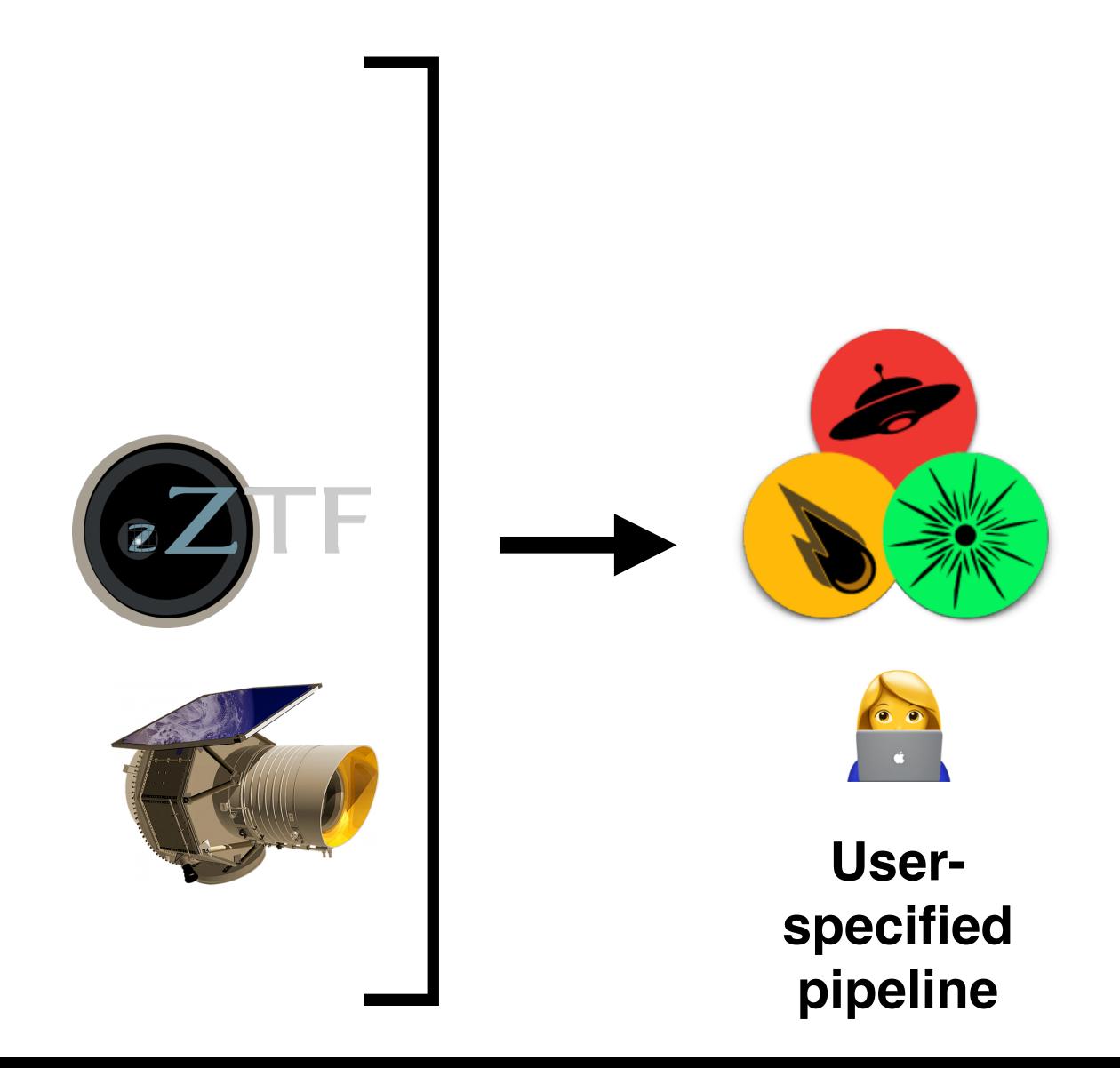
SMBH

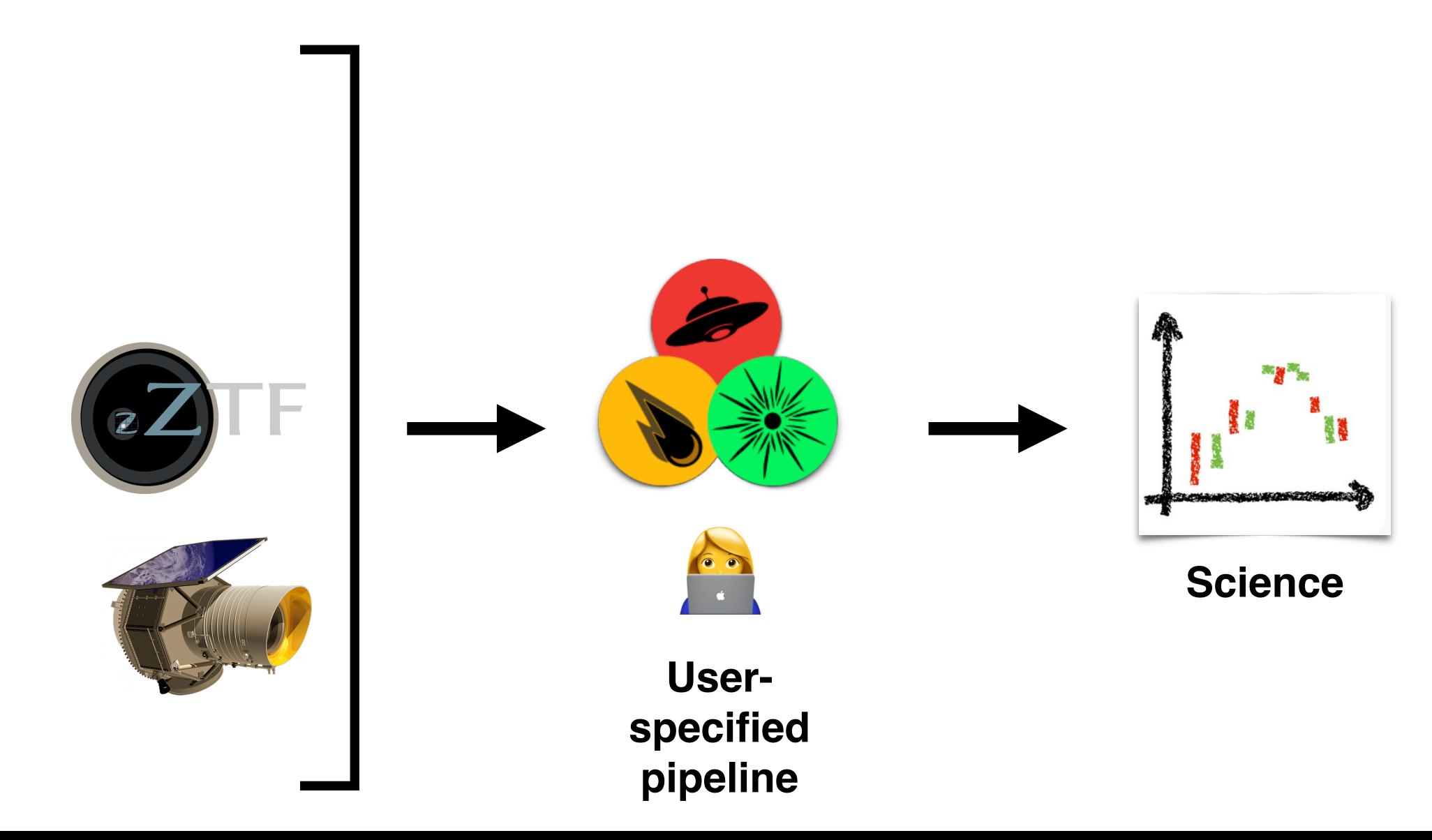
GW signals unlikely to be accessible to LVK

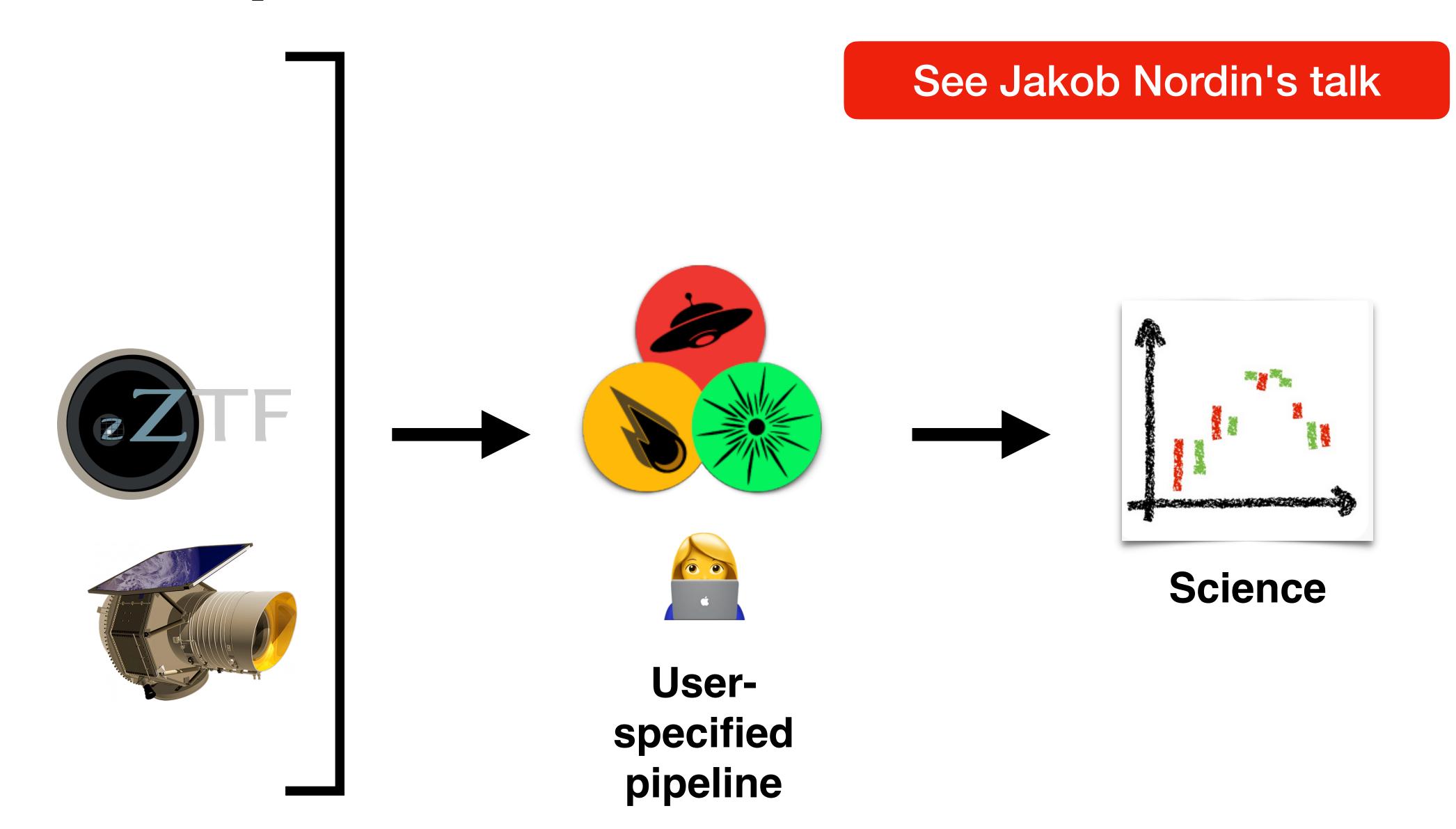
luced

But one can potentially learn about BH evolution and binary formation in AGN disks



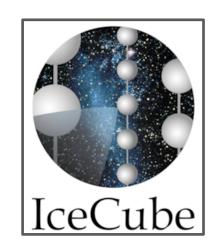






#### There is so much more data to be combined







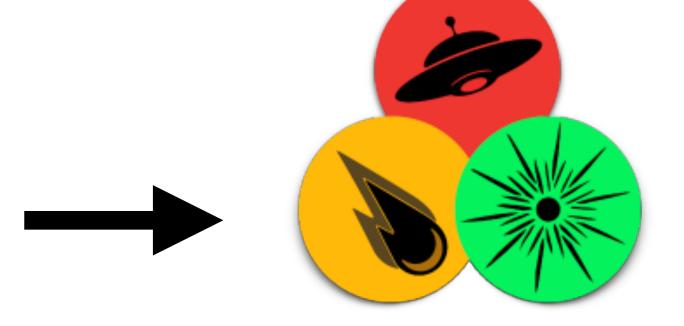






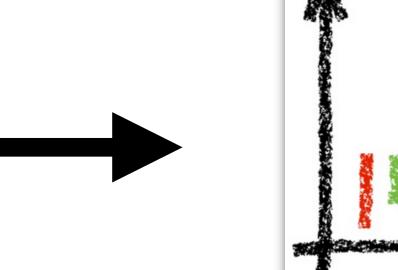
















Wealth of data should increase the likelihood of serendipitous discovery

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→ But only if we have the right tools for it!

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How about automatic coincidence searches for **all** O4 alerts?

Wealth of data should increase the likelihood of serendipitous discovery

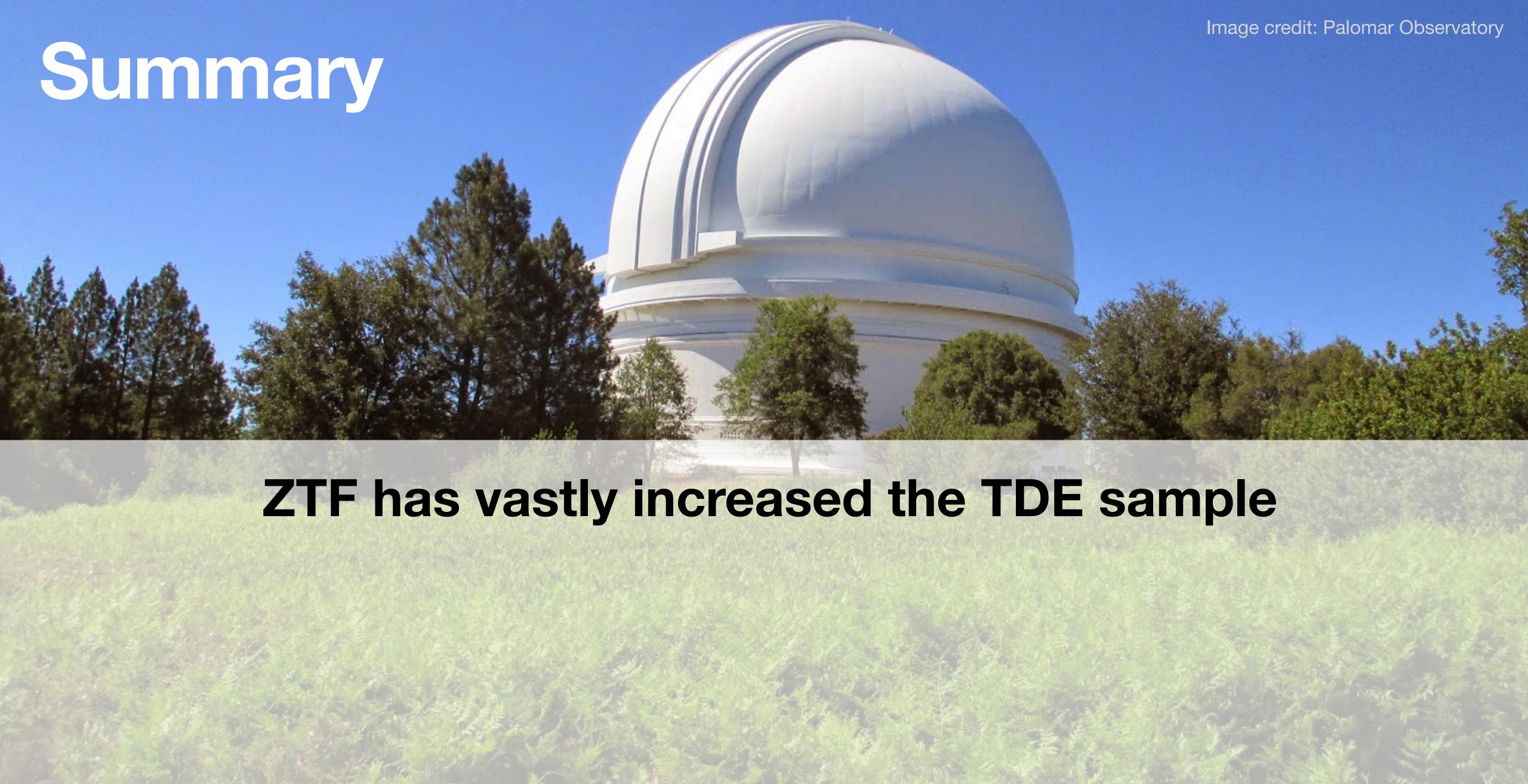
→ But only if we have the right tools for it!

Rapid follow up is great! But maybe there is something to be gained from archival studies

How about automatic coincidence searches for **all** O4 alerts?

Background studies using the normal survey ("blind")







ZTF has vastly increased the TDE sample Created a flux-limited, unbiased sample of nuclear events



ZTF has vastly increased the TDE sample
Created a flux-limited, unbiased sample of nuclear events
Classification effort ongoing

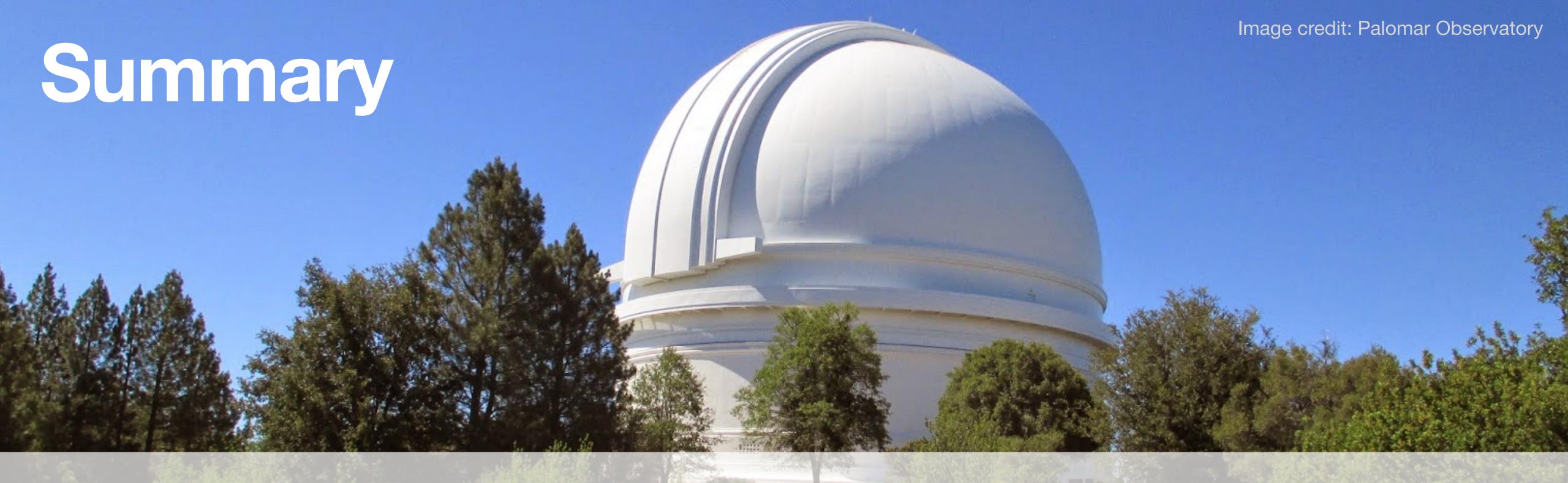


ZTF has vastly increased the TDE sample
Created a flux-limited, unbiased sample of nuclear events
Classification effort ongoing
Explore the wealth of transients contained in the sample



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AMPEL enables all this

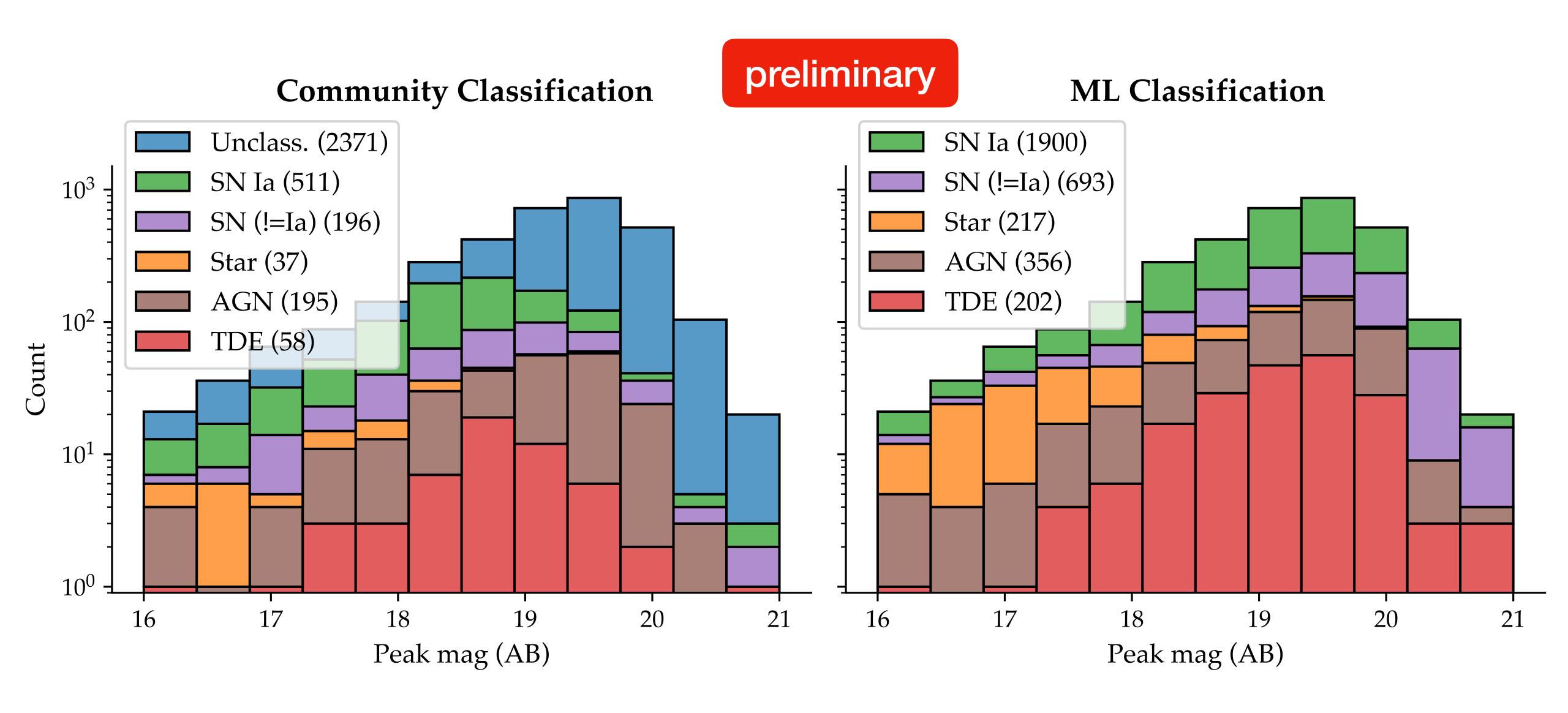


ZTF has vastly increased the TDE sample
Created a flux-limited, unbiased sample of nuclear events
Classification effort ongoing
Explore the wealth of transients contained in the sample

