



# **NOMAI: A real-time photometric classifier for superluminous supernovae identification**

*Fink general meeting*

**Etienne Russeil**

# Superluminous supernovae (SLSNe)



- Core collapse supernovae with  $M < -20$
- Wide variety of behaviour
- < 500 known SLSNe
- LSST:  $10^4$  SLSN per year (Villar+2018)

Mostly ZTF



Credit: Palomar Observatory/Caltech

# Searching for SLSNe in the ZTF stream

## Already multiple classifiers



# Searching for SLSNe in the ZTF stream

Already multiple classifiers

Now Fink also has one!



## NOMAI: A real-time photometric classifier for superluminous supernovae identification

A science module for the Fink broker

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### ABSTRACT

Superluminous supernovae (SLSNe) are one of the most luminous stellar explosions known, yet they remain poorly understood. Because they are intrinsically rare, efficiently identifying them in the large alert streams produced by modern time-domain surveys is essential for building larger observational samples and enabling spectroscopic follow-up. We present NOMAI, a machine learning classifier designed to identify SLSN candidates directly from photometric alerts in the ZTF stream, using light curves accumulated over at least 30 days. It does not require any spectroscopic redshift and is running in real time within the Fink broker. ZTF light curves are transformed into a set of physically motivated features derived primarily from model-fitting procedures using SALT2 and Rainbow, a blackbody-based multi-band fitting framework. These features are used to train an XGBoost classifier on a curated dataset of labeled ZTF sources constructed using literature samples of SLSNe, along with TNS and internal ZTF labeled sources. The final training dataset contains 5280 unique sources, including 225 spectroscopically classified SLSNe. On the training sample, the classifier reaches 66% completeness and 58% purity. Deployed within the Fink broker, NOMAI has been running continuously since 18/12/2025 on the ZTF alert stream and publicly reports SLSN candidates every night by automatically posting them to dedicated communication channels. Based on this, we also report the first two-month as an evaluation period, where the classifier successfully recovered 22 of the 24 active SLSNe reported on the Transient Name Server. The achieved performances, particularly the high completeness in real-time operations, demonstrate that the classifier provides a valuable tool for experts to efficiently scan the alert stream and identify promising candidates. In the near future, NOMAI is intended to be adapted to operate on the Legacy Survey of Space and Time conducted by the Vera C. Rubin Observatory, which is expected to uncover an unprecedented number of transients, making machine learning based photometric classification essential.

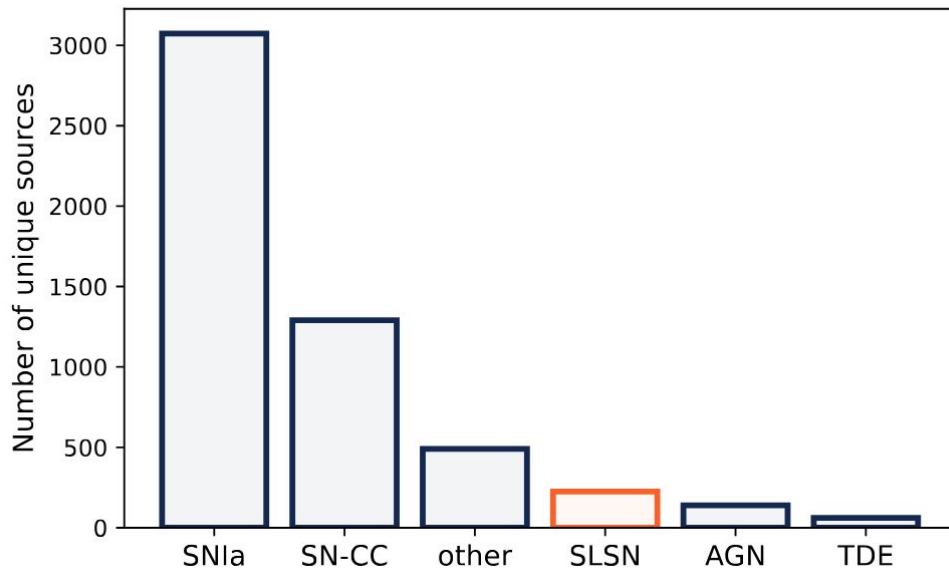
**Key words:** supernovae: general, methods: data analysis, techniques: photometric, surveys, astronomical databases: miscellaneous, methods: numerical

