

*Finding Early SNIa with the Fink broker An extraordinary journey into the transient sky 4 April 2025* 





**Emille Ishida,** Anais Moller and Julien Peloton on behalf of the Fink Team From detection to science

### The data path

every ~30 seconds down to mag ~24

10 million alerts per night...

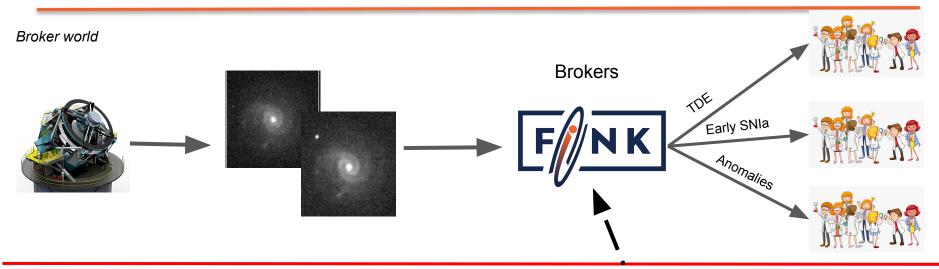
Machine learning Catalog association Streams join

BROKER

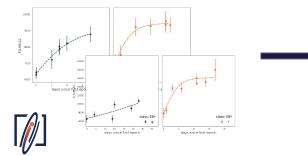
We would like the interesting ones ...



# How to classify alerts?



Domain specialist world (this is you)



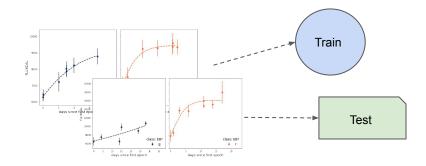
**Taylored science module** 

f(alerts; ++) => class scores Boolean

### Case study: Early SN Ia classification

Problem 1: labels are expensive, resources are limited

**Problem 2:** training (spectroscopically classified light curves) is <u>not</u> <u>representative</u> from test (purely photometric light curves)



**Goal:** optimize classification results with small training

**Strategy 1:** transfer learning Learn from simulations, apply in real data

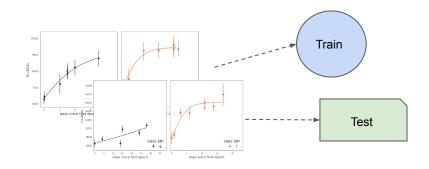




### Case study: Early SN Ia classification

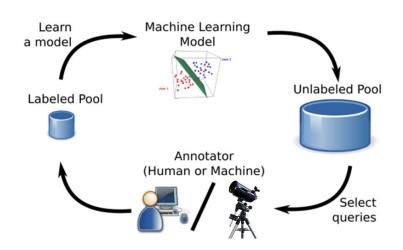
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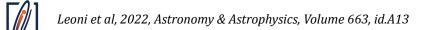
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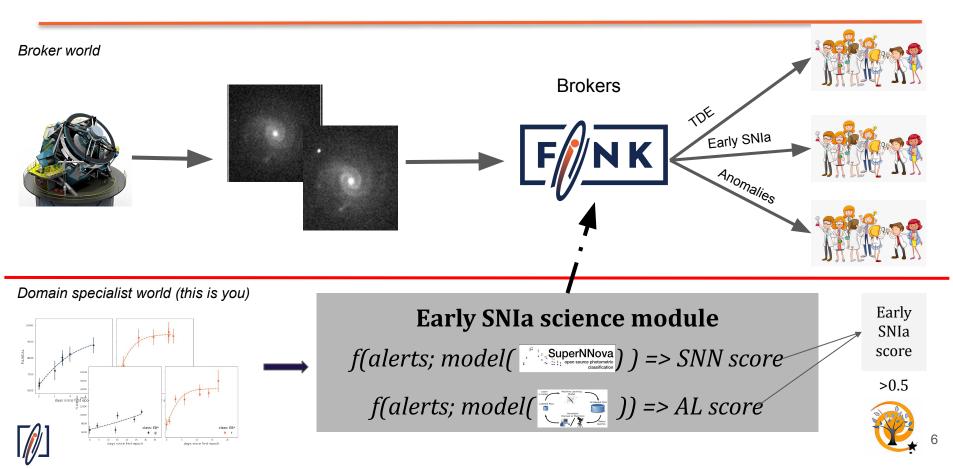
**Goal:** optimize classification results with small training