§ AMPEL enables all this
Archival analyses have their own merits

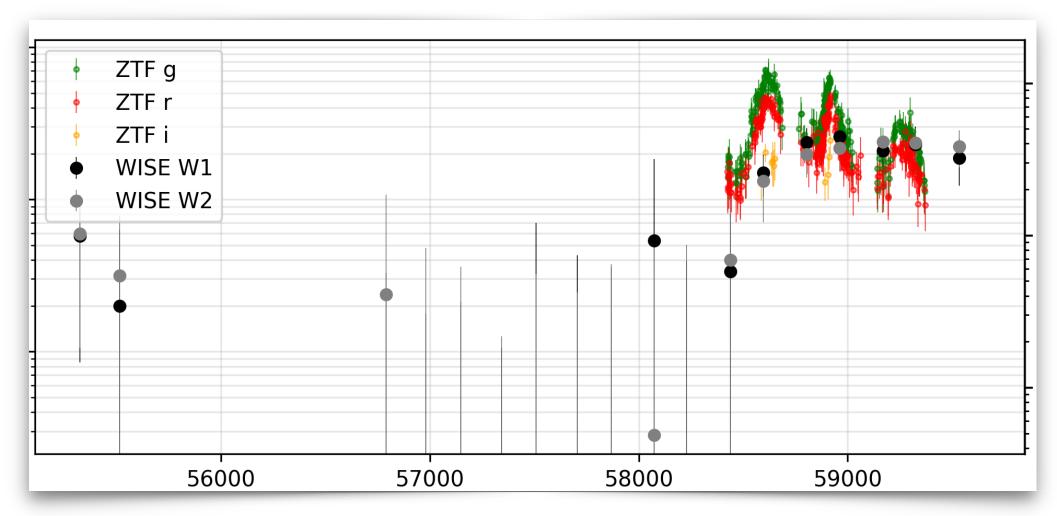
## Bonus

#### Classification results

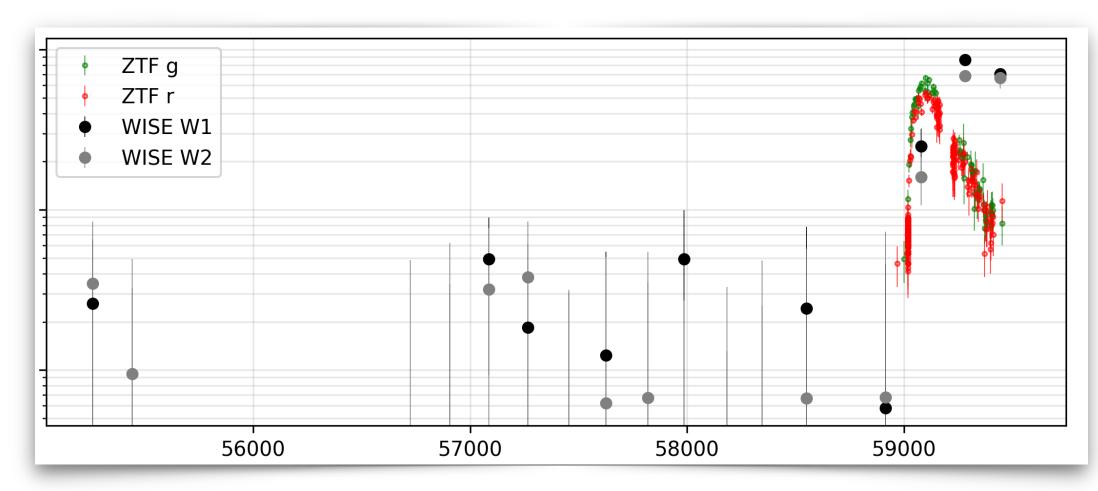


#### IR transients: Filtered by IR

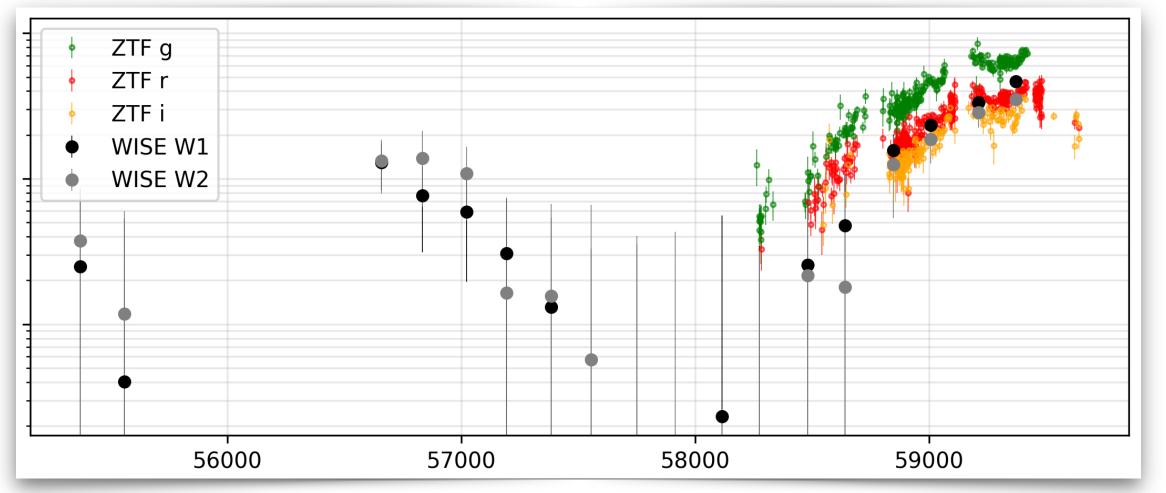
#### **Blazar candidates**



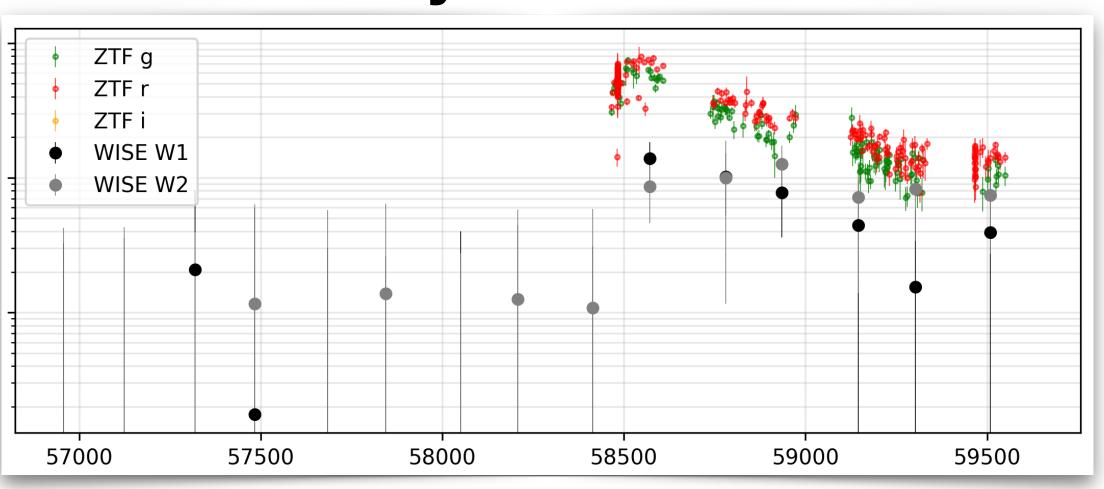
#### **TDE** candidate



#### slow rise, simultaneous IR rise

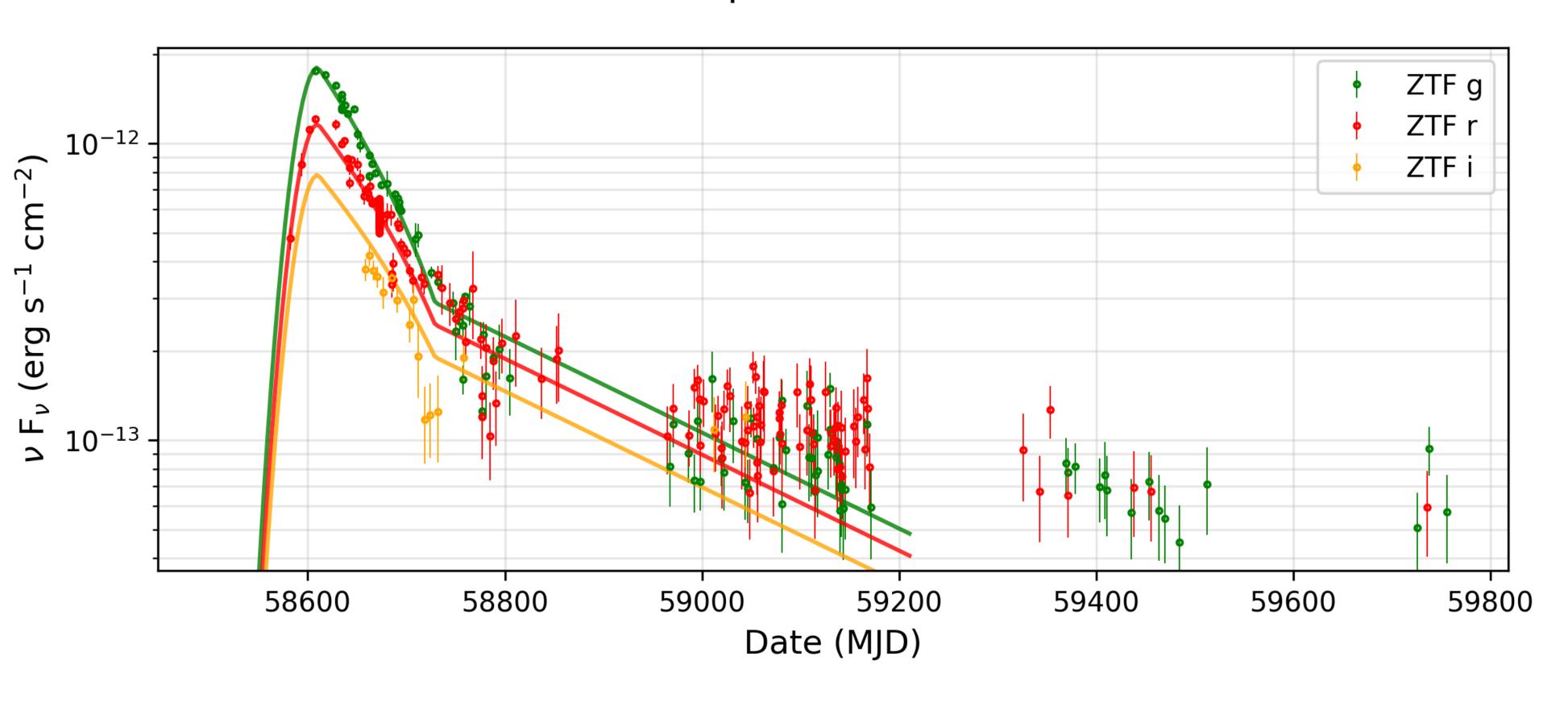


#### Tywin-like

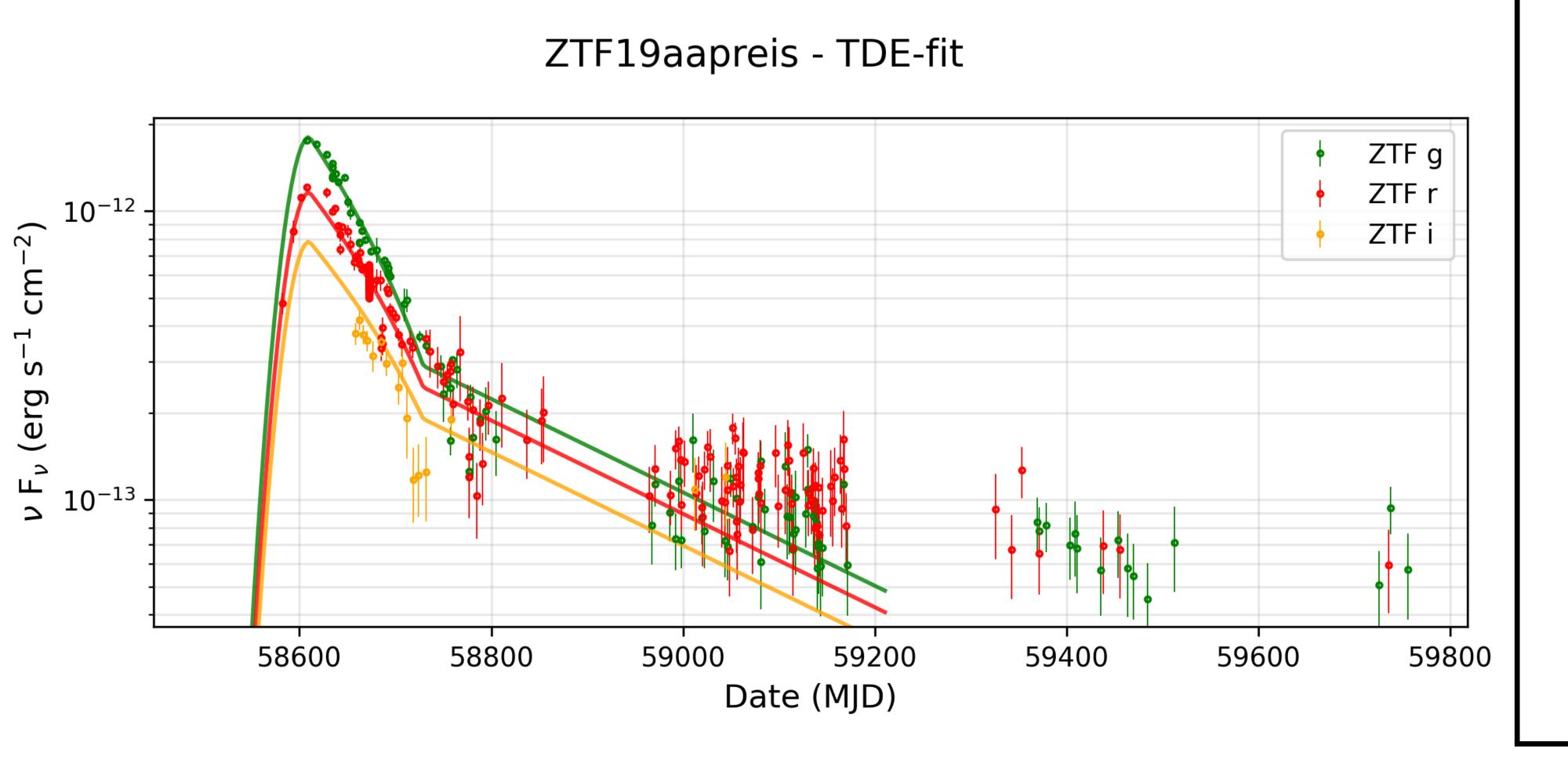


#### Feature extraction: TDE Fit

ZTF19aapreis - TDE-fit



#### Feature extraction: TDE Fit

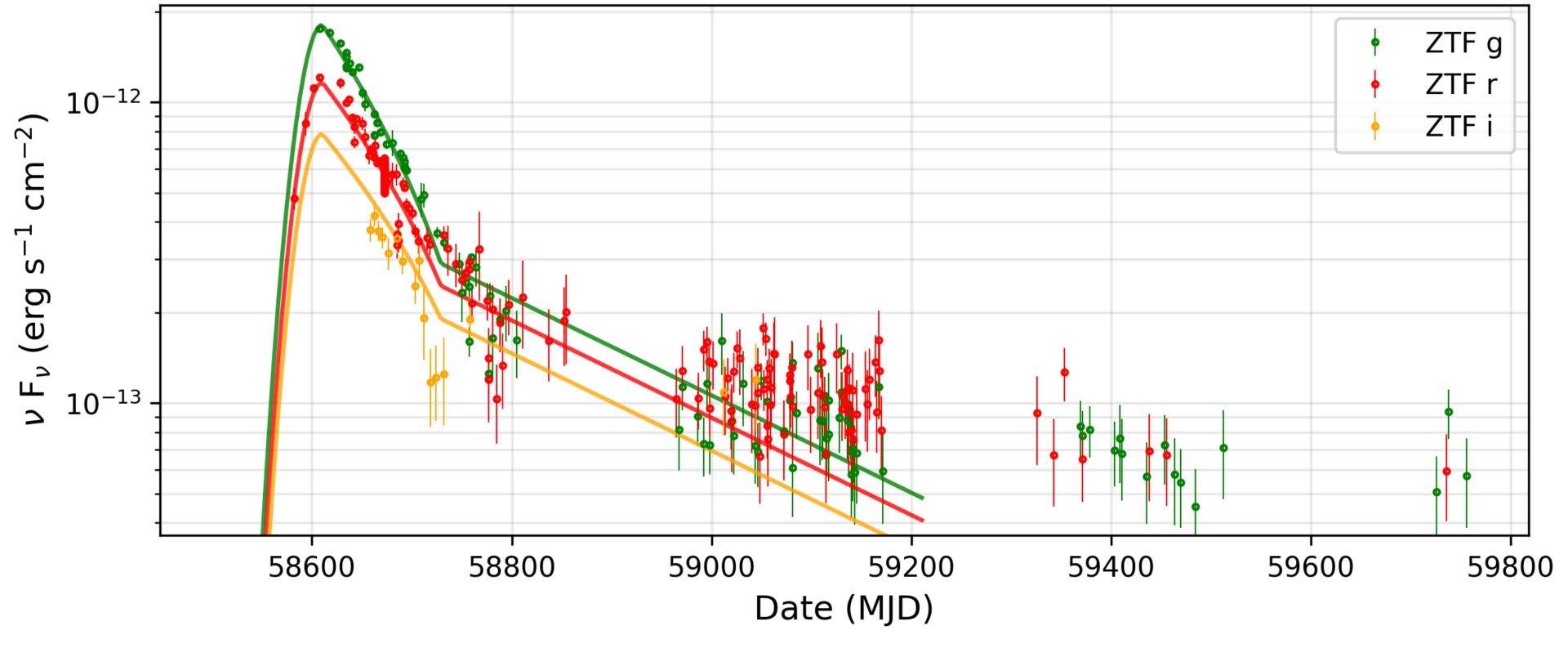


#### Fit parameters:

risetime
decaytime
BB temp
BB d\_temp
t\_evo time

#### Feature extraction: TDE Fit





#### Fit parameters:

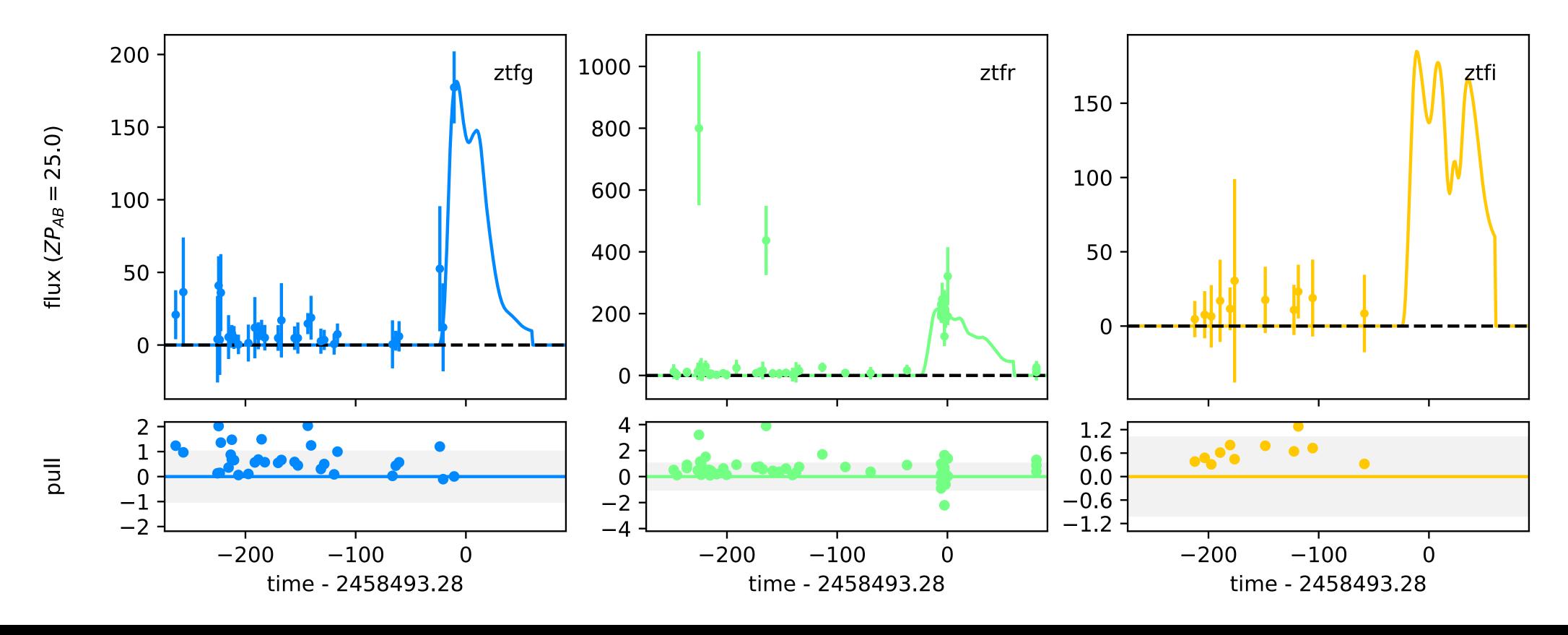
risetime
decaytime
BB temp
BB d\_temp
t\_evo time

Fitted twice (stability)
Problem: Trunaway (no UV)

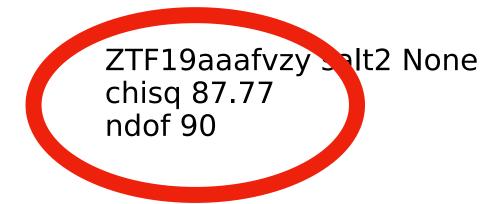
#### Feature extraction: SALT Fit (SN Ia)

ZTF19aaafvzy salt2 None chisq 87.77 ndof 90

z = 0.21352750  $t_0 = 2458493.3$   $x_0 = 2.6827823 \times 10^{-4}$  $x_1 = 8.3501711$  c = 0.17711280 mw ebv = 0.068765168 $mw r_v = 3.1000000$ 