# The training sample (spectroscopic labels)

Private ZTF labels



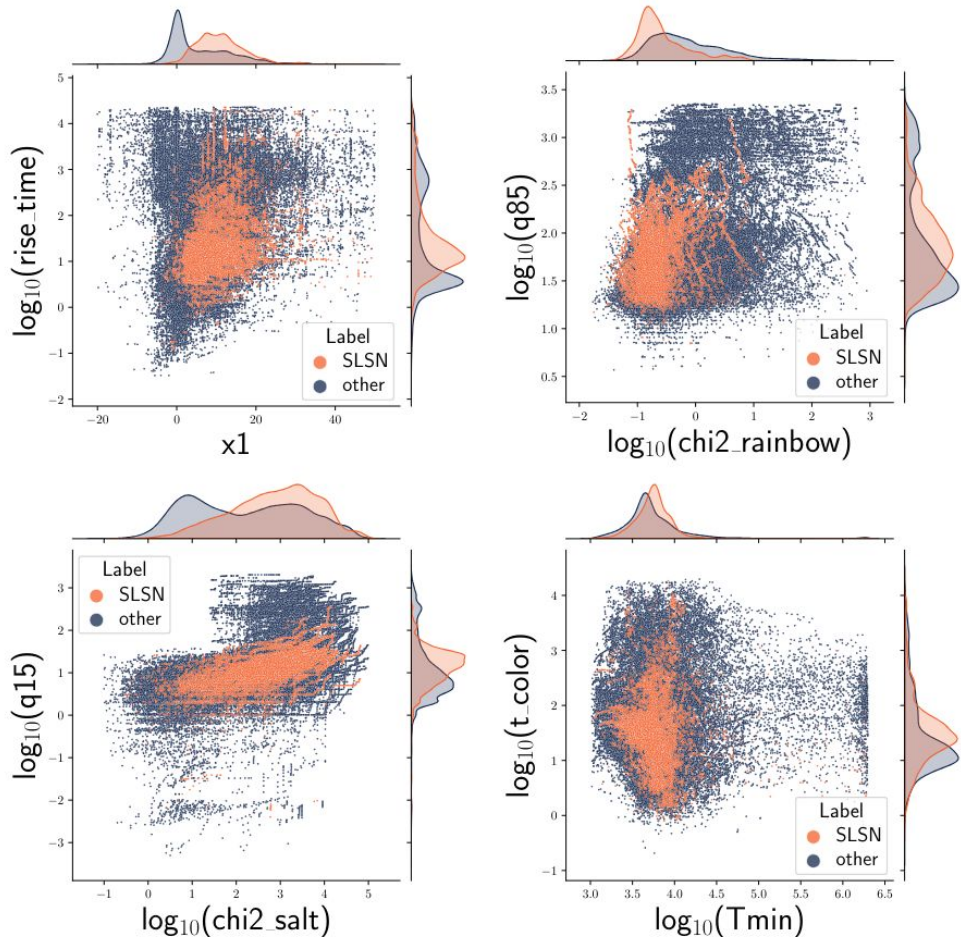
- **SLSNe from sample papers**  
*Gomez et al. (2024) & Pessi, P. J. et al. (2025)*
- **Every available TNS labels**
- **Reclassify all missed SLSNe**



# Feature extraction

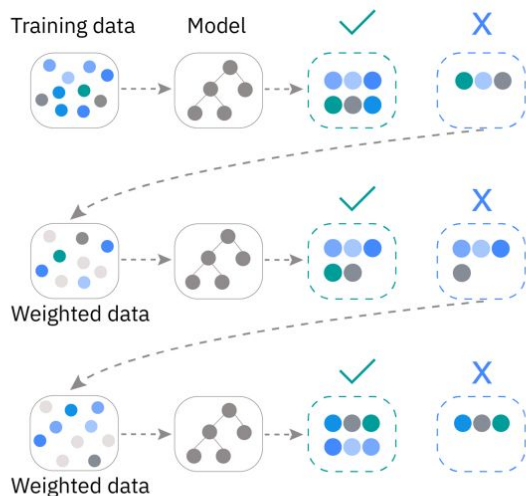
## 27 features per light curve

- SALT2 features
- Rainbow features *Russeil et al. 2024*
- Statistical features
- Distnr
- **No redshift !**