#### Strategy 2: optimize the construction of training samples



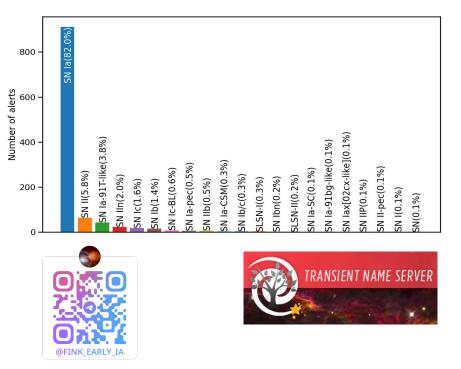


### Case study: Early SN la classification



### Case study: Early SN Ia classification

- RF trained in 310 alerts chosen by AL
- Only rising candidates
- Agreement with <u>SuperNNova</u>
- Surviving candidates sent to TNS
- from Nov/2020 to Oct/20224:
  - 2633 candidates sent
  - 1182 classified (45%)
    - 966 confirmed SNIa (82%)

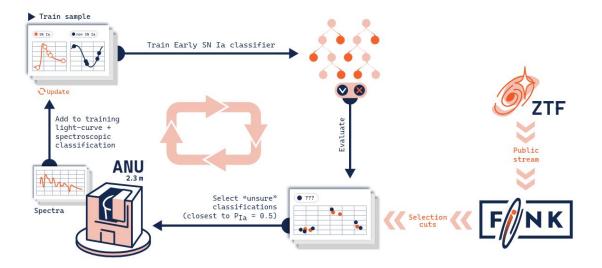




### **On-the-fly AL: Early SN la**

#### Real-Time Active Learning for optimised spectroscopic follow-up: Enhancing early SN Ia classification with the Fink broker

A. Möller,<sup>1,2</sup> E. E. O. Ishida,<sup>3</sup> J. Peloton,<sup>4</sup> O. Vidal Velázquez,<sup>1,2</sup> J. Soon,<sup>5</sup> B. Martin,<sup>5</sup> M. Cluver,<sup>1</sup> M. Leoni,<sup>4</sup> and E. Taylor<sup>1</sup>

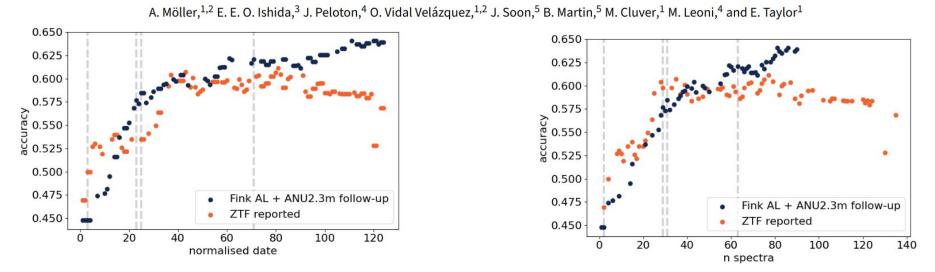




Moller et al, 2025, accepted to PASA, <u>arXiv:astro-ph:2502.19555</u>

### **On-the-fly AL: Early SN la**

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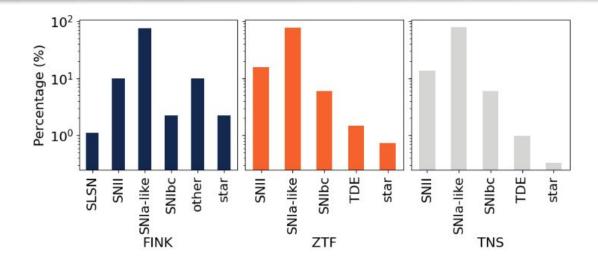


ANU 2.3m IFU for spectroscopic follow-up + extra spectra by DEBASS and ePESSTO+



Moller et al, 2025, accepted to PASA, <u>arXiv:astro-ph:2502.19555</u>

### **Target follow-up**



**Figure 4.** Spectroscopic classes for follow-up candidates in the FINK AL loop. We show from left to right panels FINK, ZTF and all TNS spectroscopic classifications. The percentage of SN families is similar to all strategies except for SLSN and other non-SN types of transients characterised by FINK.



# **For Rubin**

Robotic network @ Siding Spring Observatory

Australia





# Take home message

- Pool-based AL is only 1 possible strategy, there are others
- The output of all science modules are publicly available
- For Rubin, ideally the entire loop should be automatized
- This strategy can be applied for other well-defined classes
- It can be adapted for less well-known classes too ...

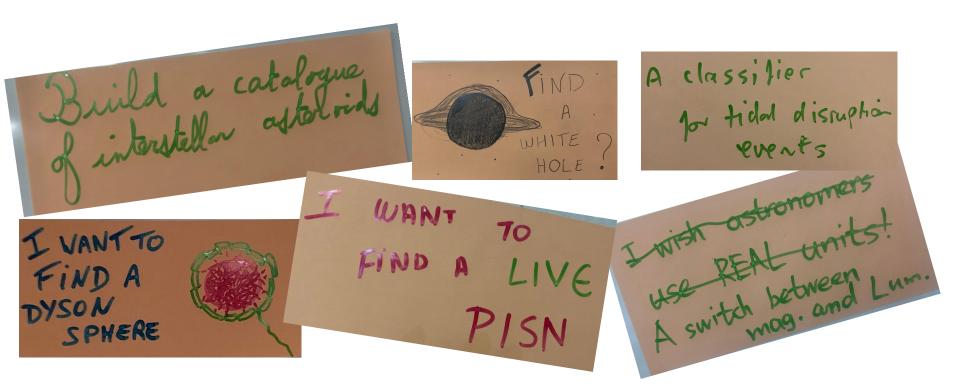


#### Personalized ML $\Rightarrow$ conversations



Get inspired

#### #FinkDreamShots



From OzFink 2023 - Melbourne, Australia - <u>https://www.ozgrav.org/ozfink-workshop-2023.html</u>