#### Feature extraction: SALT Fit (SN Ia)



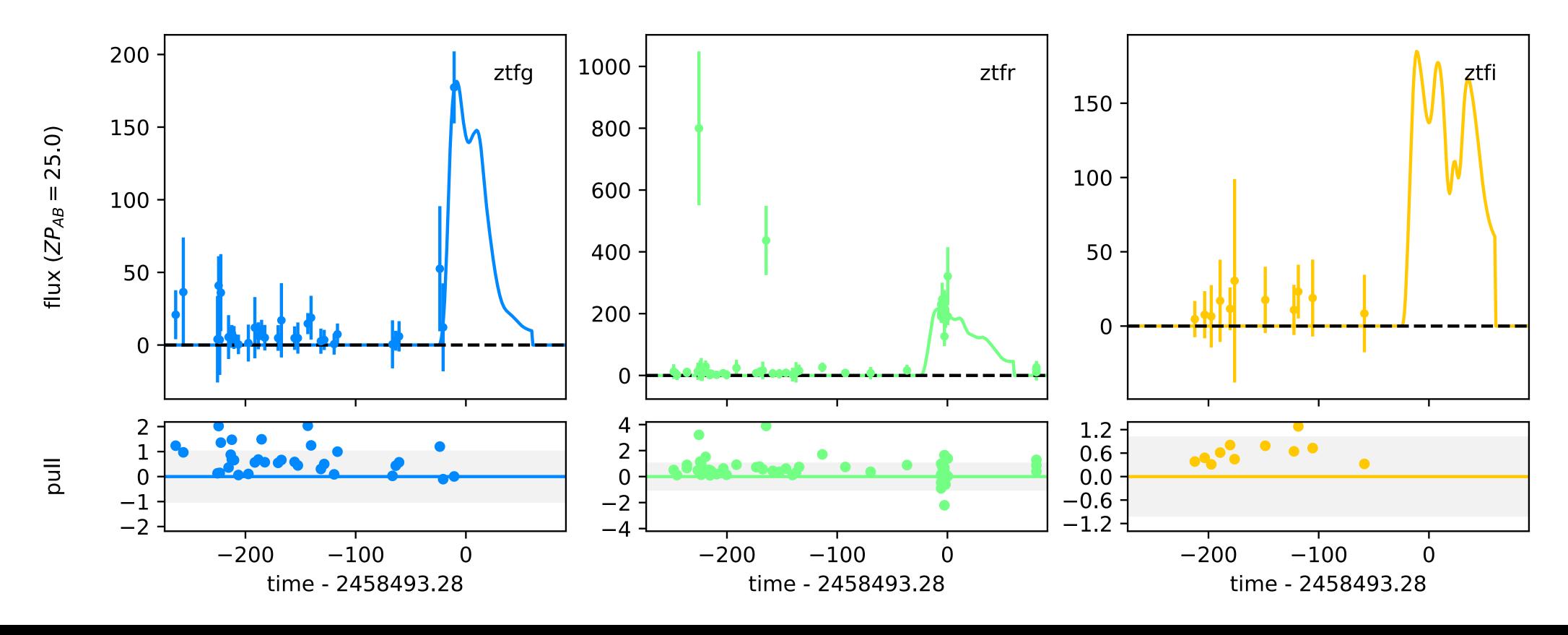
```
z = 0.21352750

t_0 = 2458493.3

x_0 = 2.6827823 \times 10^{-4}

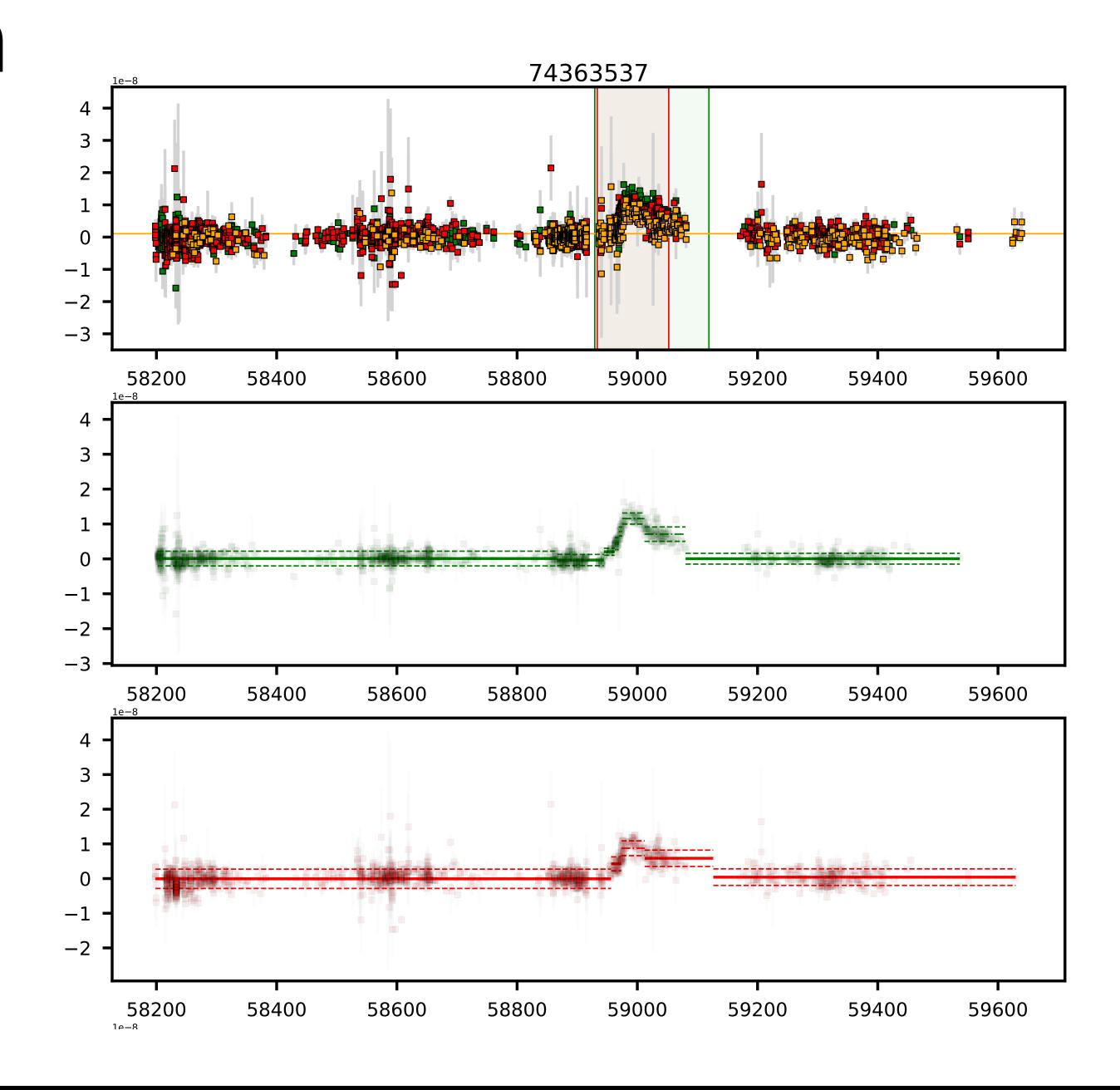
x_1 = 8.3501711
```

c = 0.17711280 mw ebv = 0.068765168 $mw r_v = 3.1000000$ 



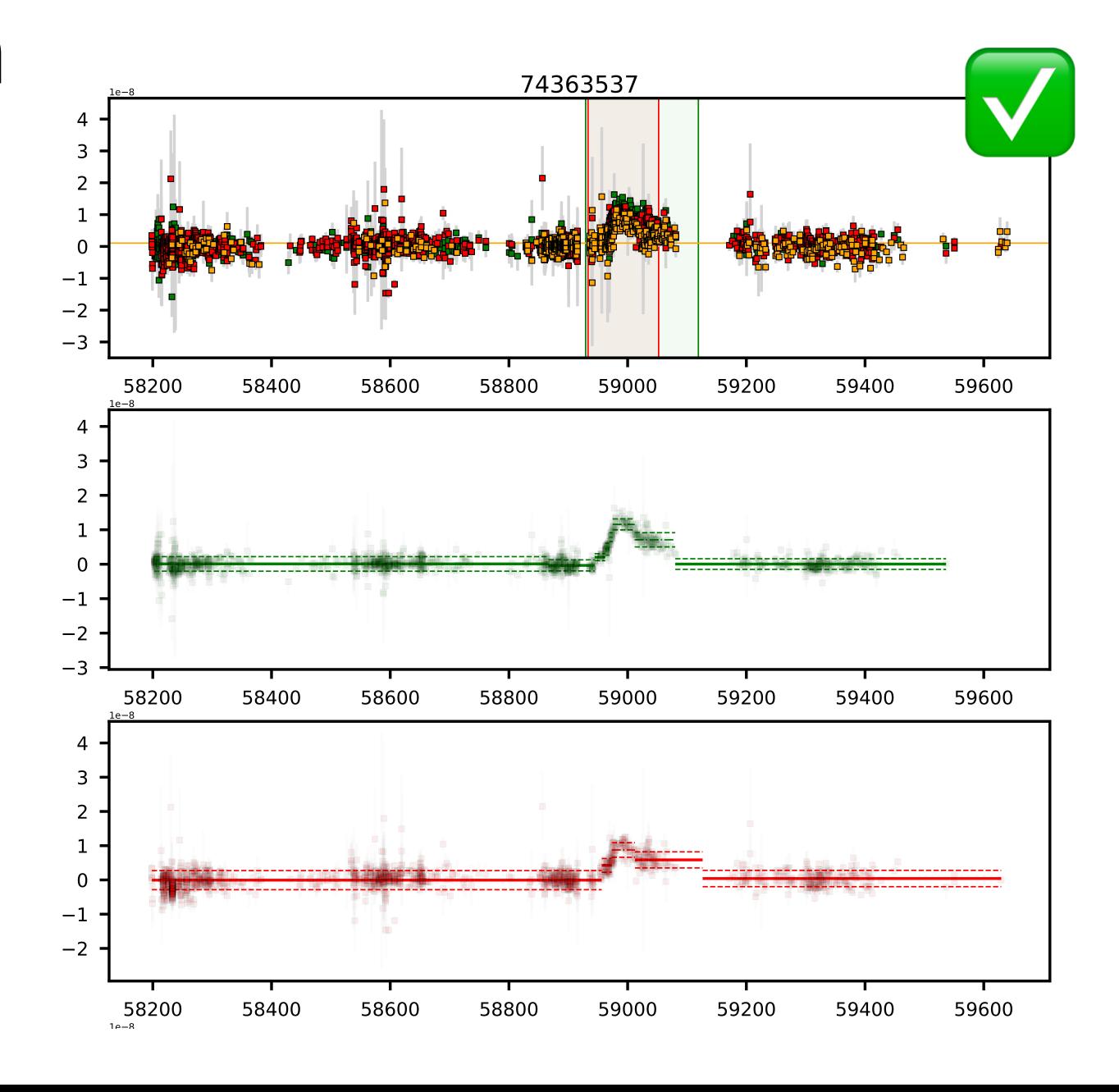
all bands

g-band



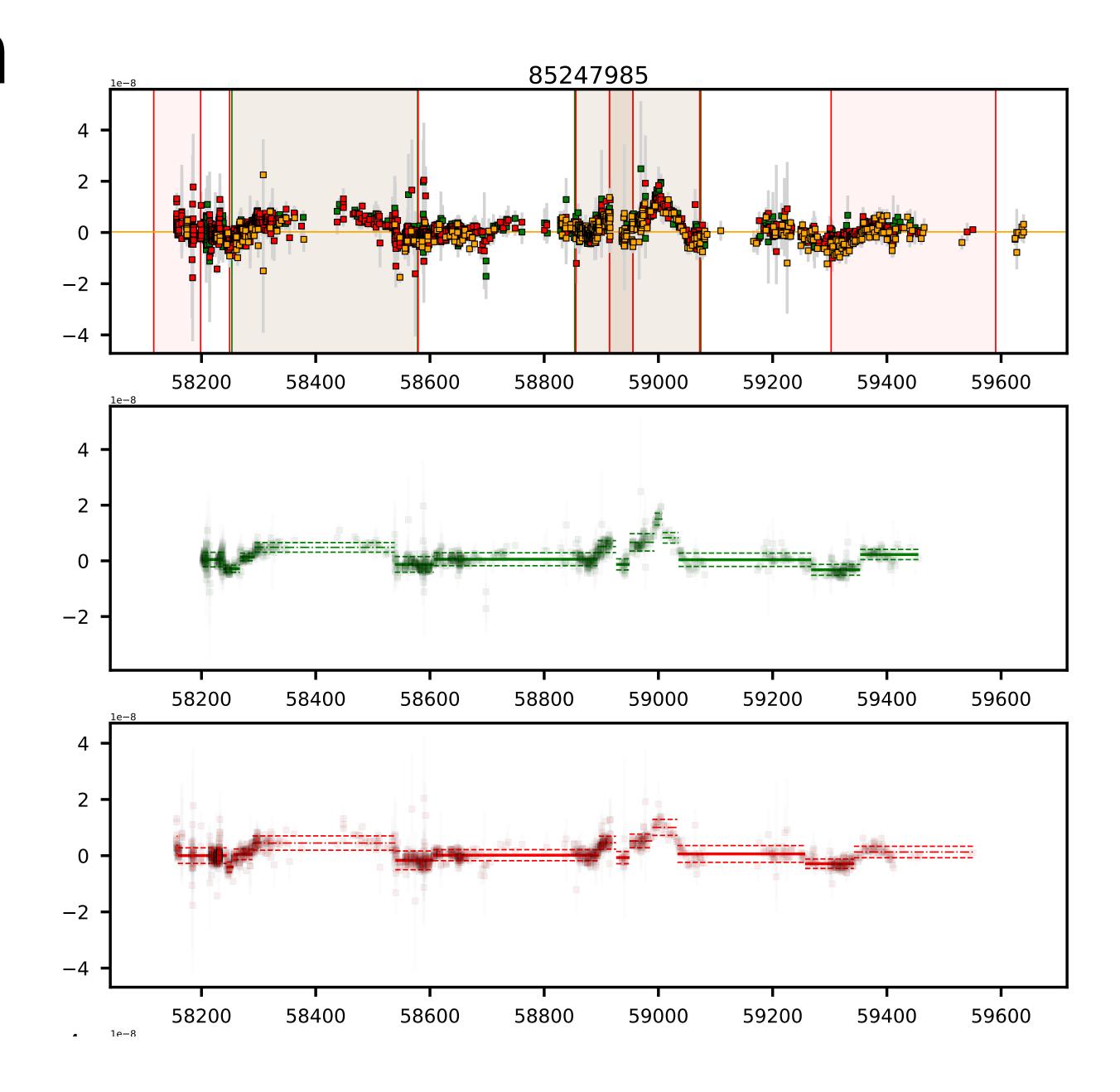
all bands

g-band



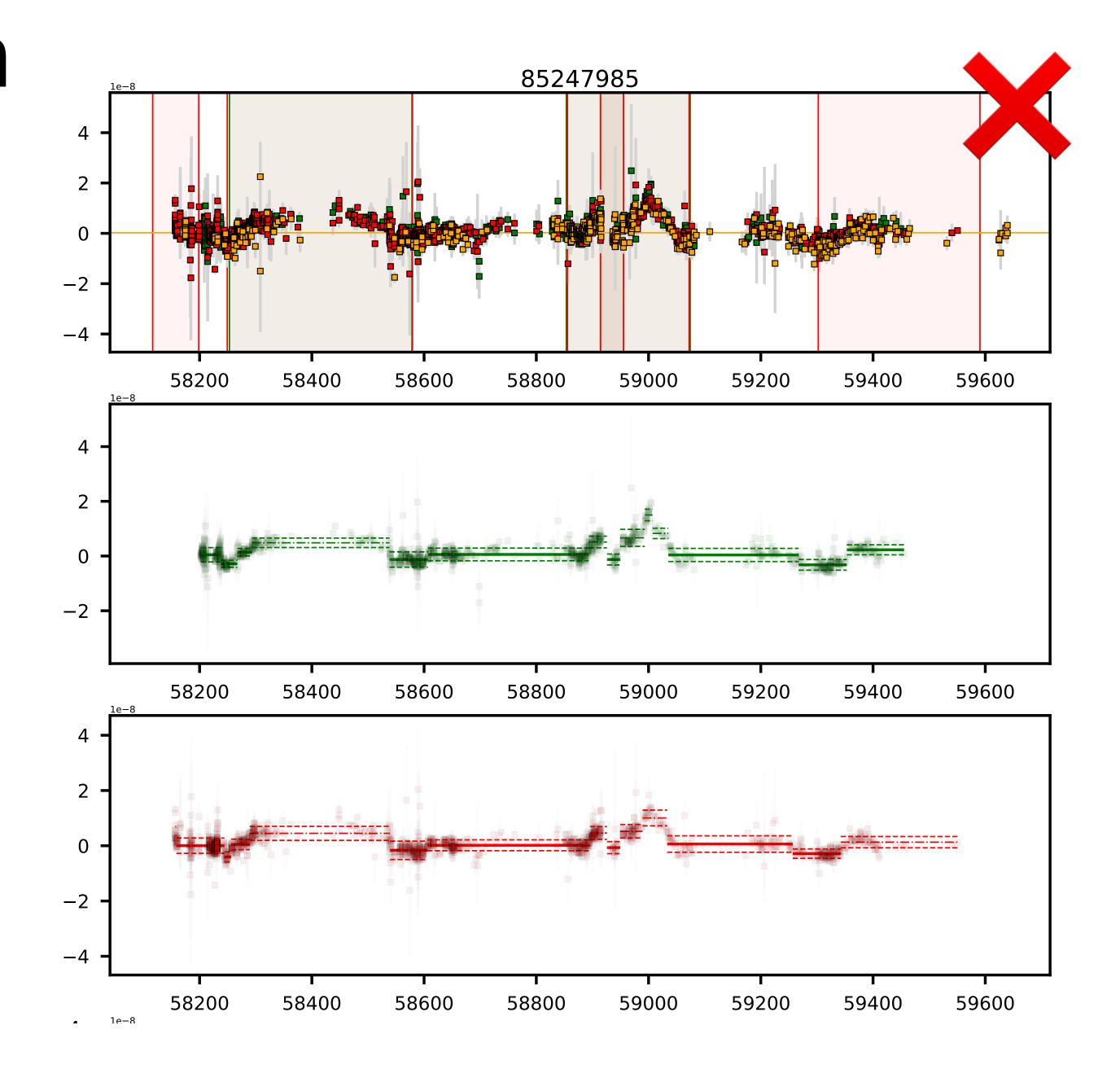
all bands

g-band



all bands

g-band



#### Features used (20 in total)

peak apparent mag

sgscore

WISE colors (with scatter for noisified child lightcurves)

SALT fit results  $(c, x_0, x_1)$ 

TDE fit results (rise, decay, temp, d\_temp)

distnr (core distance, scaled for noisified child lightcurves)

Bayesian block analysis: overlapping regions

no z (only indirectly for some of the SALT fits)

#### Noisification

K-correct (bandpasses see different flux depending on object redshift)

Redshift

Random dropout

Wiggle time a bit

Generate child lightcurves:

**SN** Ia: 12683 **TDE**: 10231 **SN**  $\neq$  Ia: 10852 **AGN**: 8220 star: 505

Children per object

**SN** Ia: 4 **TDE**: 155 **SN**  $\neq$  Ia: 10 **AGN**: 94 **star**: 0

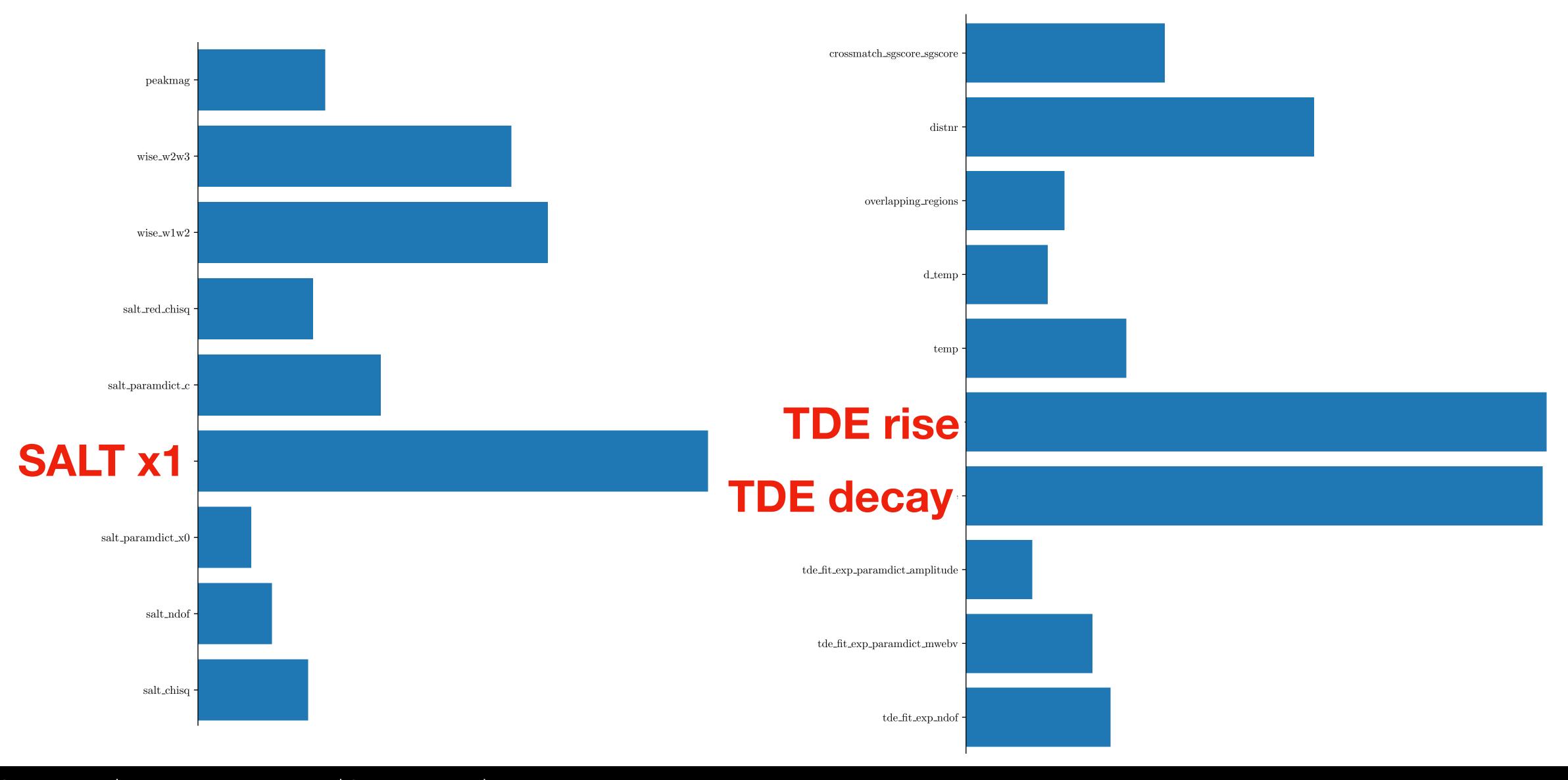
#### Train XGBoost

train-validation fraction = 0.7

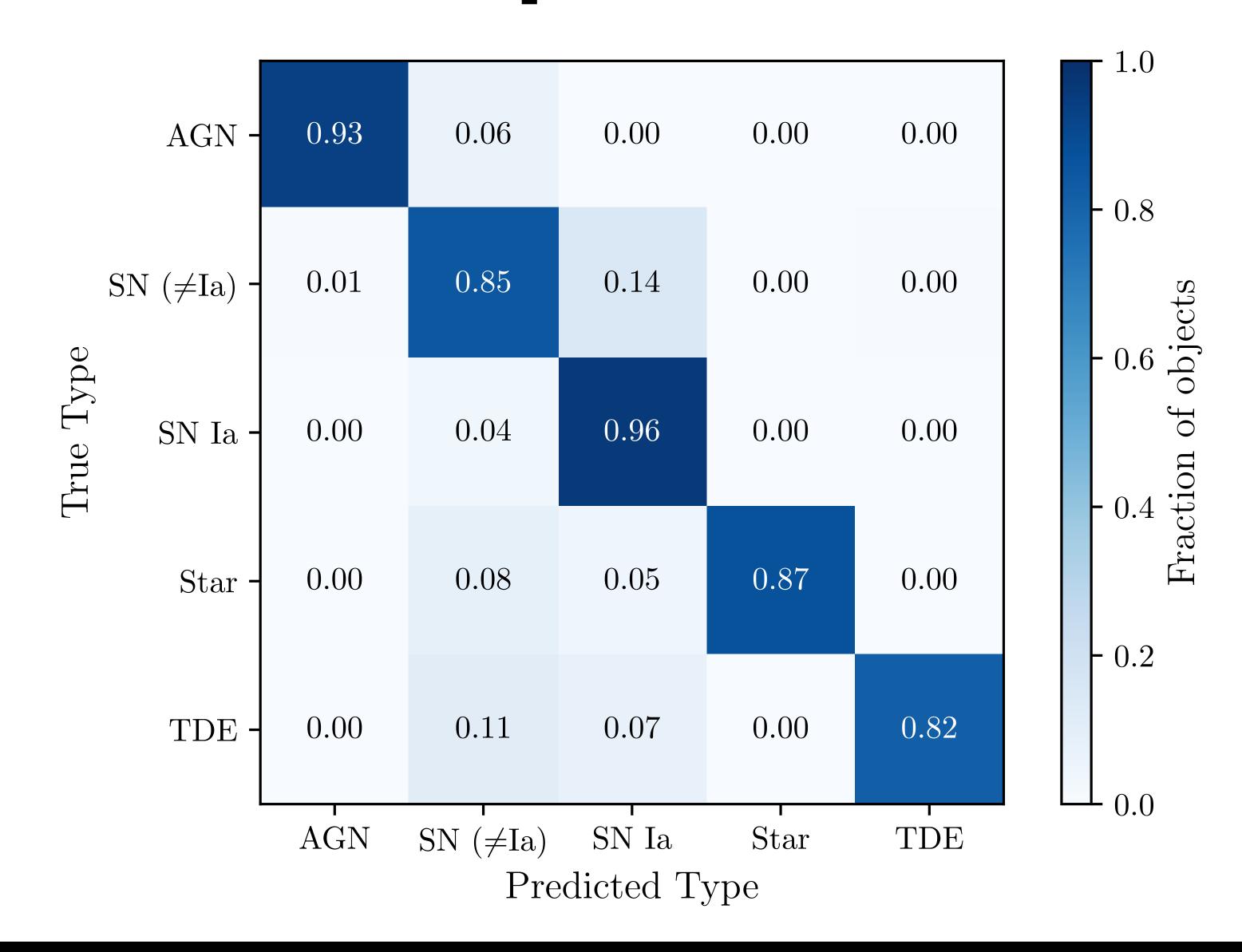
test fraction = 0.3

hyper parameter search (9 parameters) with 50 iterations

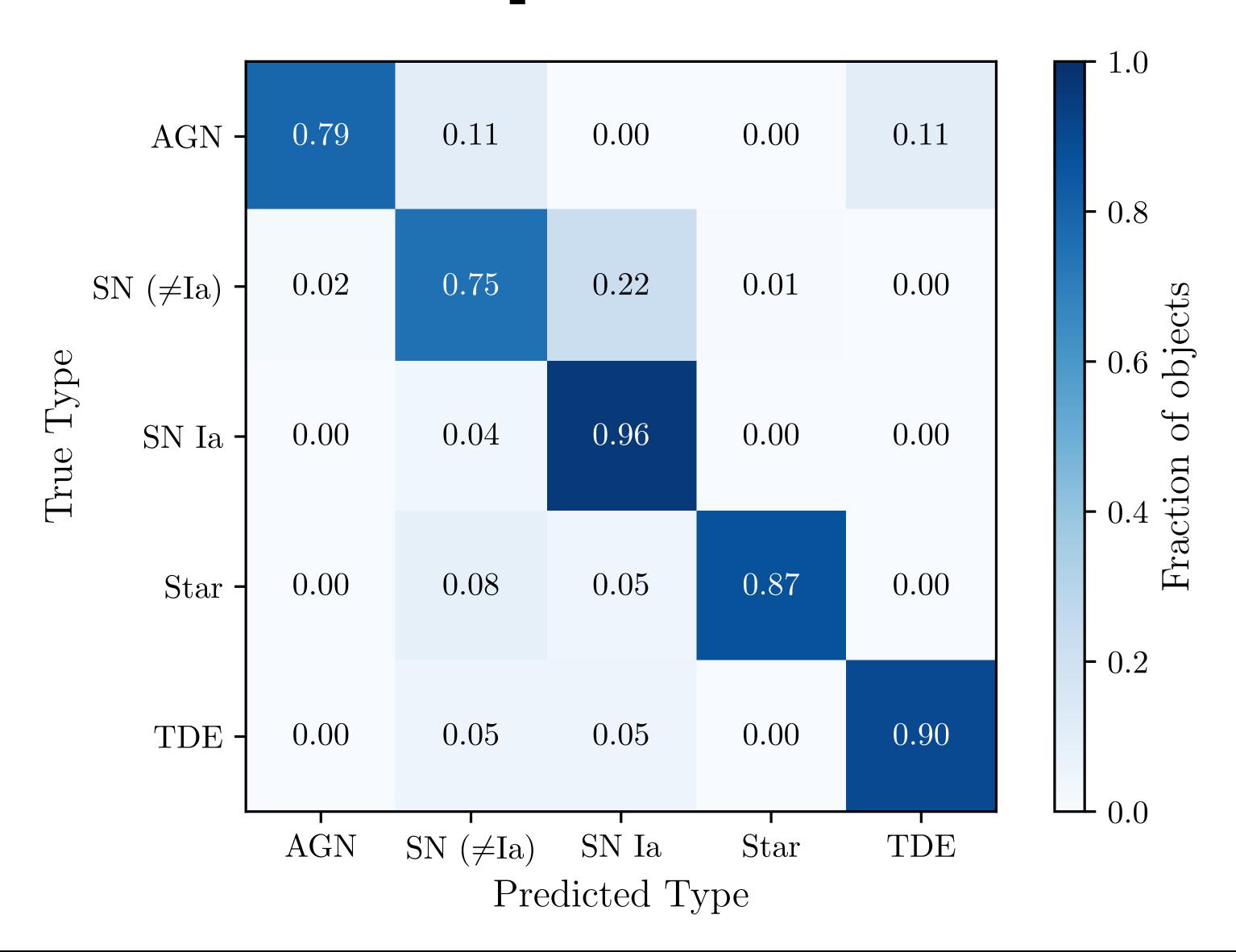
#### XGBoost feature importance



Including noisified lightcurves in the test sample

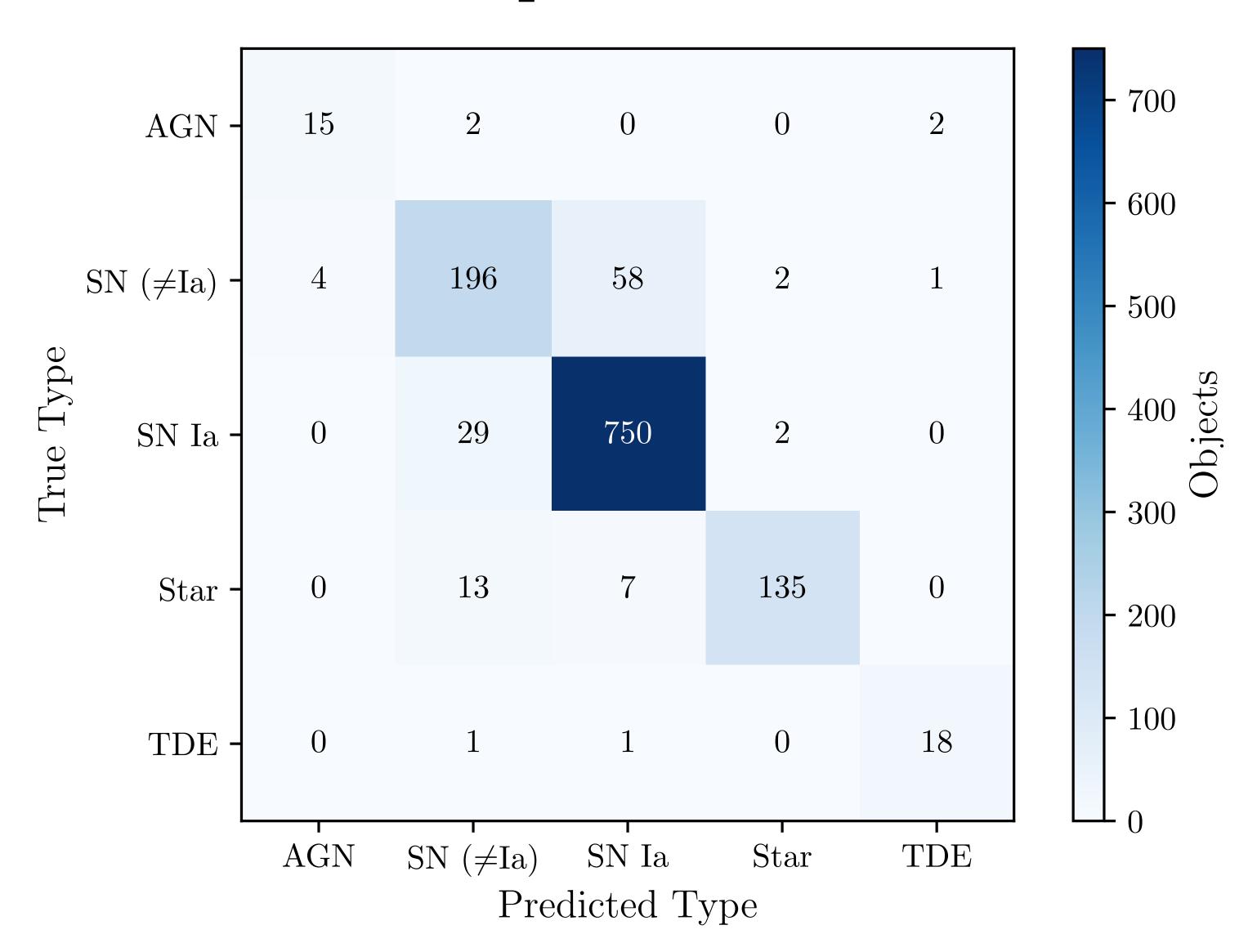


Not including noisified lightcurves in the test sample



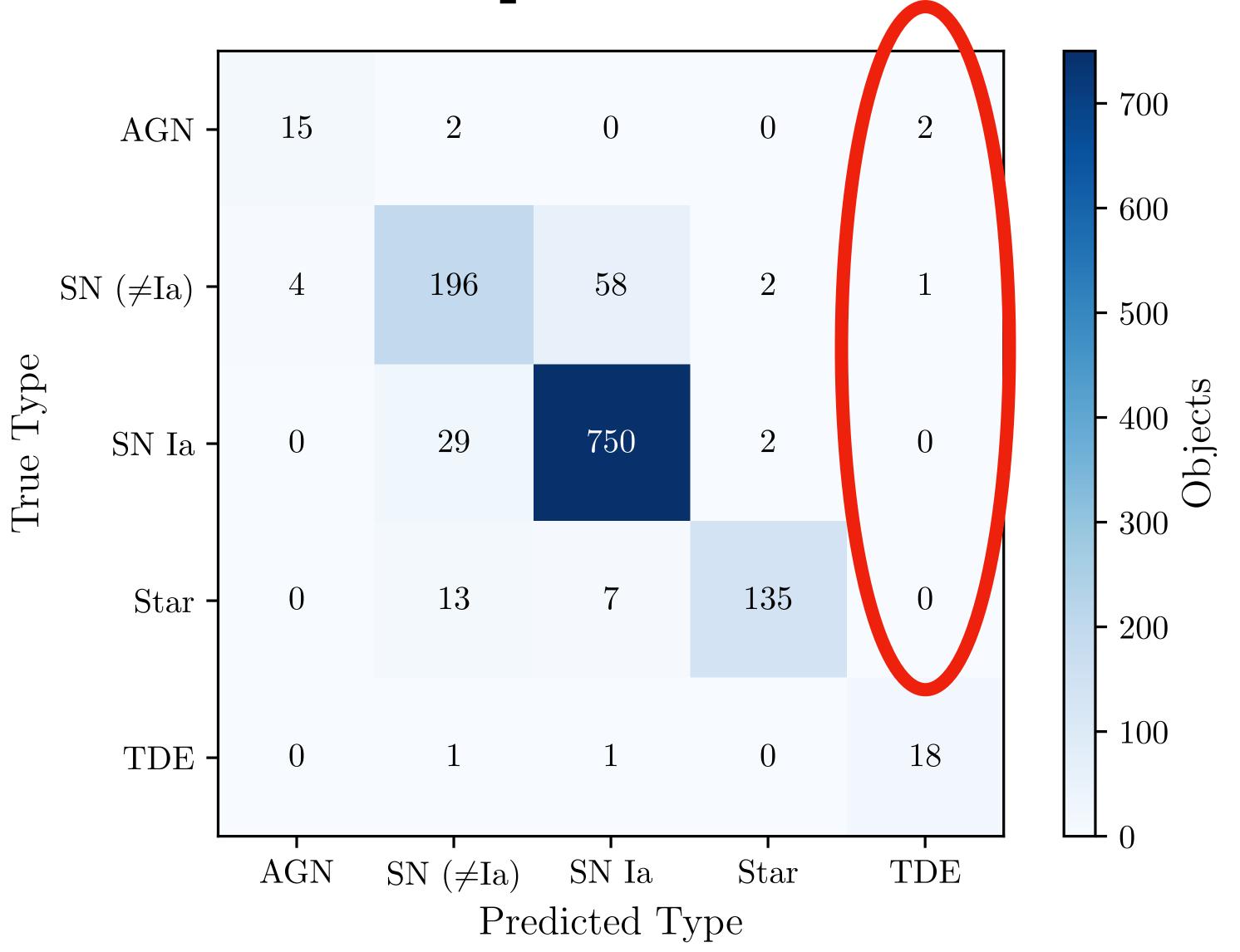
Not including noisified lightcurves in the test sample

absolute numbers



Not including noisified lightcurves in the test sample

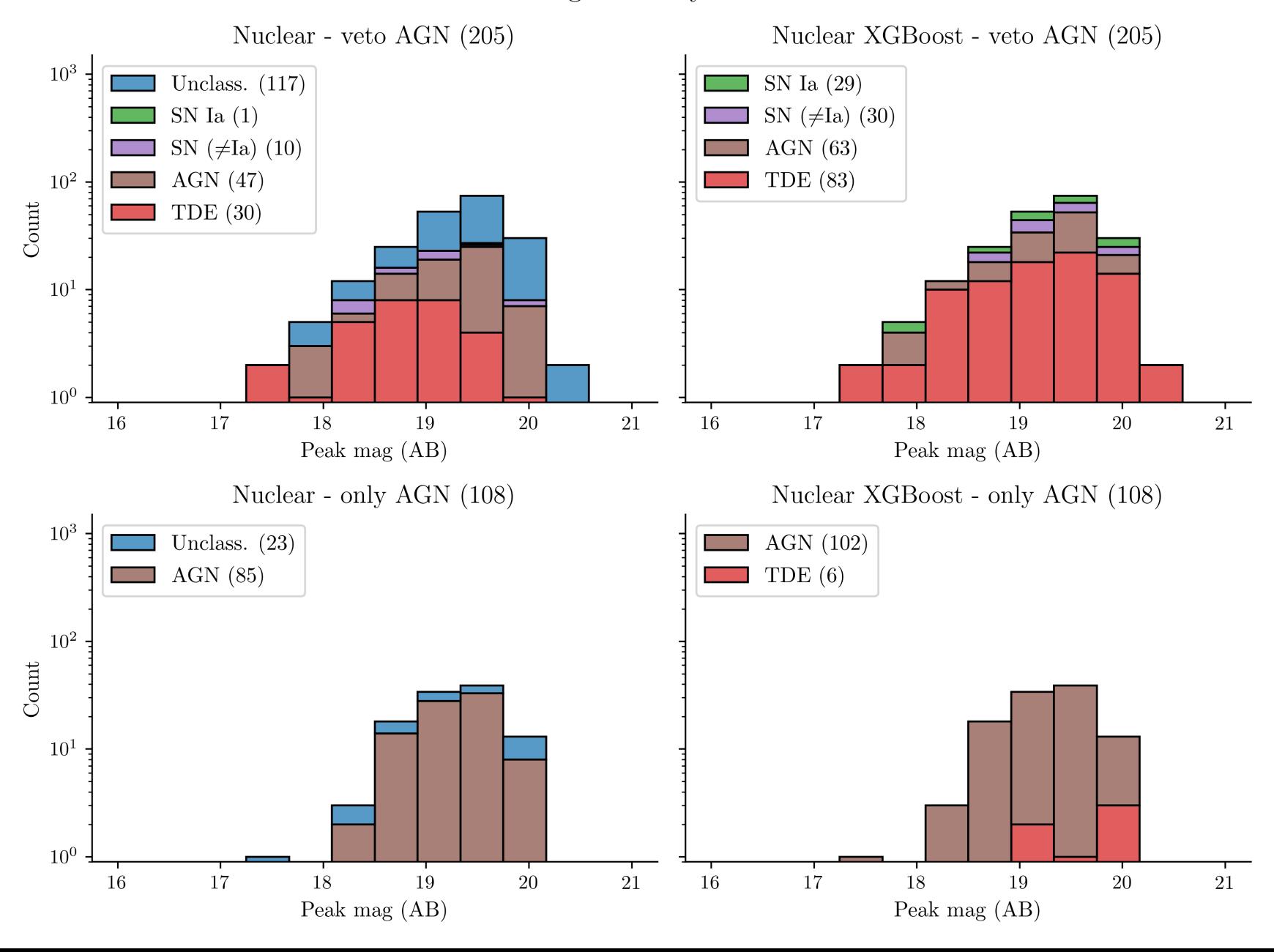
absolute numbers

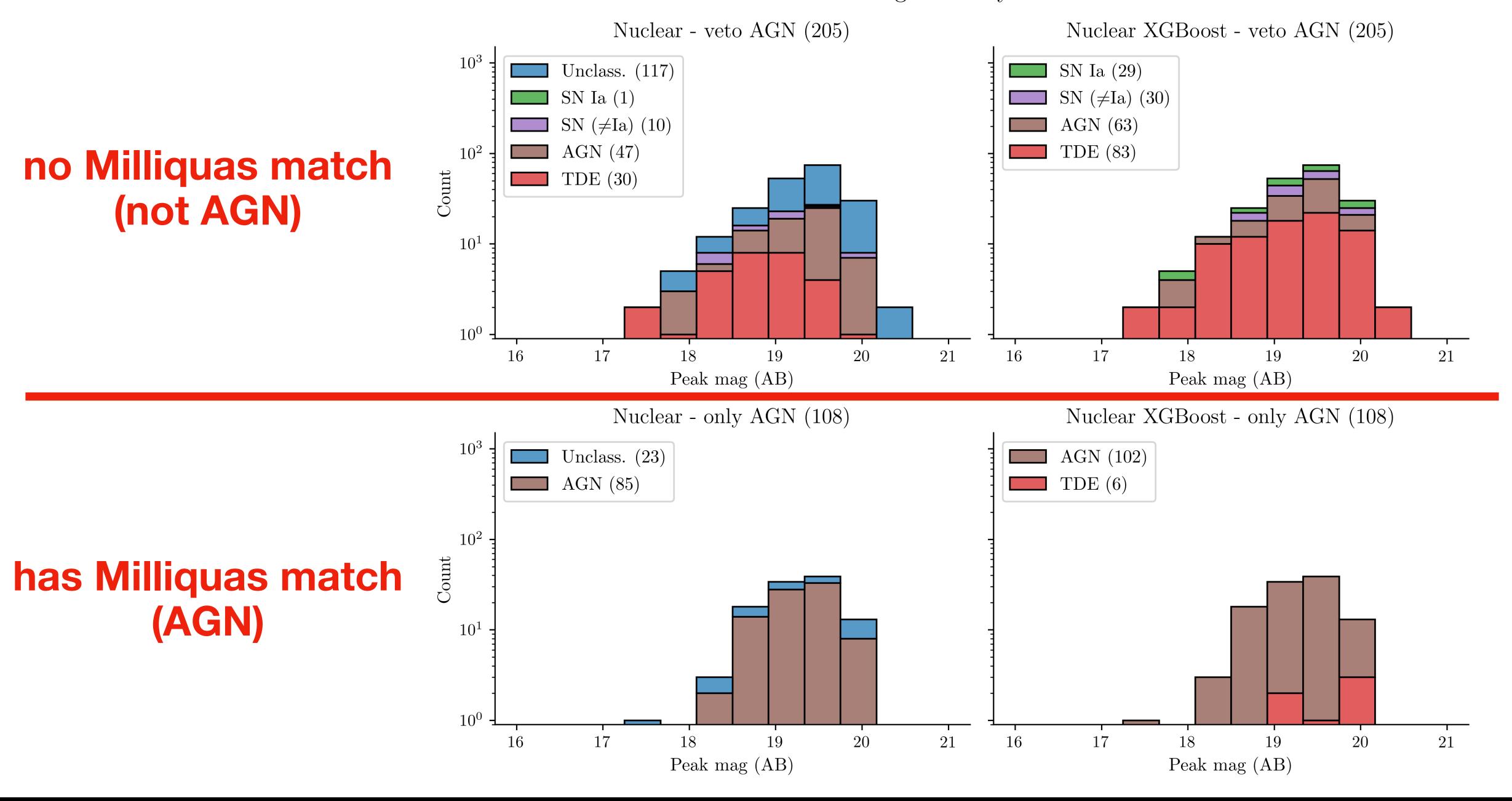


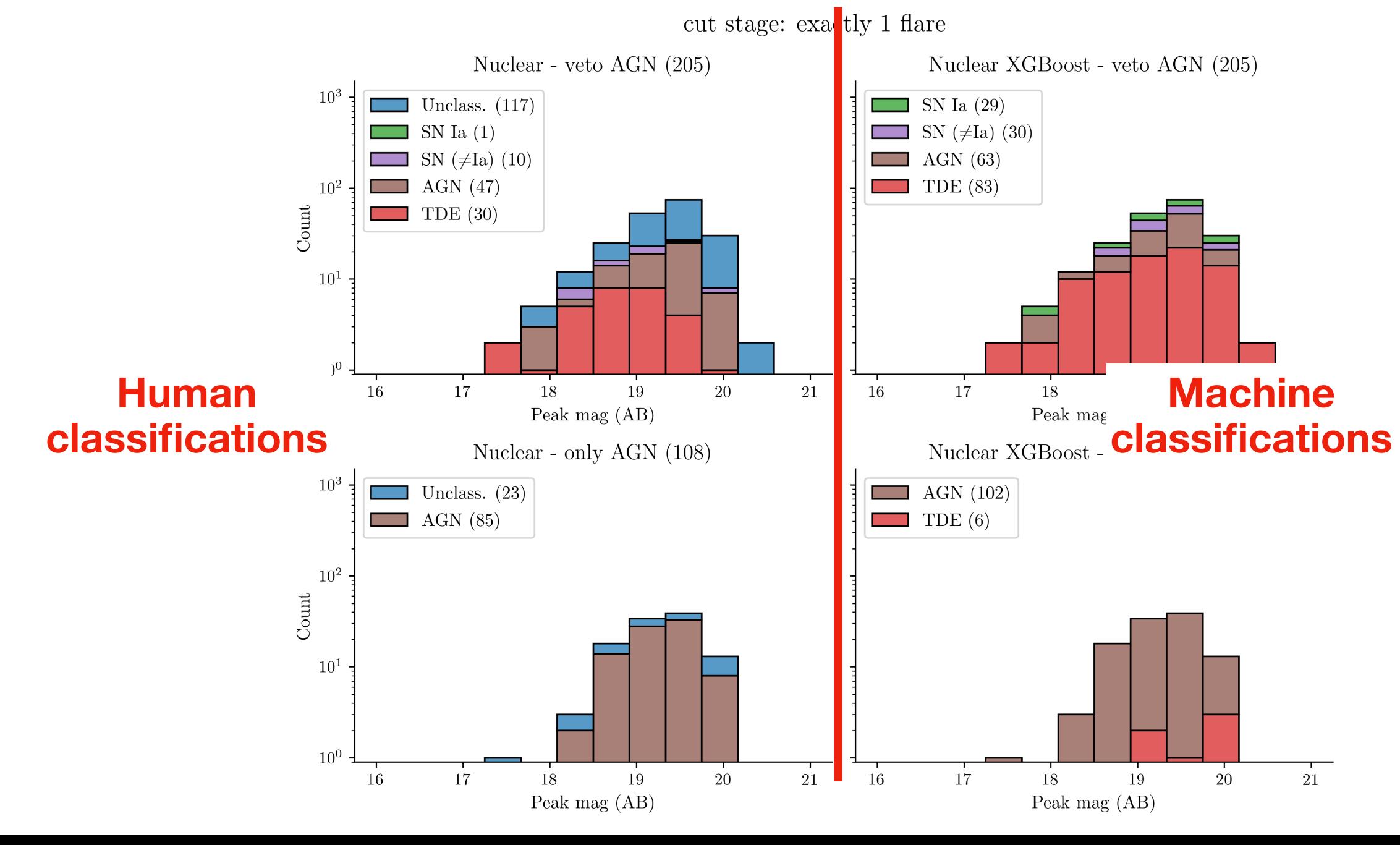
## No we apply the classifier to the nuclear sample