# Machine learning method

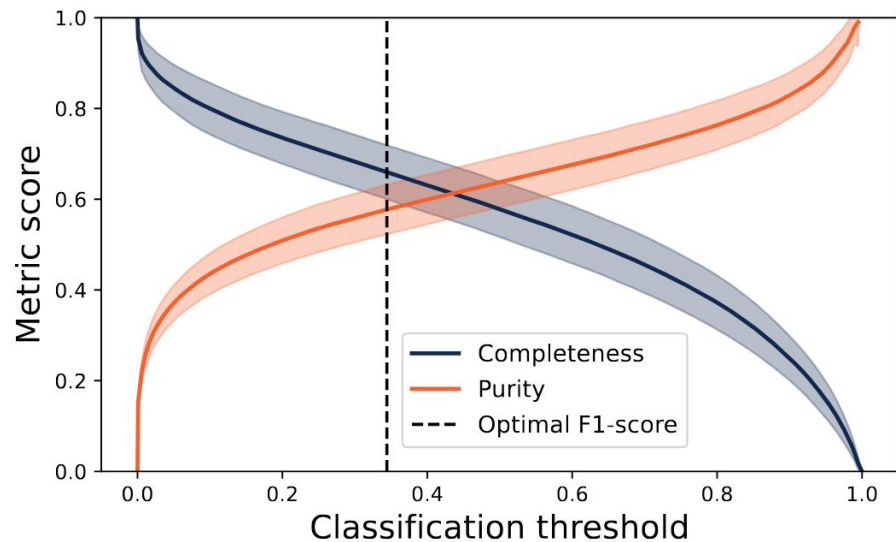
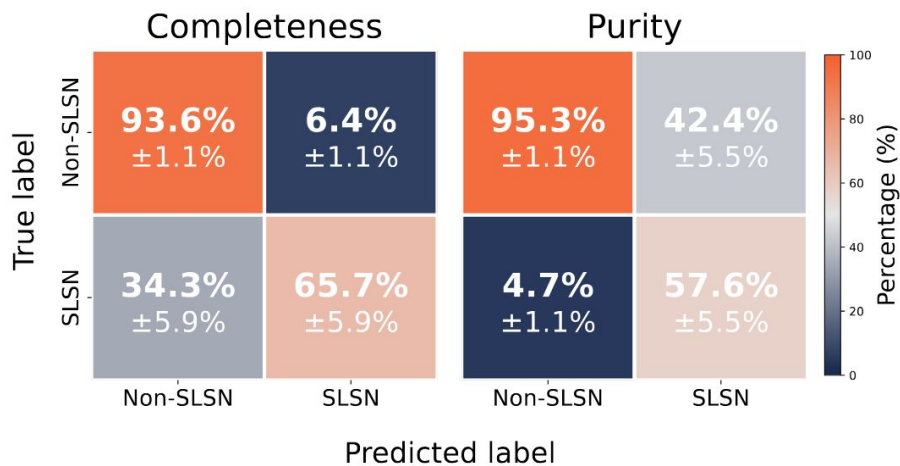
## XGBoost classifier