# What do you want to see?



### How to start a conversation?

- 1. Make a wish list of the type of light curve you would like to see
- 2. Contact the Fink team to discuss the feasibility of your proposal
- 3. Choose a data delivery method: kafka stream, bots, science portal, etc.
- 4. Inspect candidates, and return to step 2 until you are happy with the proposed candidates
- 5. Receive the data and perform your analysis!



#### Moller, Ishida et al., 2025

**Table 3.** FINK targets in the AL loop with types acquired with the ANU 2.3m spectra. Featureless and other indicate spectra which have no features consistent with a SN.

type	number
П	8
IIb	1
la	62
Ia-91T-like	2
la-pec	2
Iax[02cx-like]	1
Ibn	1
Ic-BL	1
SLSN	1
featureless	2
microlensing	2
other	5
star	2



*Mimic the spectroscopic classification procedure* 

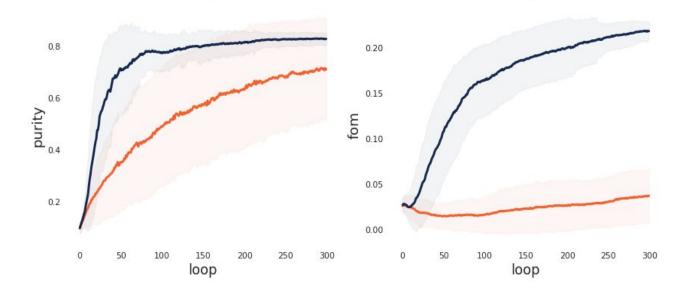
#### Model trained on ZTF alerts



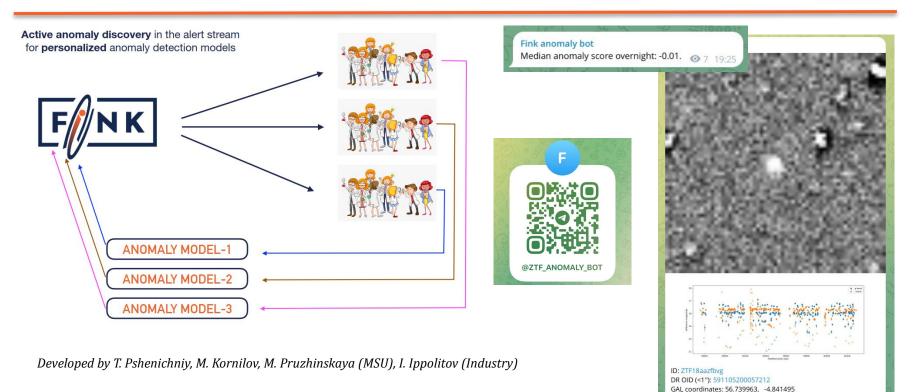
----- RandomSampling ------ UncSampling

#### Results after 300 loops:

*Training: 310 alerts Testing: > 52 000 alerts* 



### **Fink Anomaly Detection module**



Algorithm from Das, S., et al., 2017, in DEA'17, KDD workshop, <u>arXiv:cs.LG/1708.09441</u> Implementation by the <u>SNAD team</u>, via <u>coniferest package</u>

20

**○**7 16:45

EQU: 298.6919835, 18.6325727

UTC: 2023-12-07 03:12:54.999

Real bogus: 0.68 Anomaly score: -0.04

# **Fink services**

- Live streams (Kafka streams): fink-client
  - Whatfor? Live inspection, Follow-up
  - Personalisable filters to select objects/parameters of interest
  - Data received "live" (+processing delays)
- Science Portal (dash-based) & REST API: <u>https://fink-portal.org</u>
  - Whatfor? Visual inspection, small queries, daily monitoring
  - All data processed remains accessible
- Data Transfer service: <u>fink-client</u>, <u>post</u>, <u>link</u>
  - Whatfor? Bulk download, complex queries, ML/DL training, exotic analyses
- TOM module
  - Whatfor? Follow-up
  - <u>https://github.com/TOMToolkit/tom\_fink</u>



### Case study: Hostless transients



#### ELEPHANT: ExtragaLactic alErt Pipeline for Hostless AstroNomical Transients

