

Superluminous supernovae classification in big data surveys

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Superluminous supernovae

Superluminous supernovae (SLSNe) are extreme supernova events with an absolute magnitude $M < -20$, making them **10 to 100 times more luminous** than standard supernovae. They are rare events for which we still lack a complete understanding. In particular, the subject of the mechanisms responsible for the excess of energy is an active question. Explanations include **circumstellar interaction** with a thick medium, or the formation of a **magnetar** acting as a central engine. It is also theoretically predicted that population III stars could lead to SLSN through an electron-positron **pair instability** mechanism. Discovering more SLSN is crucial to better understand these events.



Photometric telescopes

Hundreds of SLSN have been discovered by the **Zwicky Transient Facility (ZTF)**, making it our current best tool to find them. In order to study them live, ZTF triggers an alert everytime a source brightness varies. Currently, ZTF generates **~300k alerts/night**. Extracting information from this amount of data is a true data challenge. Soon, the **Vera Rubin Observatory (LSST)** will start its operation. Using the largest camera ever built, it will produce **~10 million alerts** per night. Among these, we expect to discover **10⁴ new SLSNe each year** (Villar+2018), which will drastically increasing our current sample.



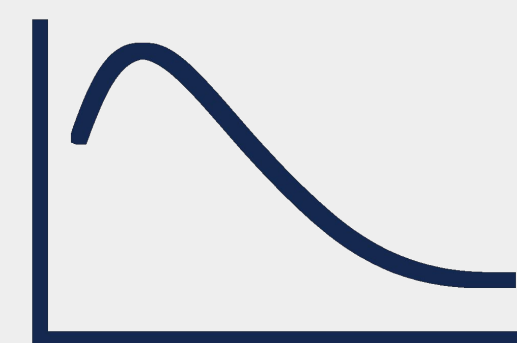
The Zwicky Transient Facility located in California has an 1.2 m mirror and a 47 square degrees field of view.



The Vera Rubin Observatory located in Chile has an 8.4 m mirror and a 9.6 square degrees field of view. It will observe with 6 distinct filters

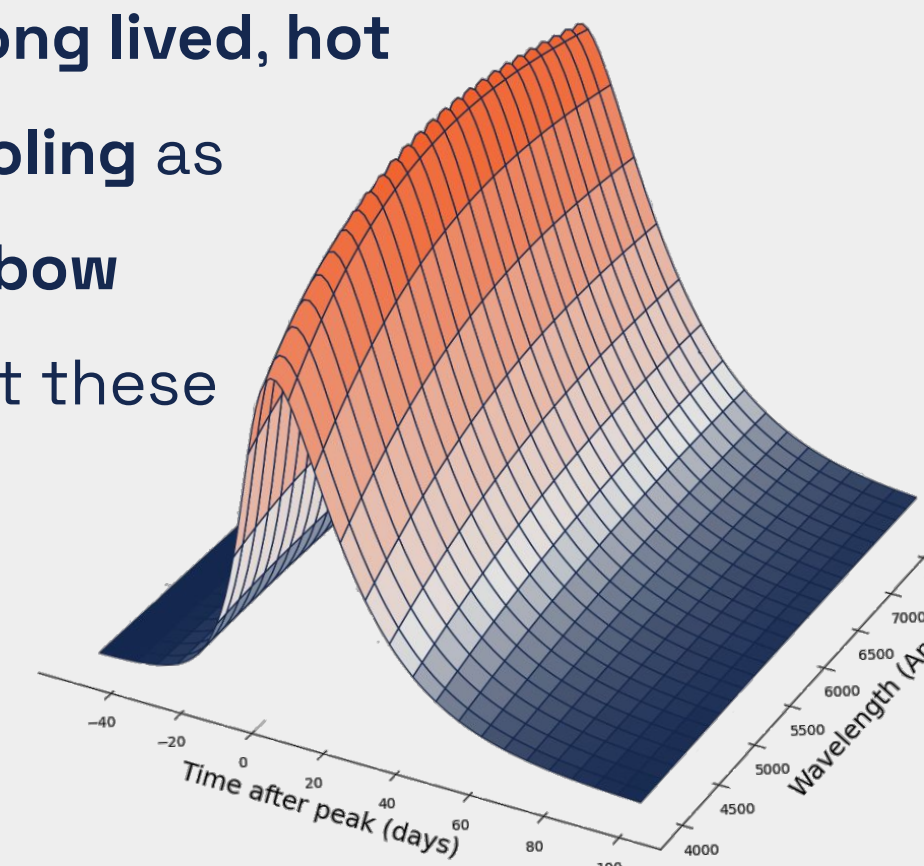
The Fink broker

Given the volume of alerts produced every night, both LSST and ZTF rely on **data brokers**. **Fink** is one of the seven brokers chosen to receive LSST data, and is currently processing ZTF alerts. It has the role of continuously **ingesting, filtering, augmenting** and **distributing** the stream. In this context, **science modules** can be build and attached to broker to produce real time photometric classification. The SLSNe classifier will soon constitute a **new science module**, whose predictions will be **publicly available**.



Light curves

Although spectroscopic observation is often required to confidently classify a transient source, the light curve contains information to distinguish various type of events, even **without redshift**. In our case, SLSNe light curves are mostly **long lived, hot** and exhibit a **thermal cooling** as they decay. We use **Rainbow** (Russeil+2024) to extract these properties. It is a fast **2D parametric fit** which assumes a **blackbody** source and models its **temperature evolution**.



Visualization of the 2D surface fitted by Rainbow on a supernovae.

The classifier

Using the parameters extracted from the **Rainbow fit**, along with a series of **statistical features**, we train a machine learning model to separate SLSN light curves from other transients. The data sample has been chosen for its **spectroscopic high quality labels**.

We train an **XGBoost** model and optimize the hyperparameters to **maximize completeness**. In this way we are able to find rare/unusual SLSN.

We manage to **identify 3/4 of all SLSN alerts!** And this high completeness is achieved while maintaining a **purity of 50%**.

True label \ Predicted label	Non-SLSN	SLSN
Non-SLSN	89.37% ± 0.57%	10.63% ± 0.29%
SLSN	24.39% ± 1.94%	75.61% ± 2.79%

Confusion matrix of the classifier applied on the testing sample. Normalized for completeness

Looking at the events wrongly classified as SLSN, we discover that a large proportion are very **luminous SN-IIn**, for which the question of the **continuum** with the **SLSN-II** class is still under debate.

These results are promising and should lead to an efficient method to discover interesting SLSN in the public stream of ZTF, and soon LSST.