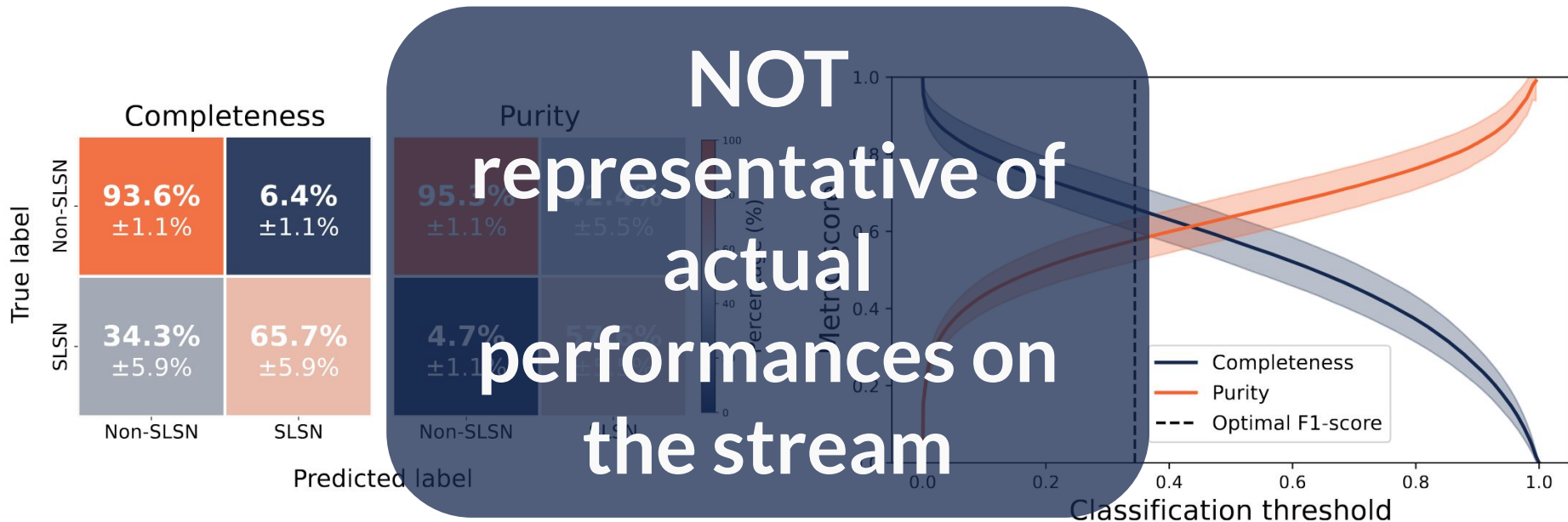
## Hyperparameter optimization

*[max\_depth]*  
*[learning\_rate]*  
*[max\_delta\_step]*  
*[reg\_lambda]*  
*[reg\_alpha]*  
**1000 random sets**

# Training results



# Training results





Has been running live  
for several months now

## SLSN SLACK BOT



**slns-bot** APPLI 21 h 20

Mardi 3 mars ▾

Number of candidates for the night 20260303: 4 (4 unique objects).

	objectId	slns_score
0	ZTF25abxjtkv	0.961665
1	ZTF25abuovxv	0.768442
2	ZTF24ablzlmw	0.642631
3	ZTF26aabpbkj	0.434377

21 h 21 =====

TNS classification: SN IIn

Fink: <https://ztf.fink-portal.org/ZTF25abxjtkv>

Fritz: ZTF25abxjtkv

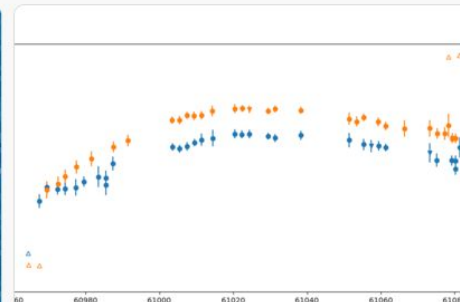
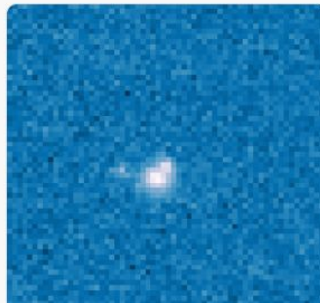
Score: 0.962

E(B-V) = 0.048

SDSS photo-z = 0.084 +- 0.017

Peak M = -19.93 (-20.36 < M < -19.41)