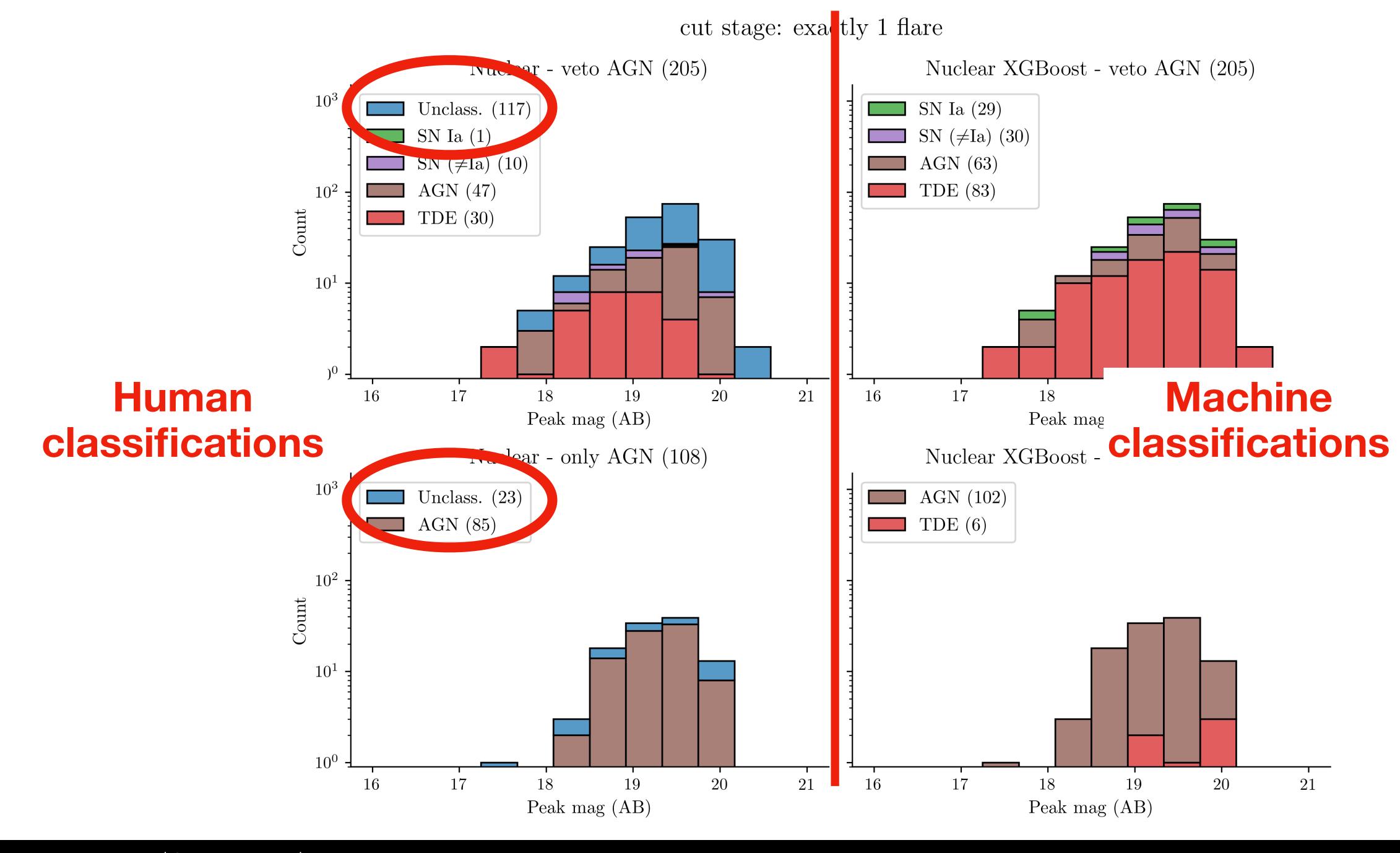
### after all cuts

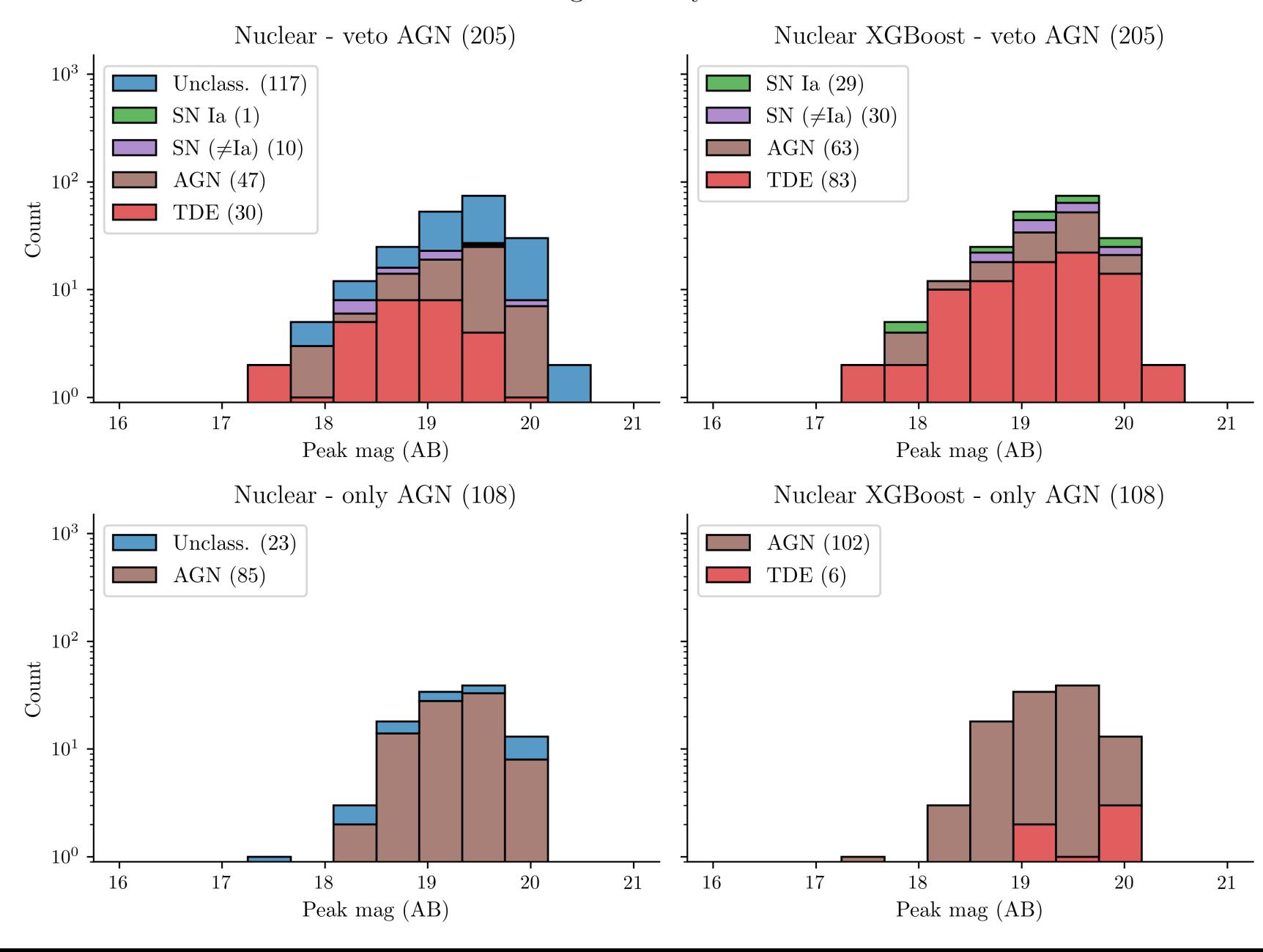
cut stage: exactly 1 flare



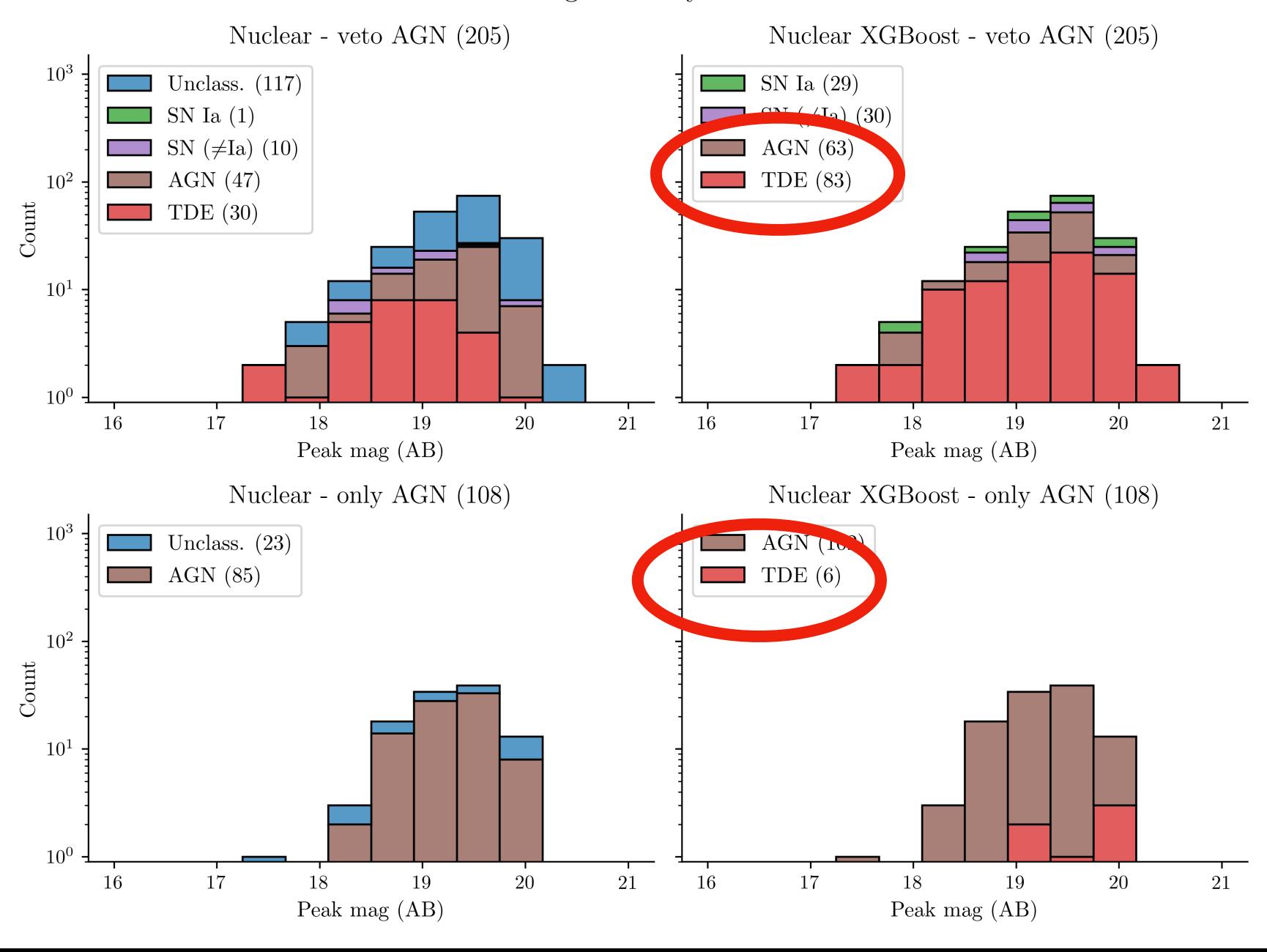








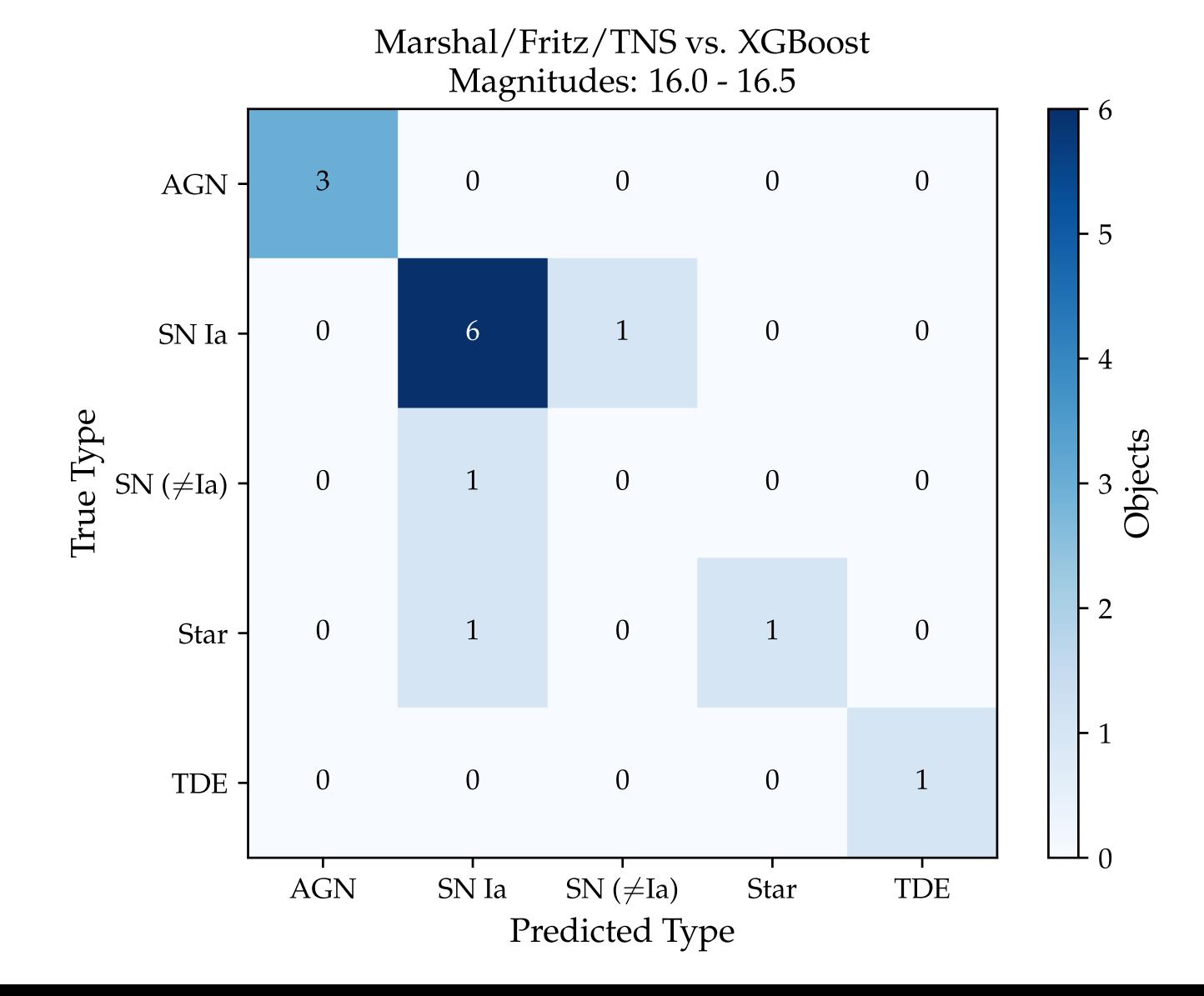
#### cut stage: exactly 1 flare



# Compare to unsecure classifications pulled from the internet

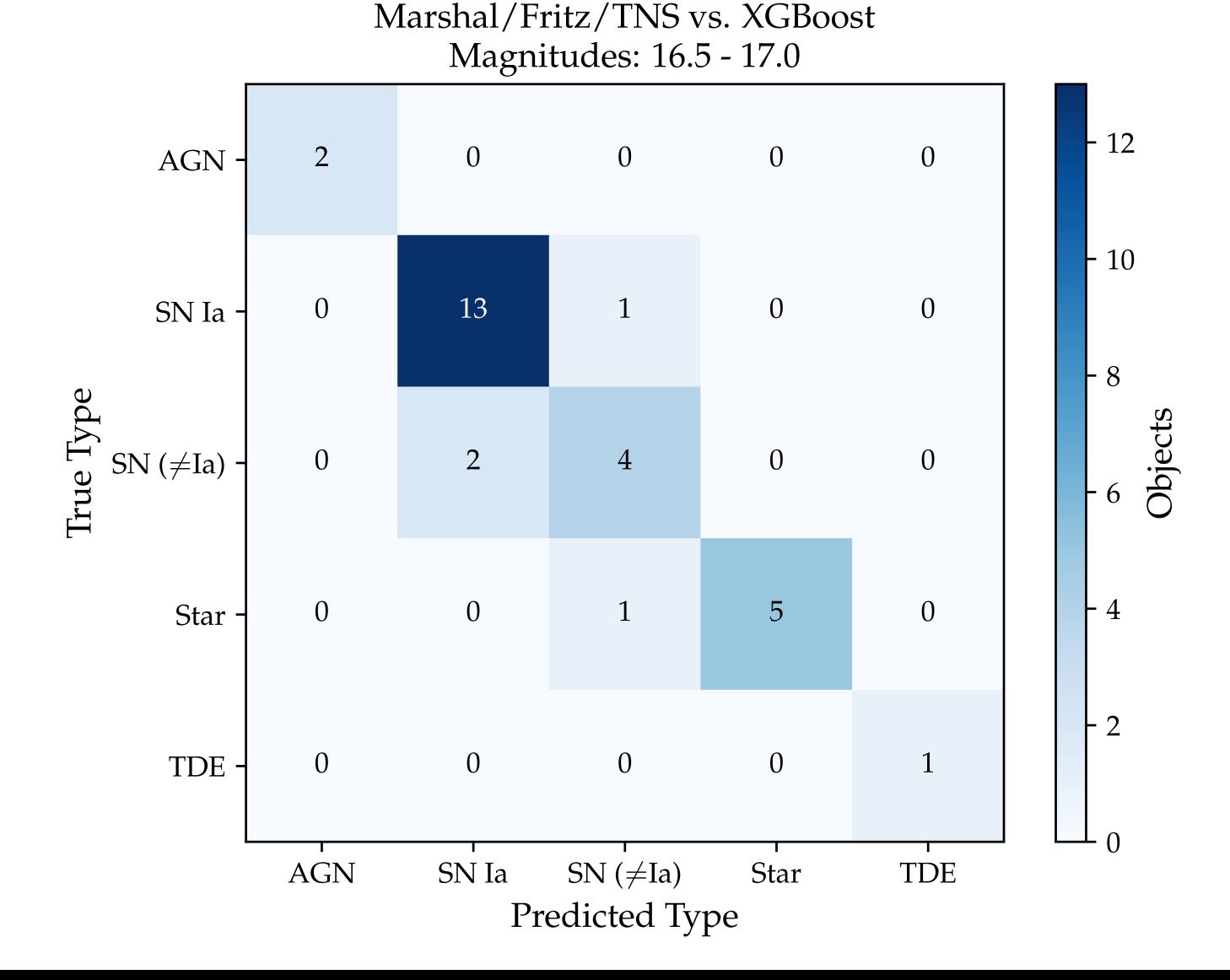
Mag:

16.0-16.5



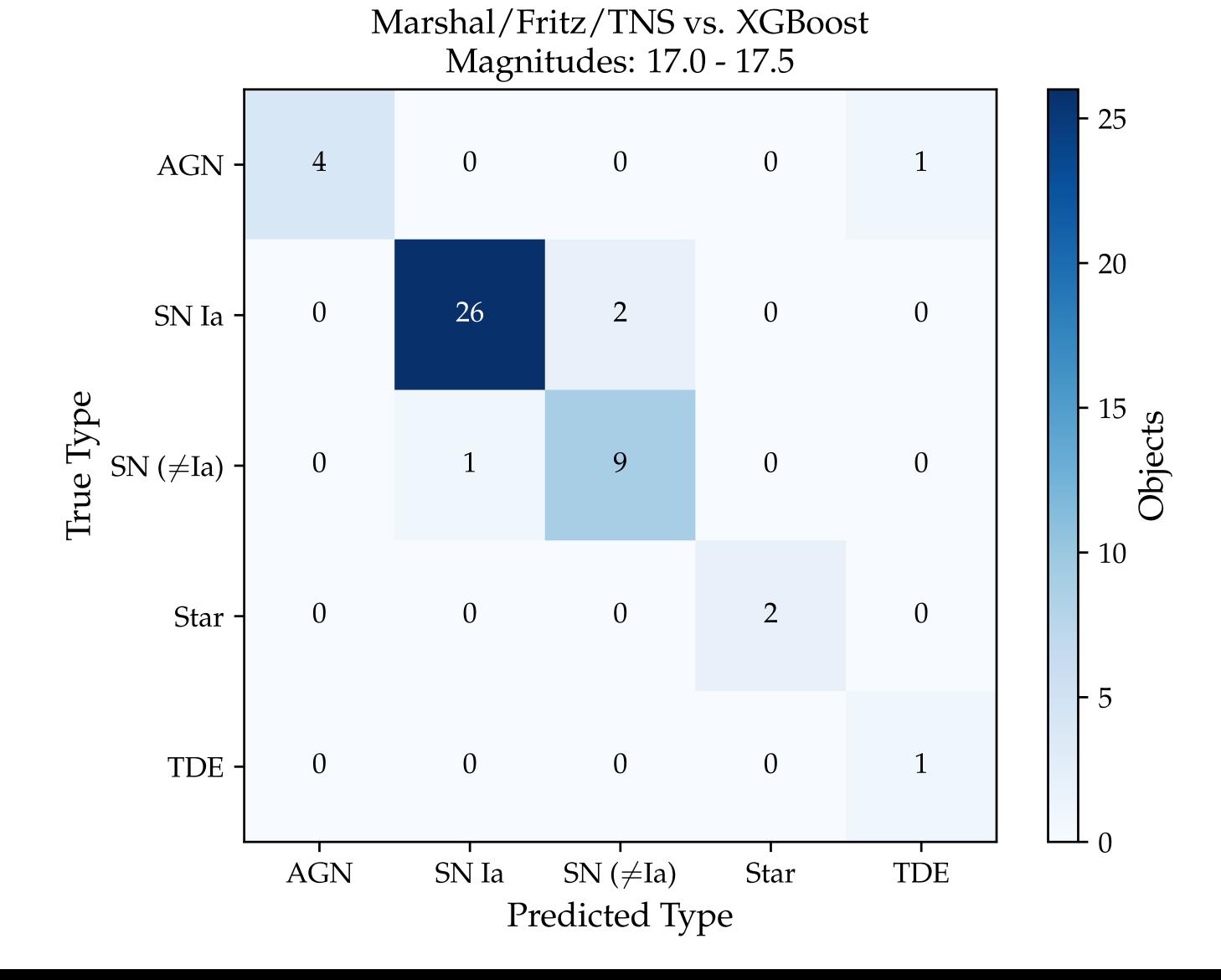
Mag:

16.5-17.0



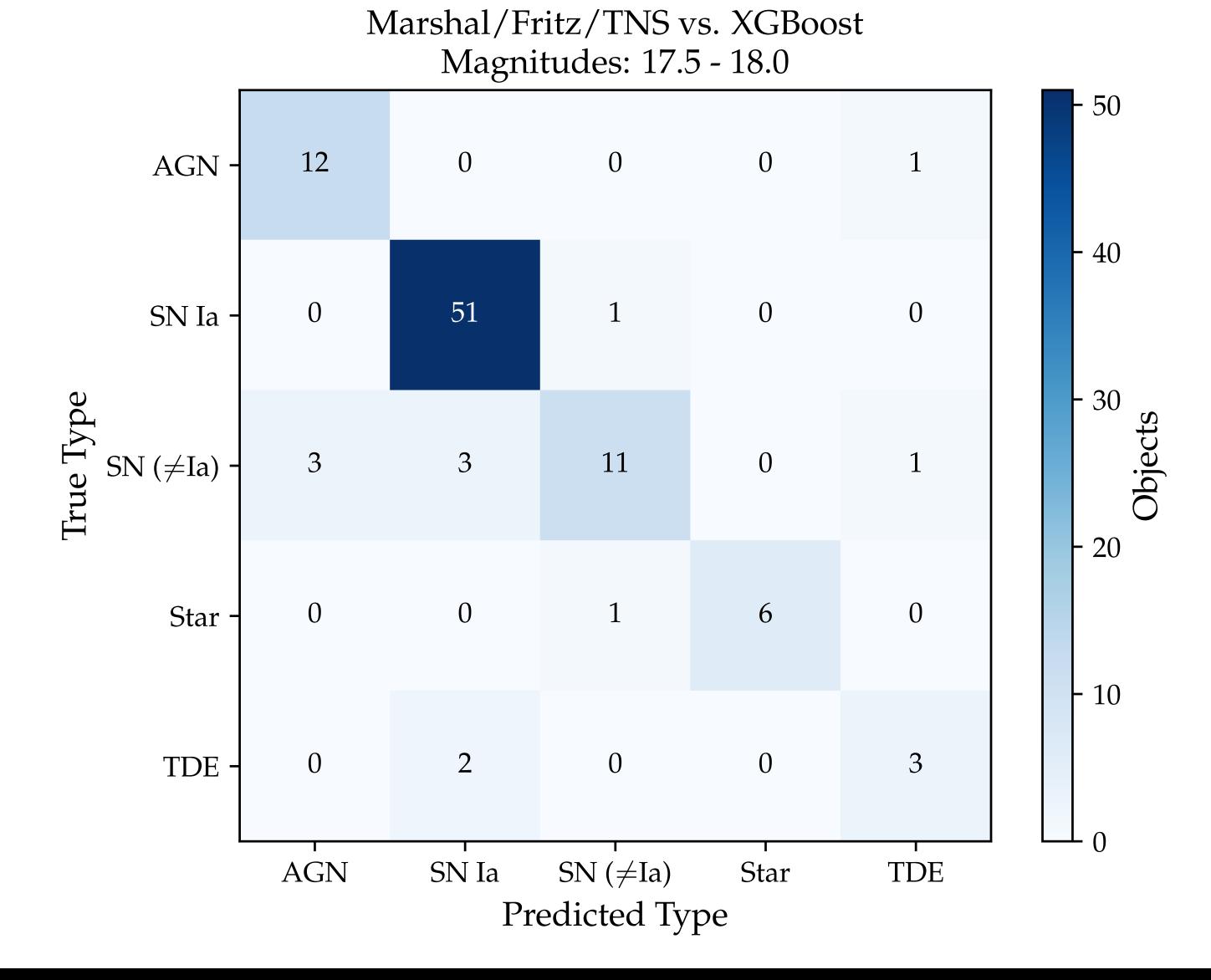
Mag:

17.0-17.5



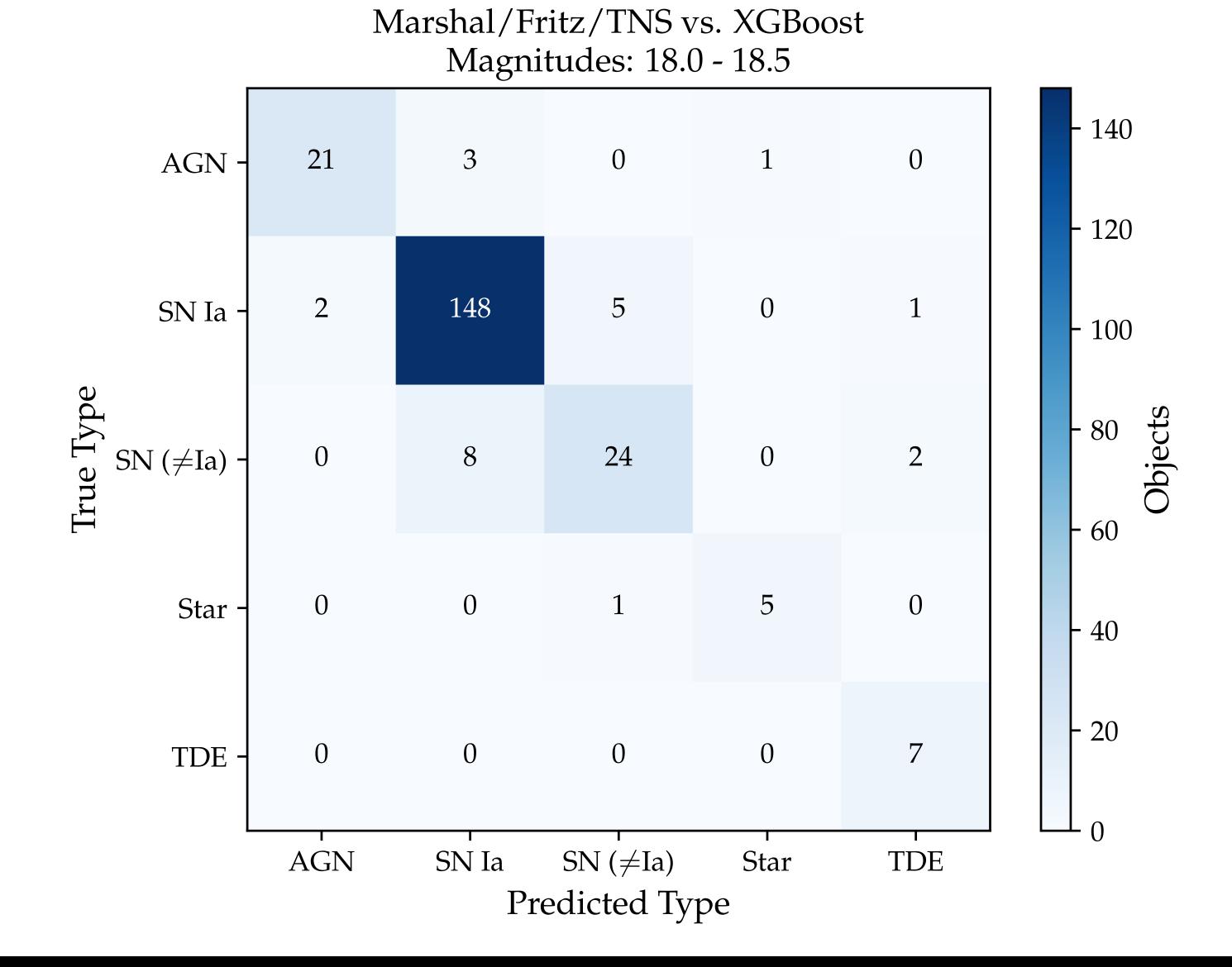
Mag:

17.5-18.0



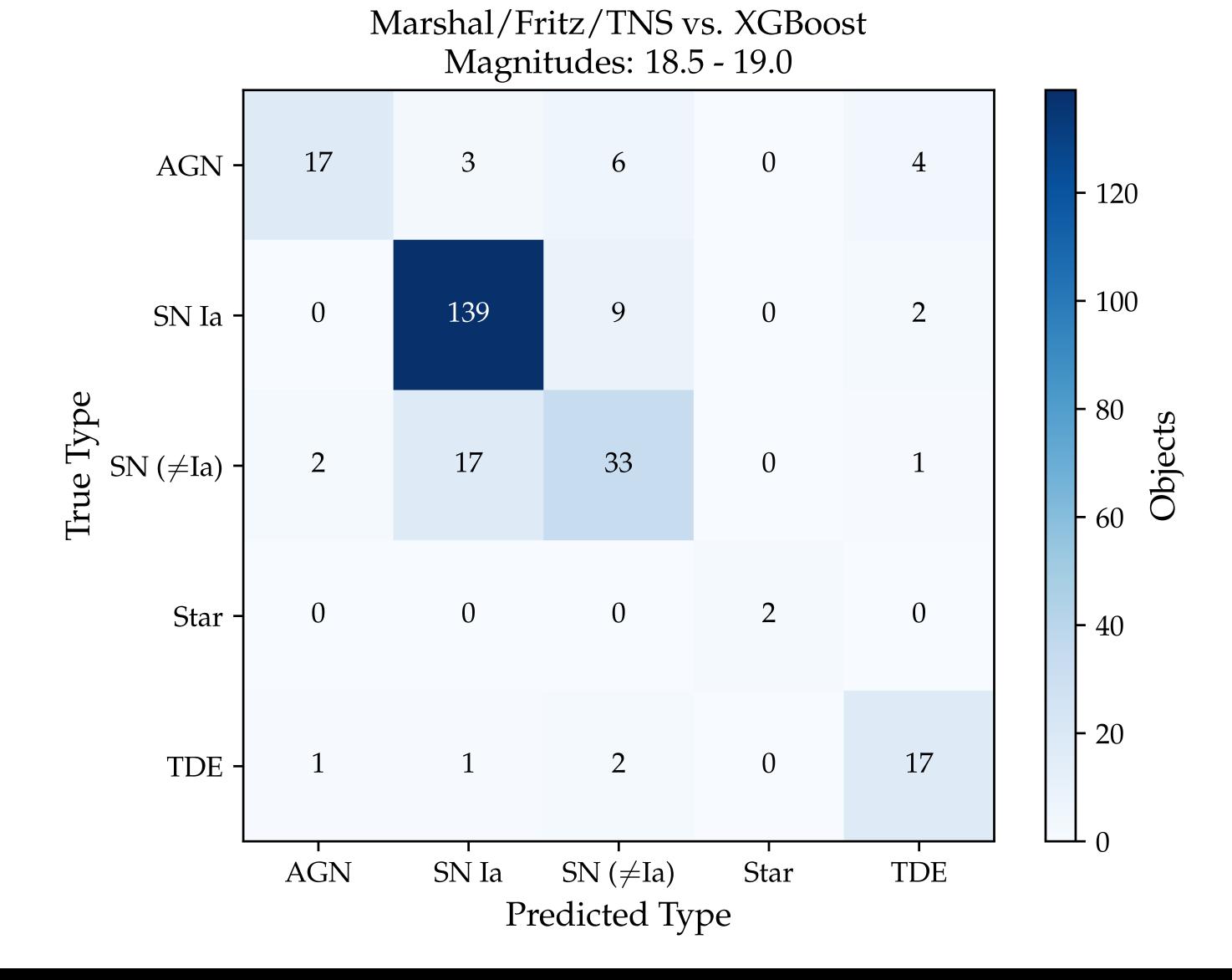
Mag:

18.0-18.5



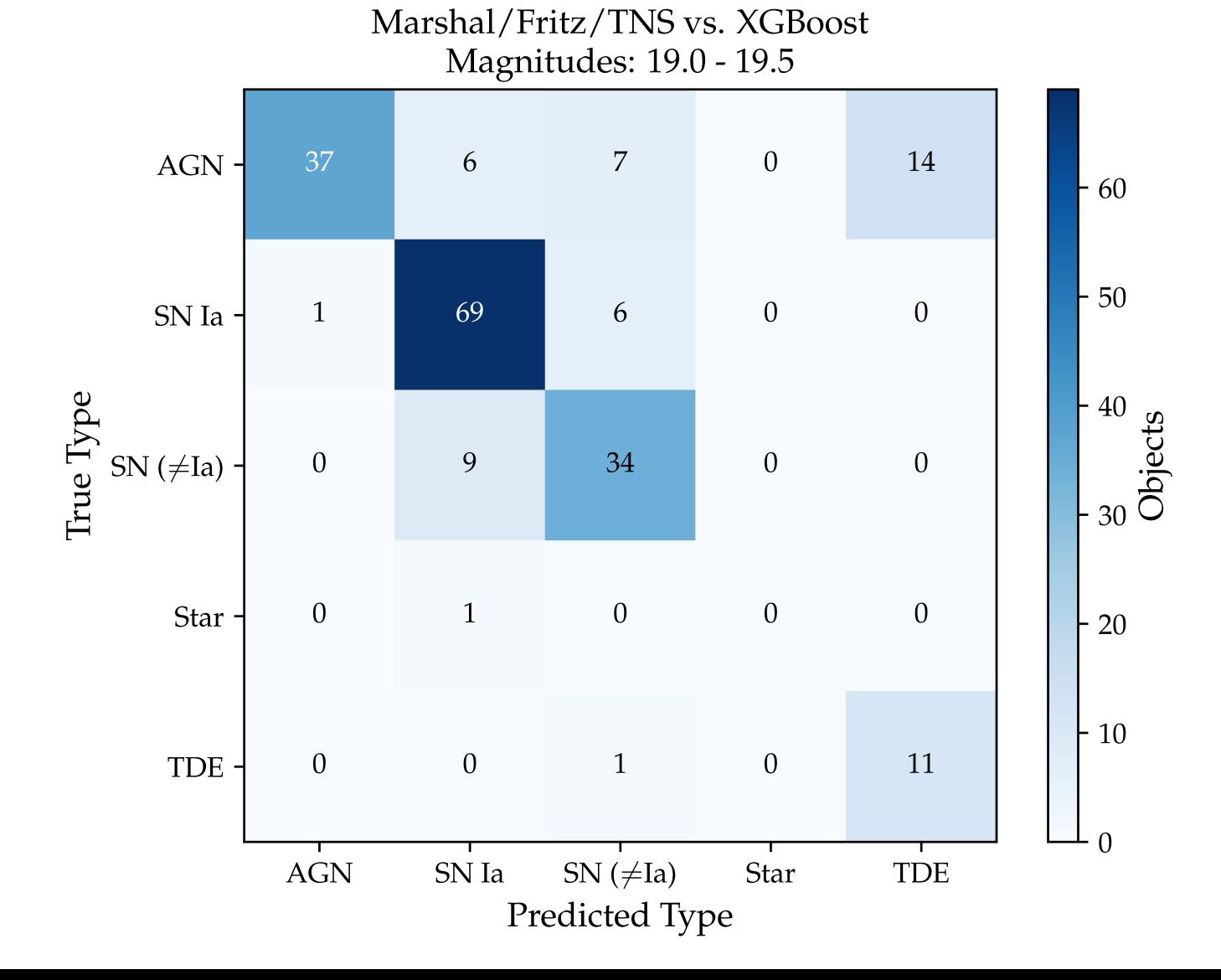
Mag:

18.5-19.0



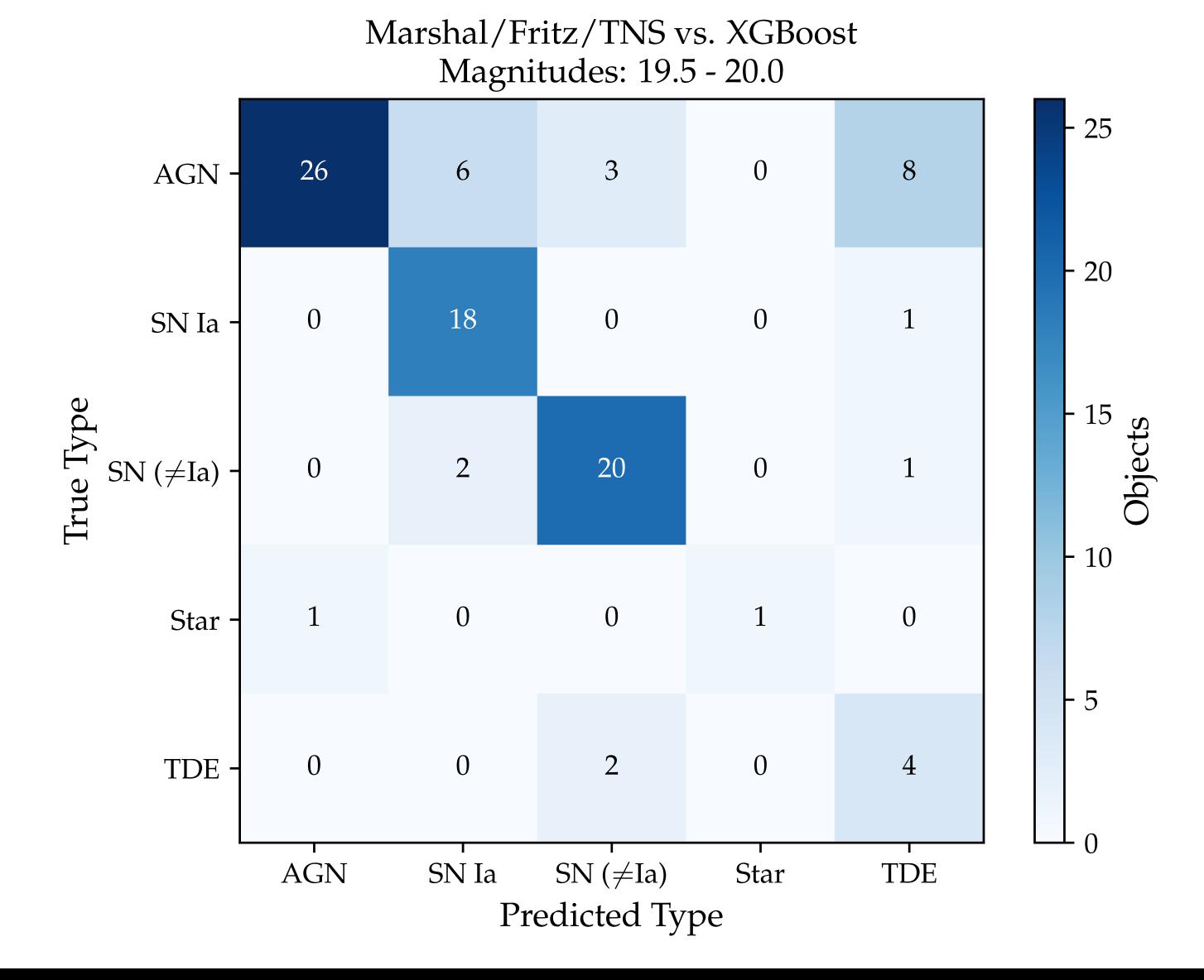
Mag:

19.0-19.5



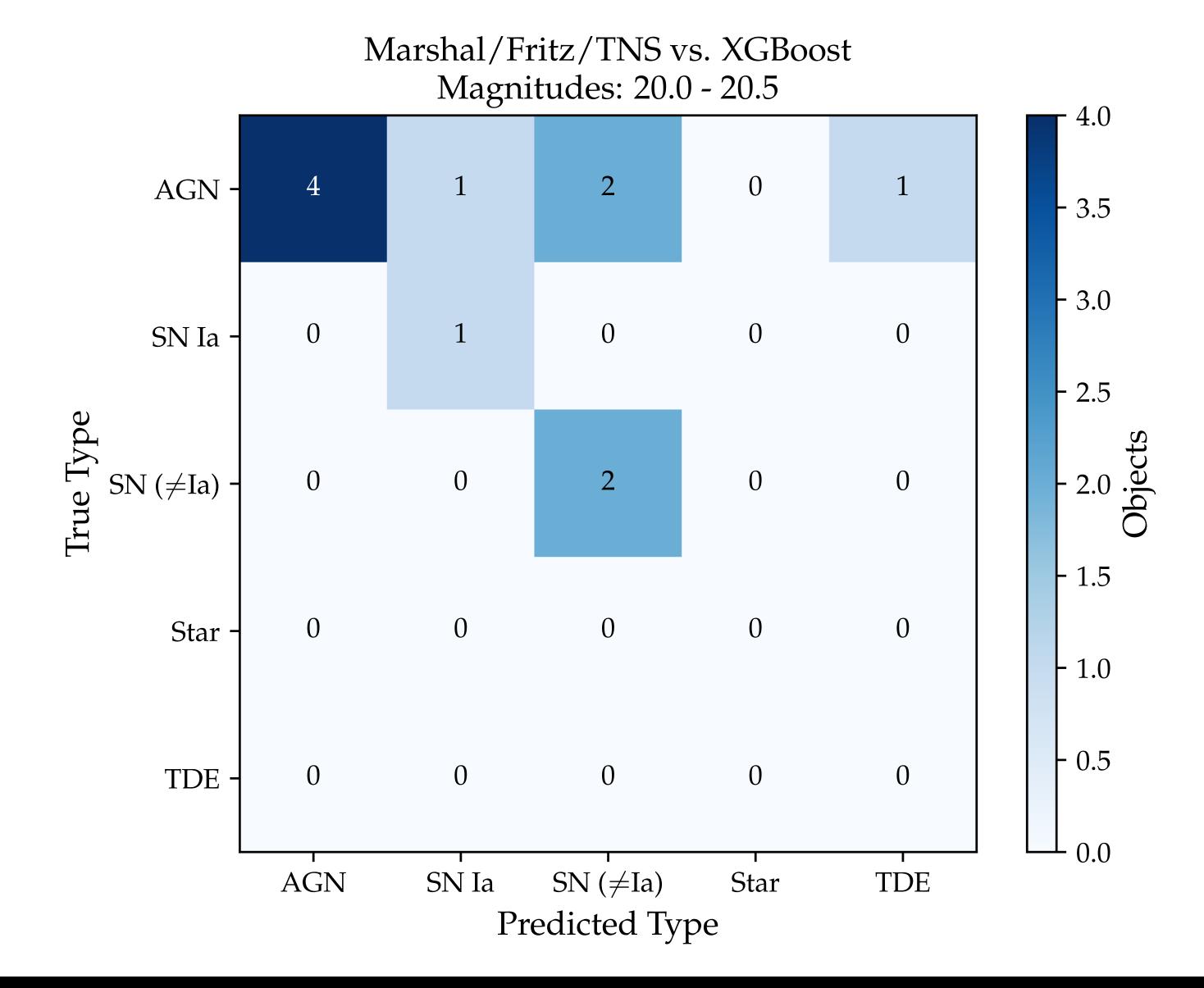
Mag:

19.5-20.0



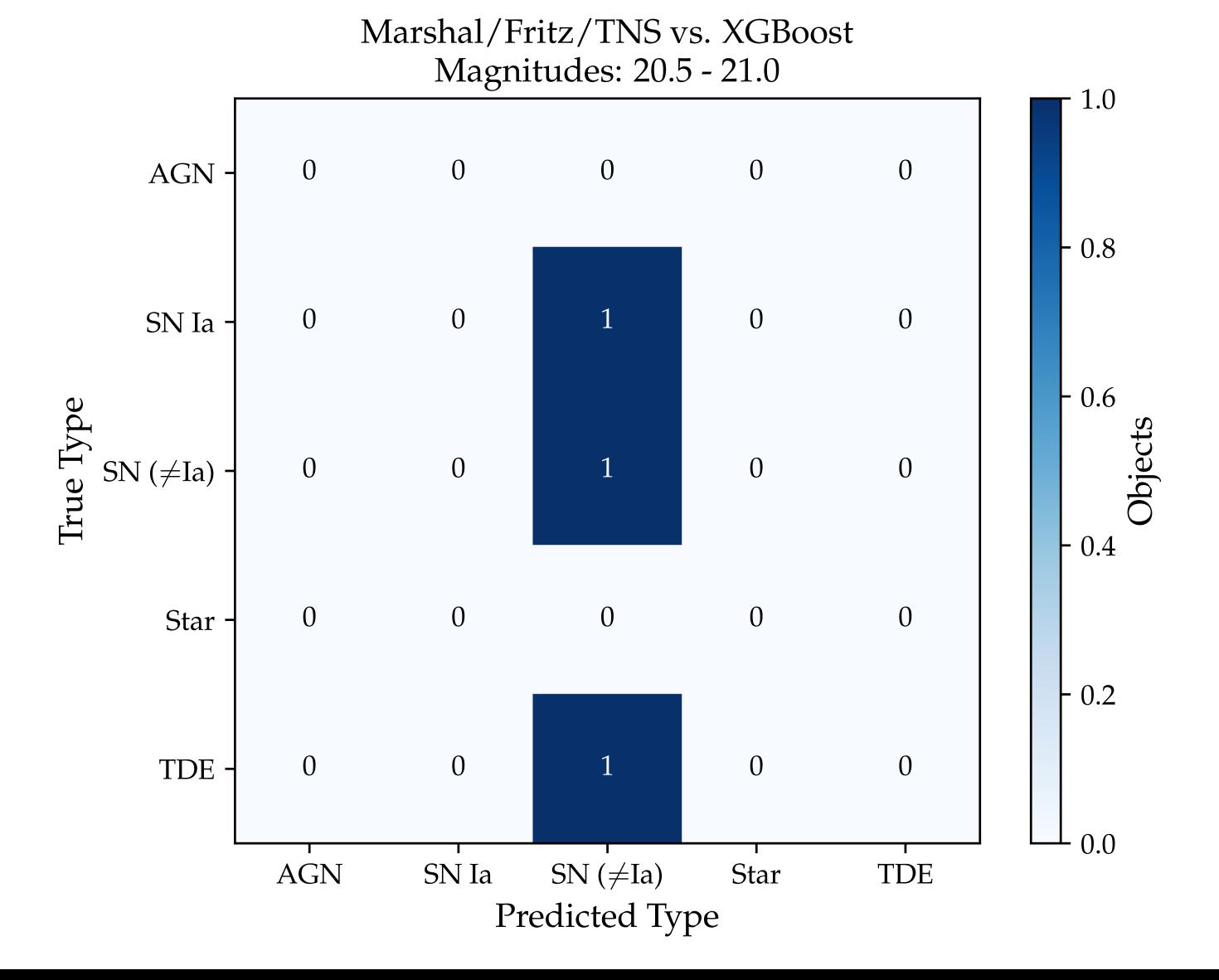
Mag:

20.0-20.5



Mag:

20.5-21.0



#### Questions

How to address uncertain classifications?

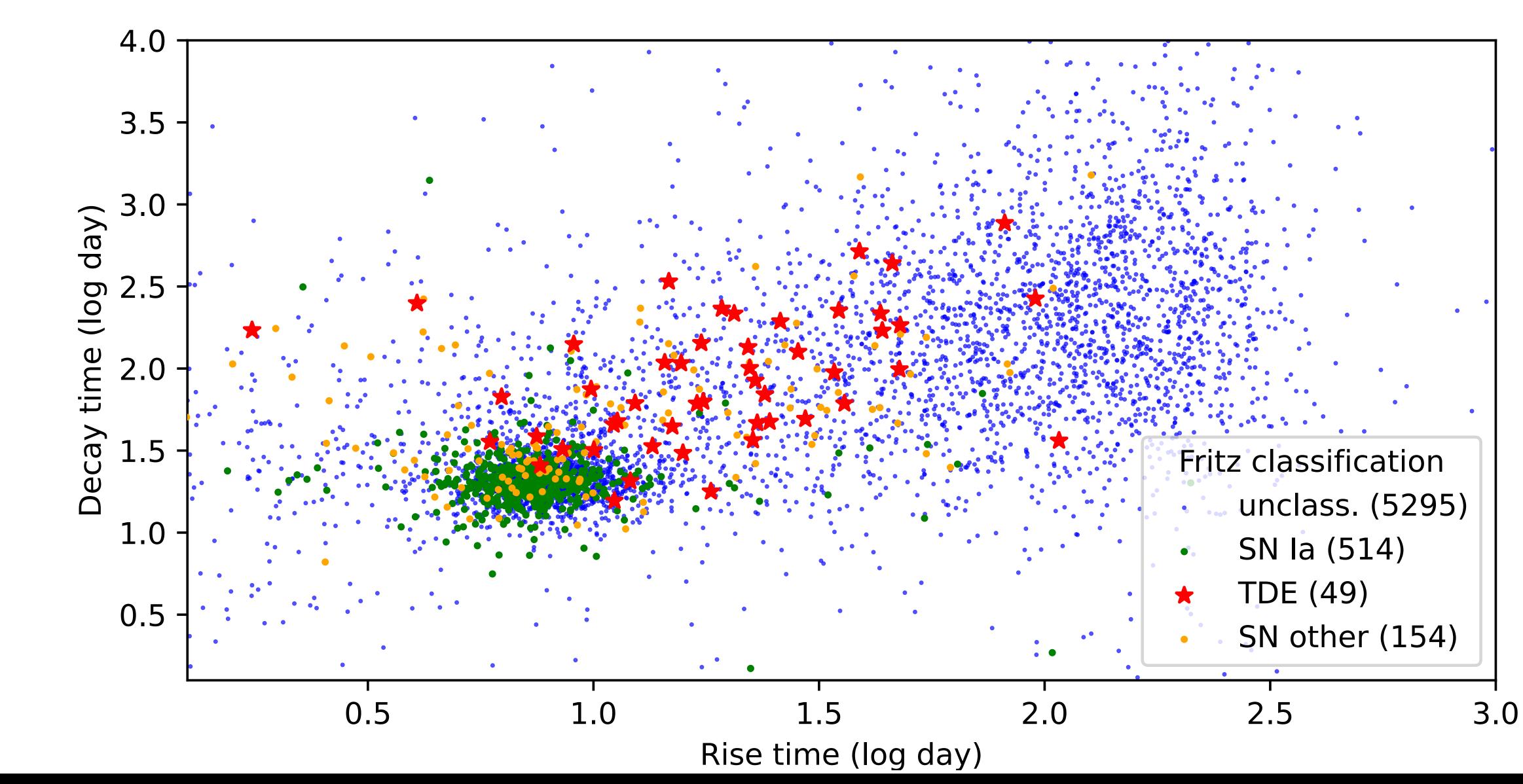
Way to discard "bad" predictions?

Waterfall plots?

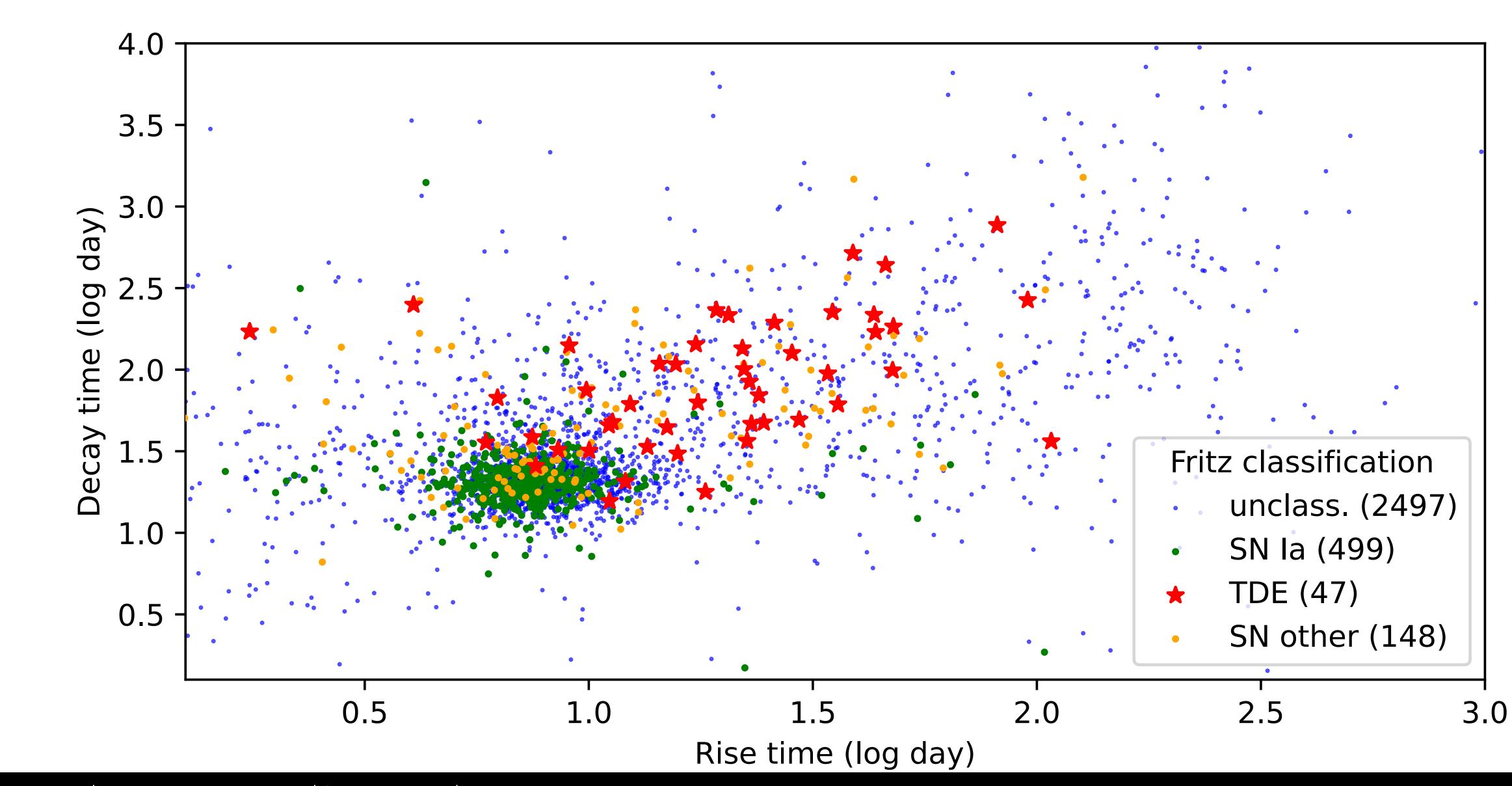
Suggestions to improve on the training? E.g. normalization of features (numerical values)?

Anomaly detection?

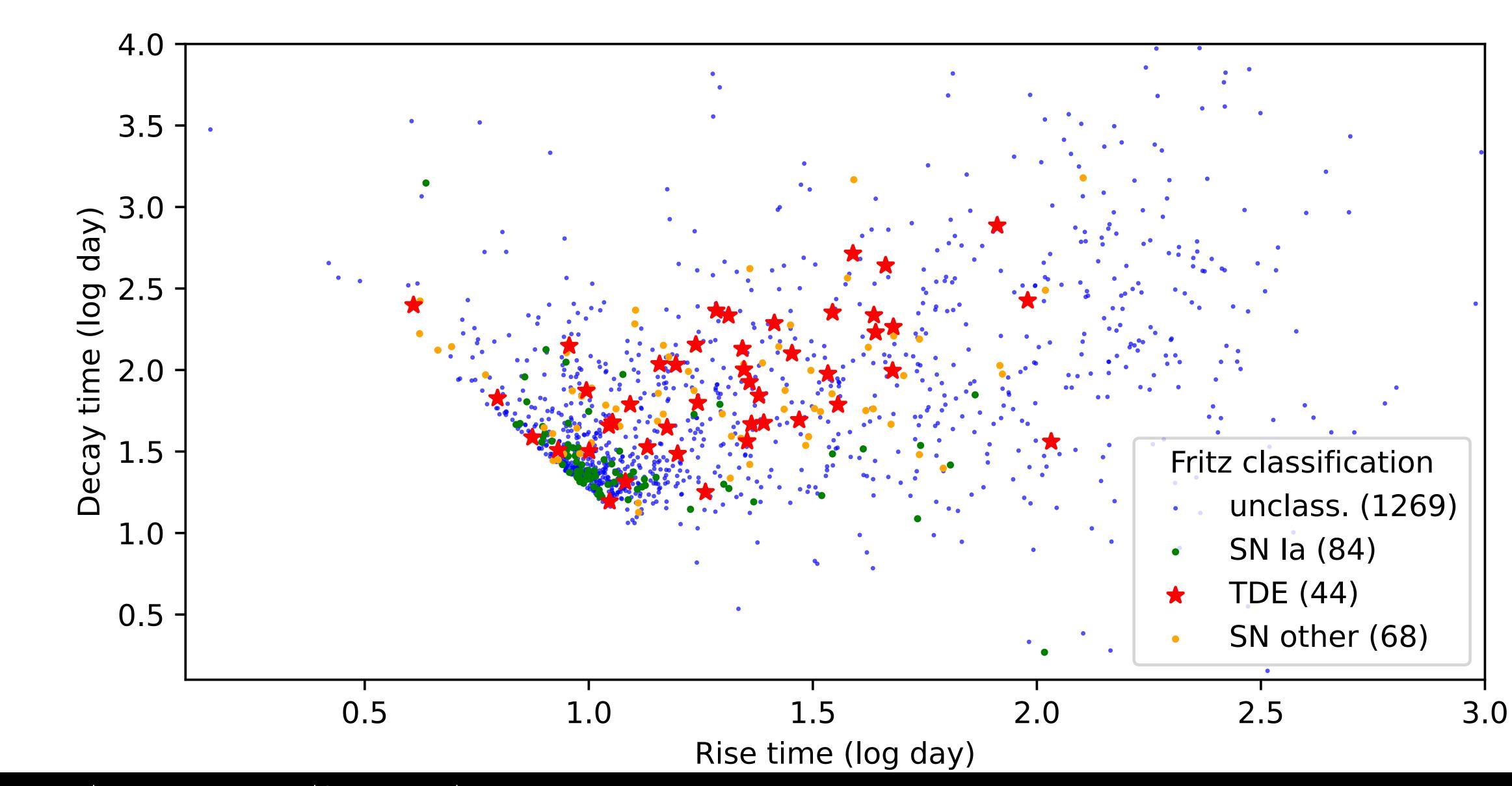
No cut (6012 transients) Purity: 0.8% / Efficiency: 100.0 %



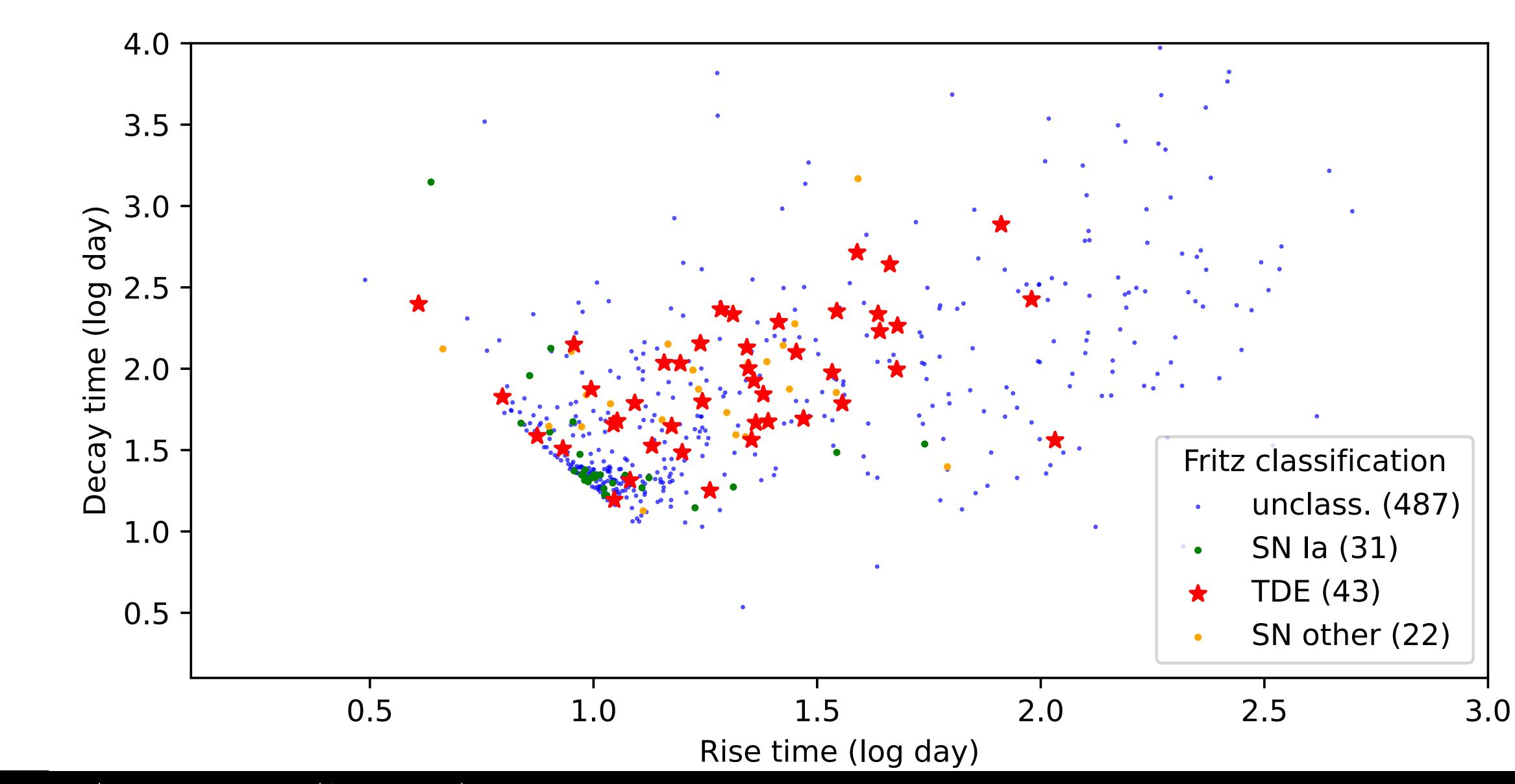
#### WISE colors (3191 transients) Purity: 1.5% / Efficiency: 95.9 %



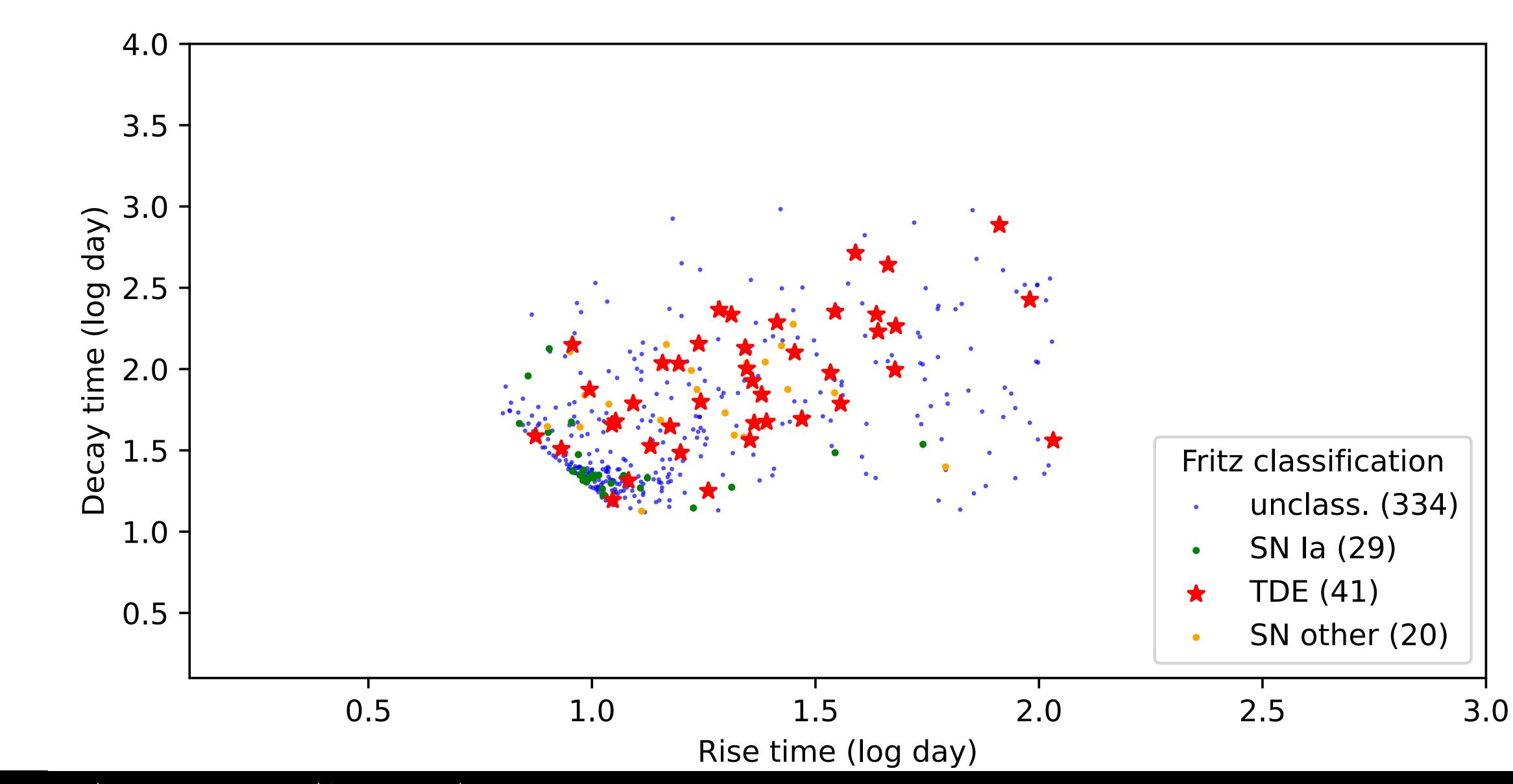
#### Diagonal cut (1465 transients) Purity: 3.0% / Efficiency: 89.8 %