2 fichiers ▾ |  Tout télécharger



# What results after 2 months ?

*From 18/12/25 to 18/02/25*

**Purity**

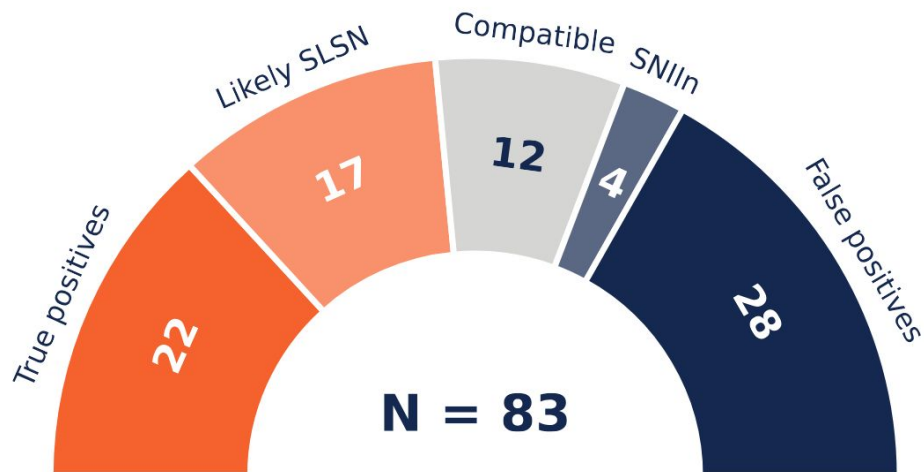
**Completeness**

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Purity

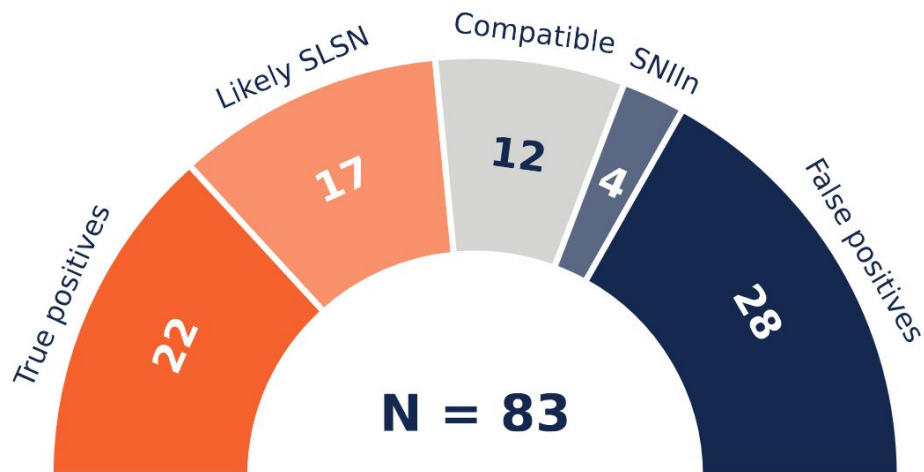
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# What results after 2 months ?

From 18/12/25 to 18/02/25

## Purity



## Completeness

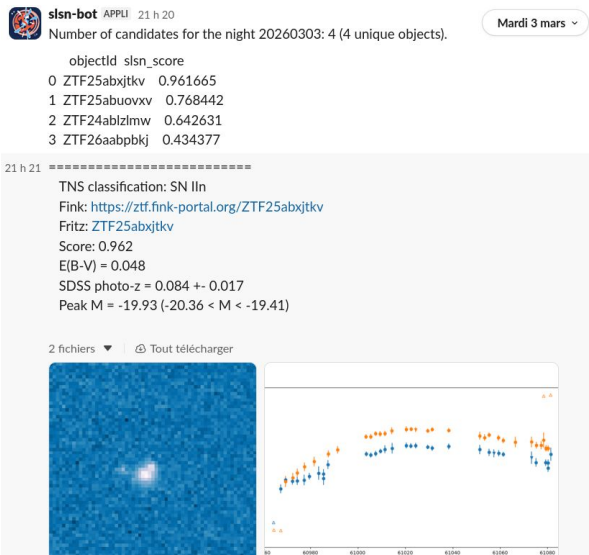
22 / 24 = 92 %

**Not bad!**

(But of course impossible to know the true value)

# How to access NOMAI's predicitions ?

## Slack bot (slsn-bot)

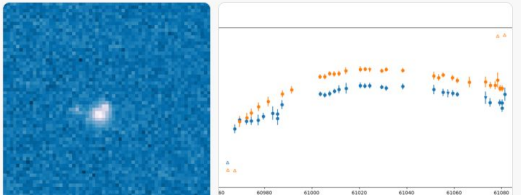


**slsn-bot** APPLI 21 h 20  
Number of candidates for the night 20260303: 4 (4 unique objects). Mardi 3 mars

objectid	slsn_score
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21 h 21  
-----  
TNS classification: SN IIn  
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Score: 0.962  
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Peak M = -19.93 (-20.36 < M < -19.41)

2 fichiers Tout télécharger




## Fink data transfer

With a query using  
the alert value:

`slsn_score > 0.35`

## Fink class



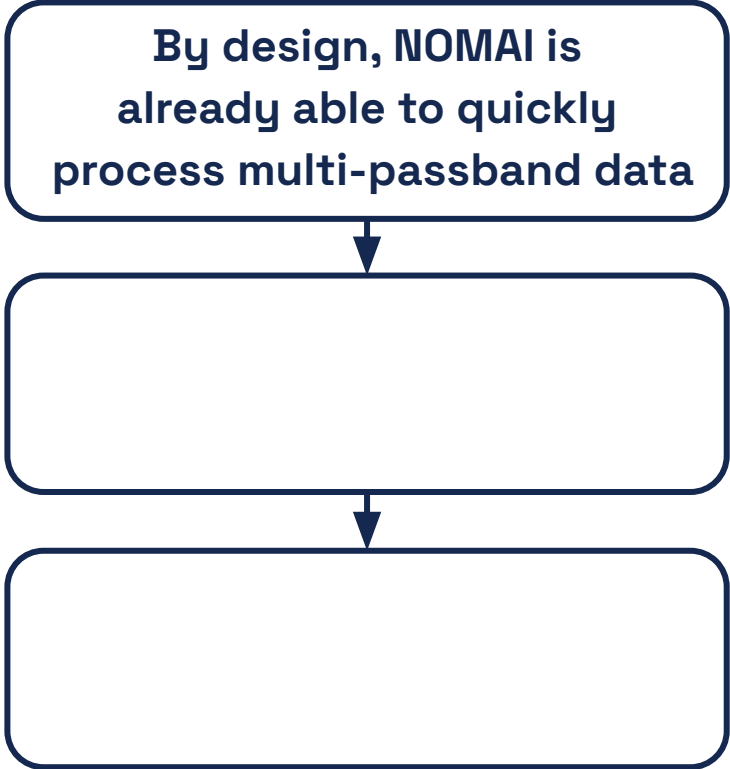
Quick fields: [class](#) [trend](#) [last](#) [radius](#) [after](#) [before](#) [window](#)

**CLASS**  
Class based search / SLSN candidate  
class=SLSN candidate

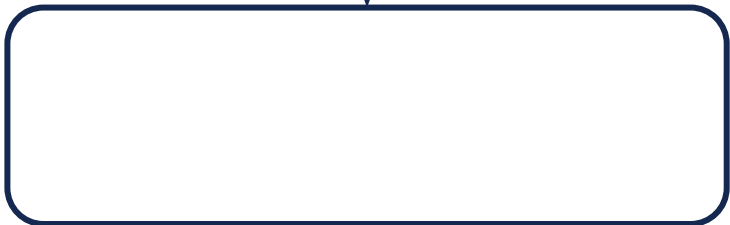
Check out the new 'SLSN candidate' class from NOMAI <https://arxiv.org/abs/2604.13761>

# NOMAI for LSST ?

By design, NOMAI is  
already able to quickly  
process multi-passband data



```
graph TD; A[By design, NOMAI is already able to quickly process multi-passband data] --> B[ ]; B --> C[ ]
```