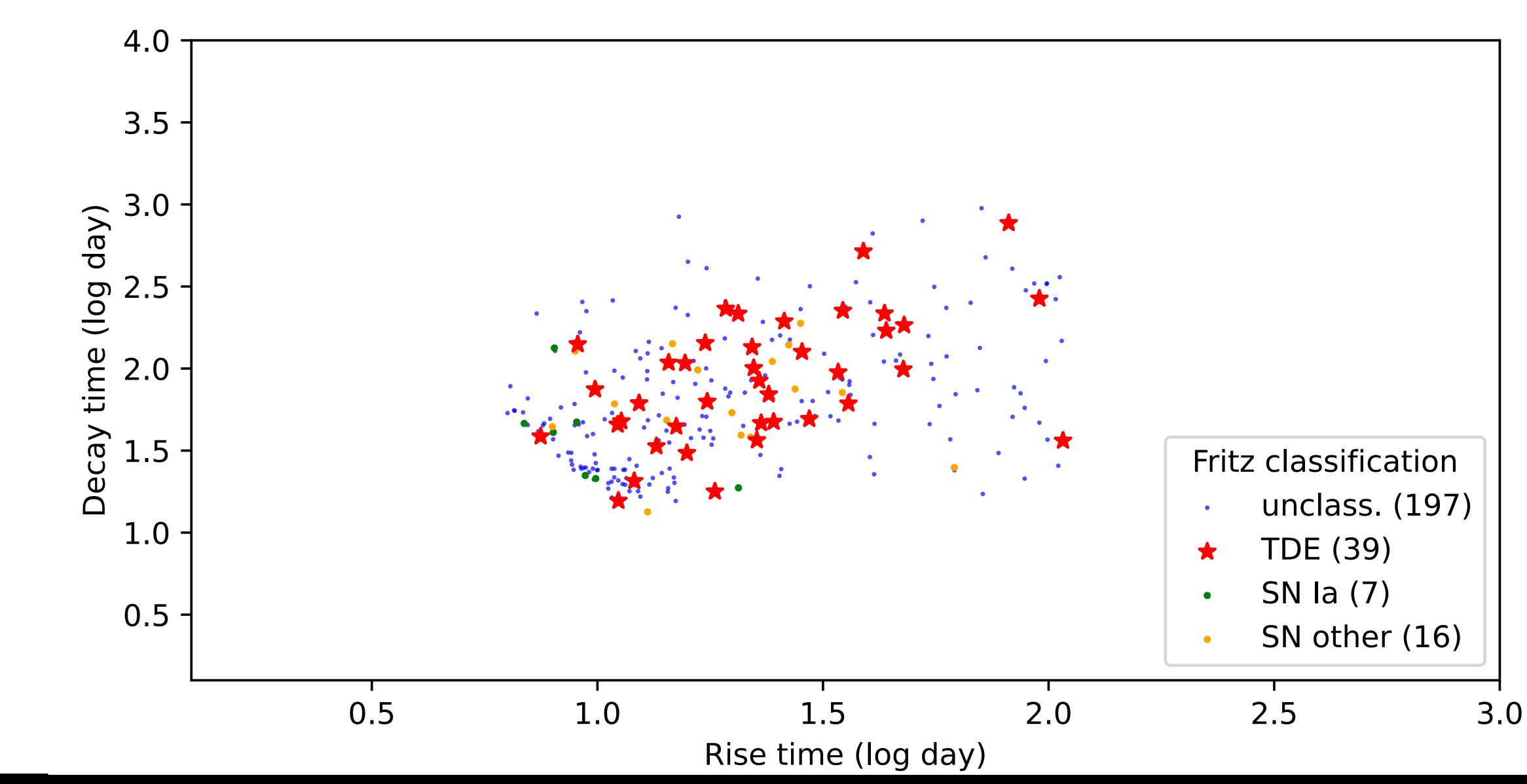
#### Temperature cut (583 transients) Purity: 7.4% / Efficiency: 87.8 %



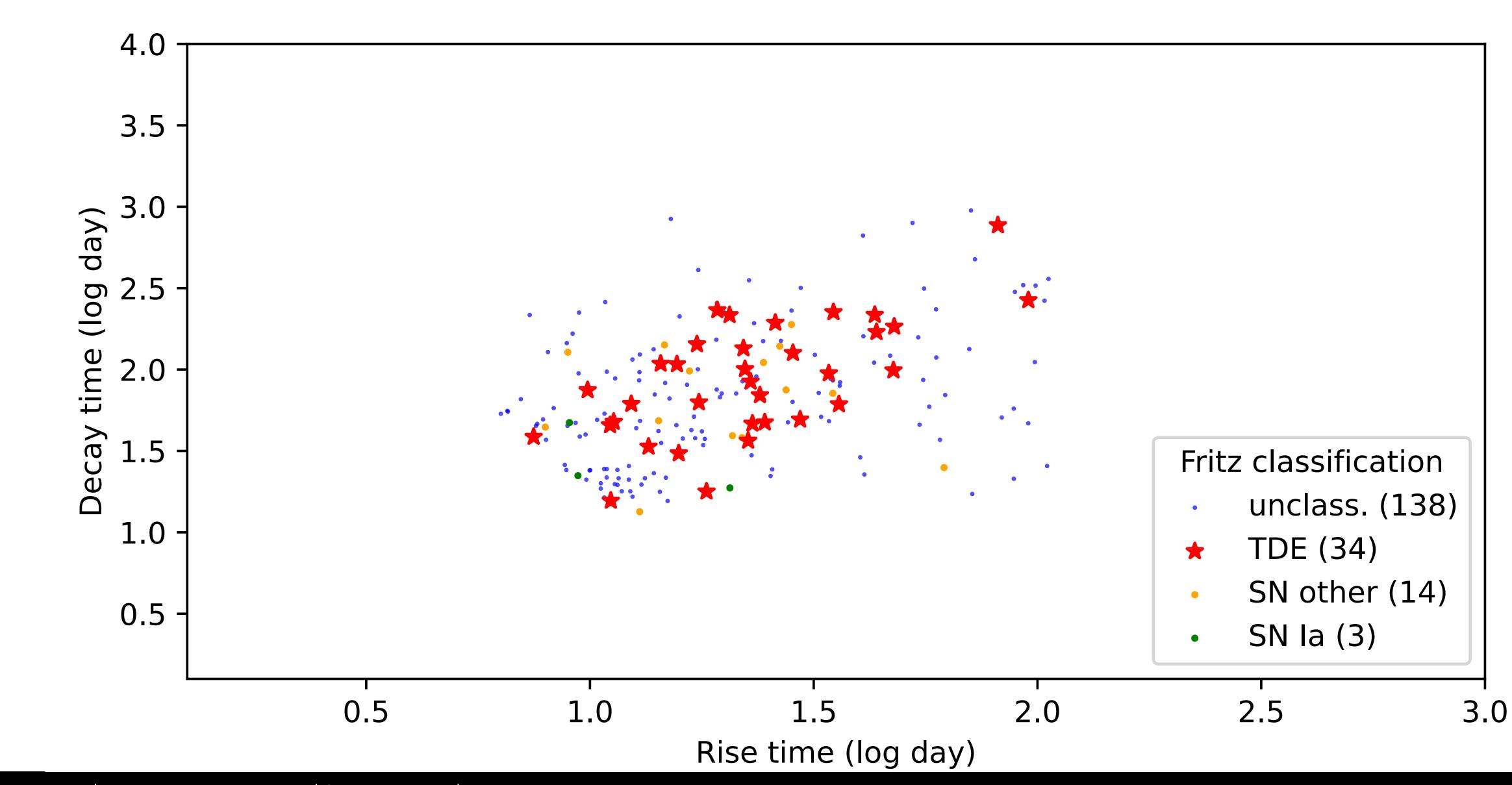
#### Rise-decay cut (424 transients) Purity: 9.7% / Efficiency: 83.7 %



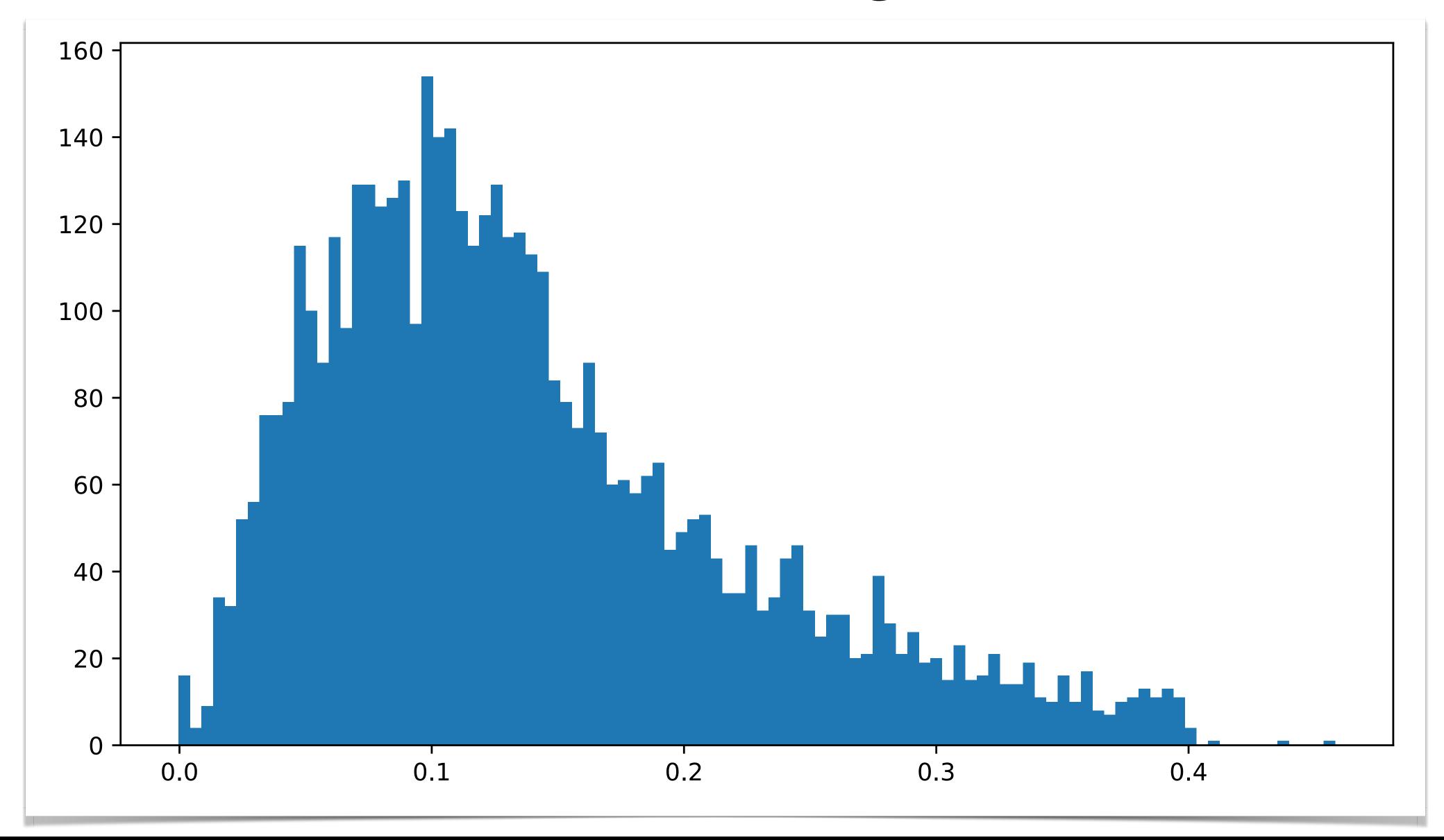
#### Chisquare cut (259 transients) Purity: 15.1% / Efficiency: 79.6 %



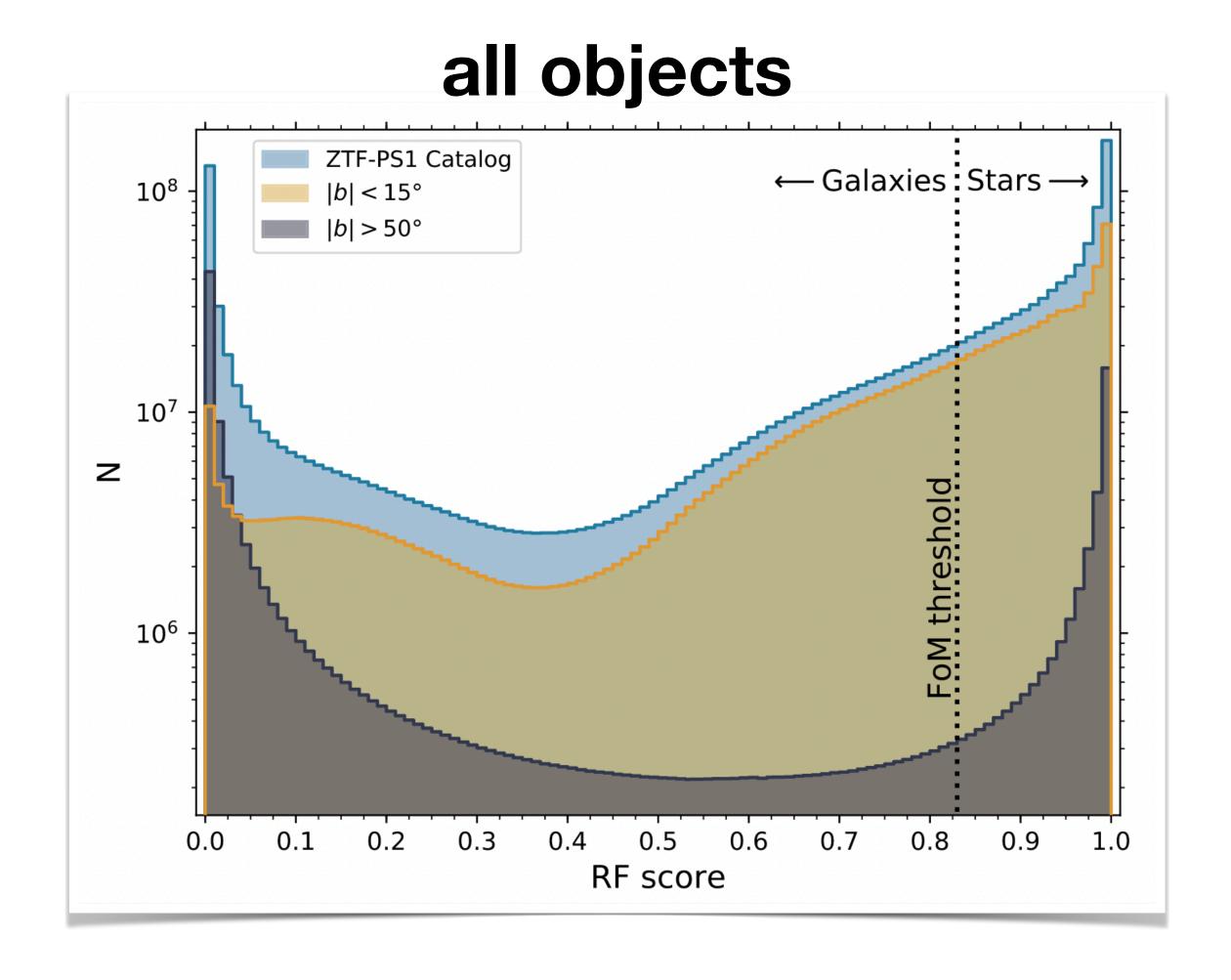
#### Bayesian block cut (189 transients) Purity: 18.0% / Efficiency: 69.4 %

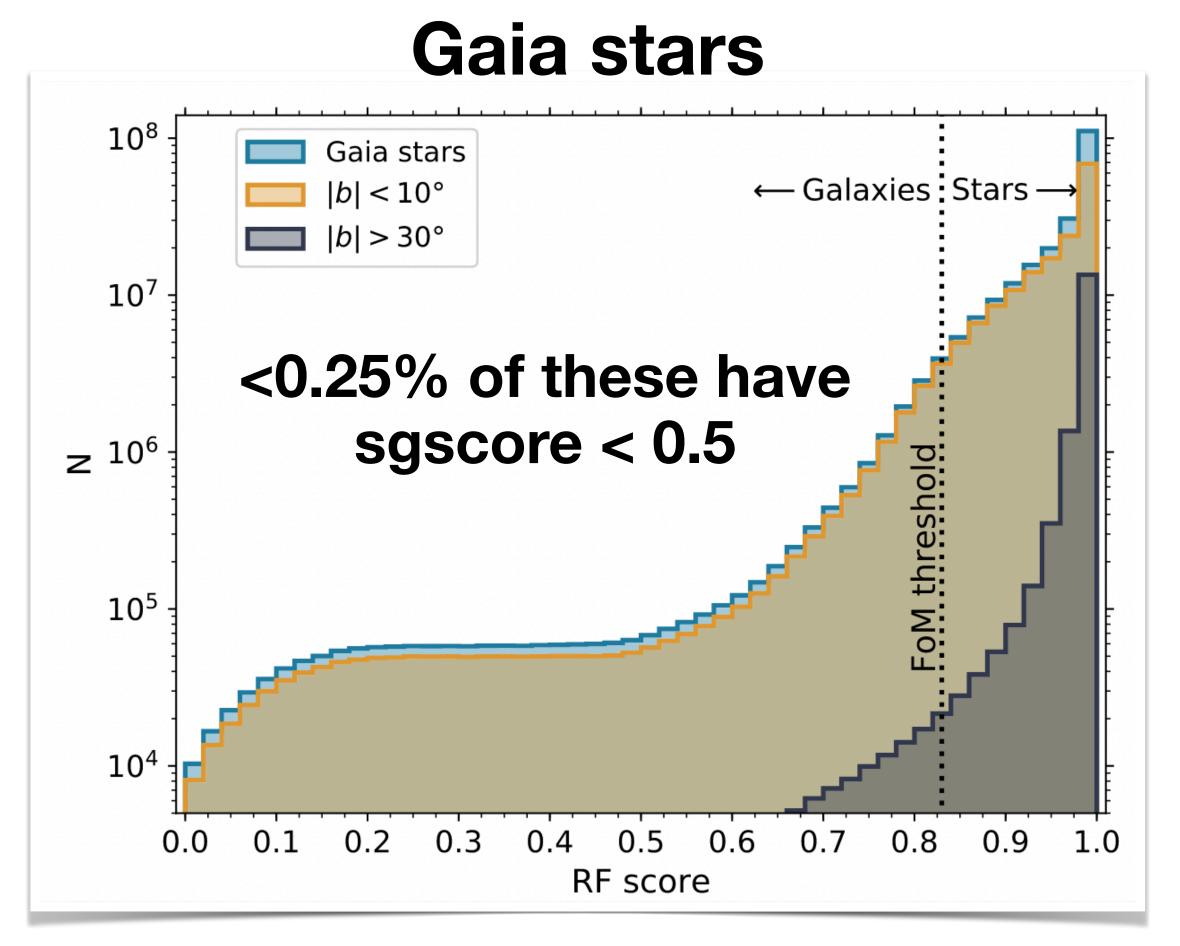


#### Redshift distribution (including photoz, n=4883)



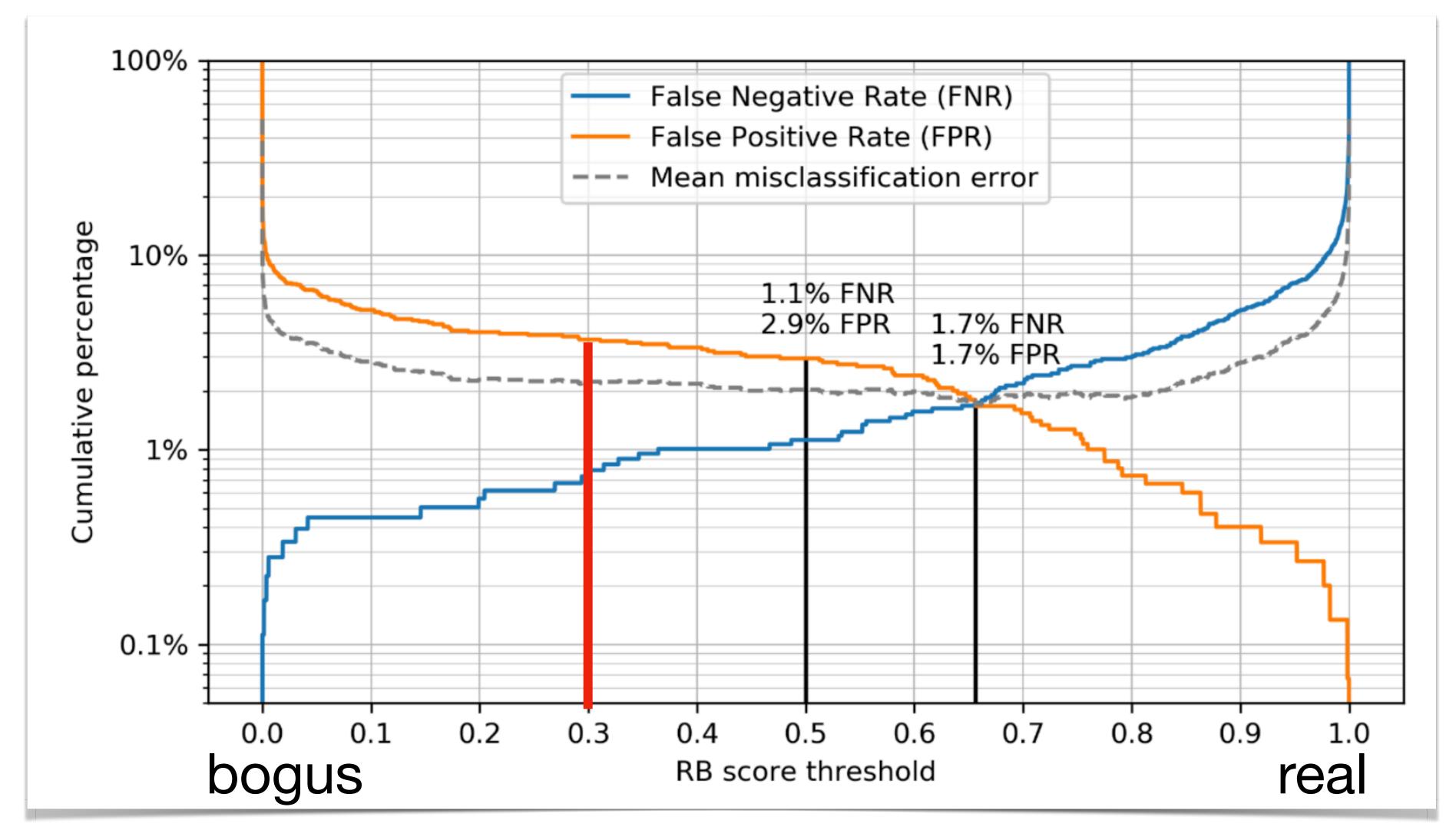
#### sgscore distribution





Tachibana & Miller, 2018 (PASP)

#### rb distribution



0.7% FNR4 % FPR

Duev et al., 2019 (MNRAS)