# NOMAI for LSST ?

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A first crude model will  
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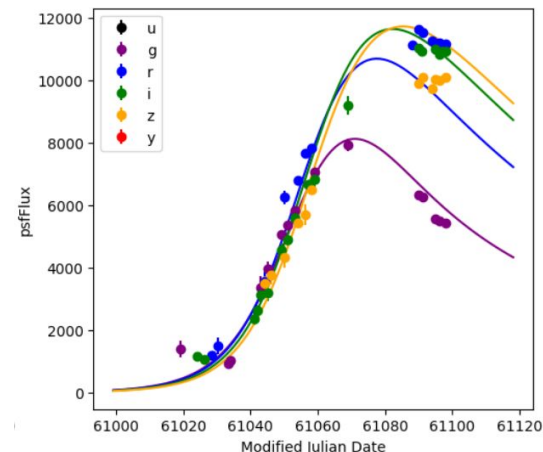
As LSST will discover more  
SLSN, we will improve  
the model with real data

# NOMAI for LSST ?

By design, NOMAI is already able to quickly process **multi-passband data**

A first crude model will be trained on ELAsTiCC

As LSST will discover more SLSN, we will improve the model with real data



Blackbody  
model

# BUT ...

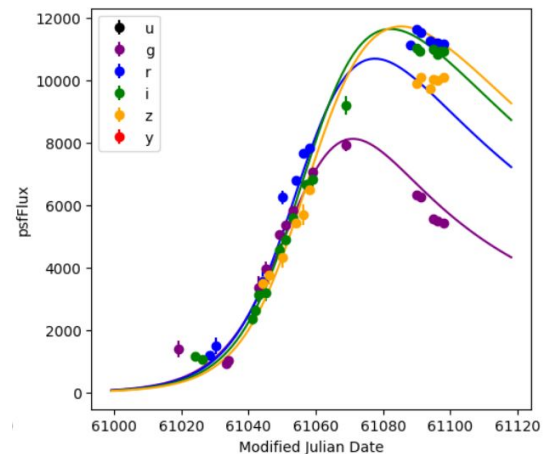
Rainbow might not  
be enough for LSST

# NOMAI for LSST ?

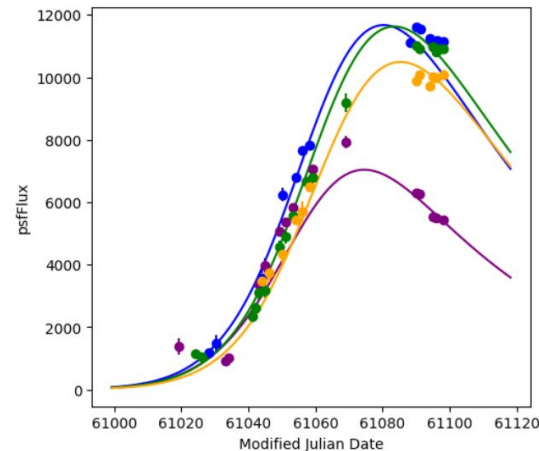
By design, NOMAI is already able to quickly process **multi-passband data**

A first crude model will be trained on ELAsTiCC

As LSST will discover more SLSN, we will improve the model with real data



**Blackbody  
model**



**UV extincted  
Blackbody  
model**

# NOMAI for LSST ?

By design, NOMAI is already able to **quickly** process multi-passband data



## YES BUT ...

Not fast enough for LSST.

An optimized version of Rainbow is being developed in Rust

A first crude model will be trained on ELAsTiCC

As LSST will discover more SLSN, we will improve the model with real data

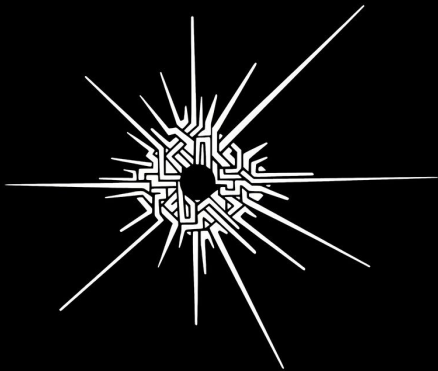
# Takeaway messages

**SLSNe were  
rare,  
but are  
becoming  
abundant**

**NOMAI is  
built to  
miss as few  
candidates  
as possible**

**Ready for  
today,  
evolving  
for  
tomorrow**





# OUTER WILDS

