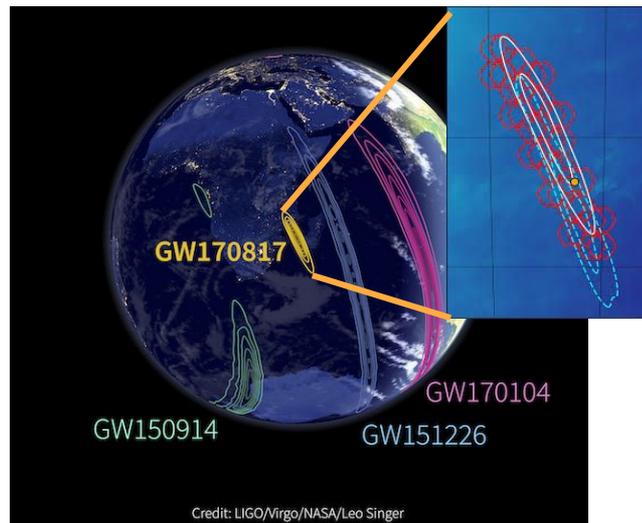
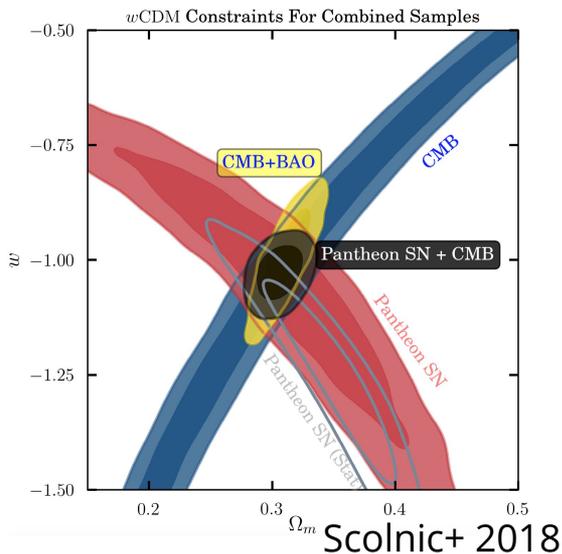
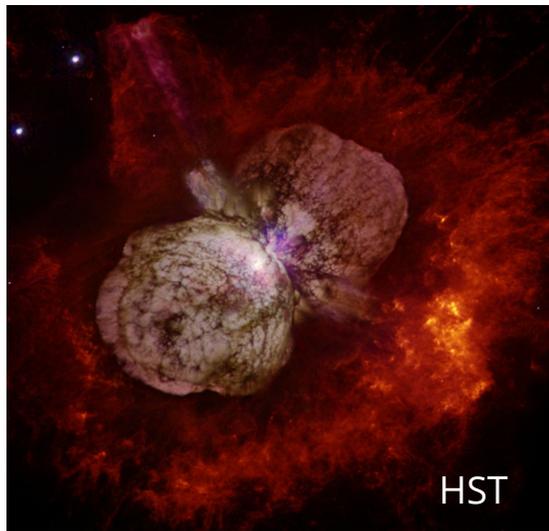

Time-domain Astrophysics in the Era of Big Data

— V. Ashley Villar —

Center for Astrophysics | Harvard & Smithsonian
Ford Foundation Dissertation Fellow

ML @ Ringberg 2019

Transients connect to all branches of astrophysics

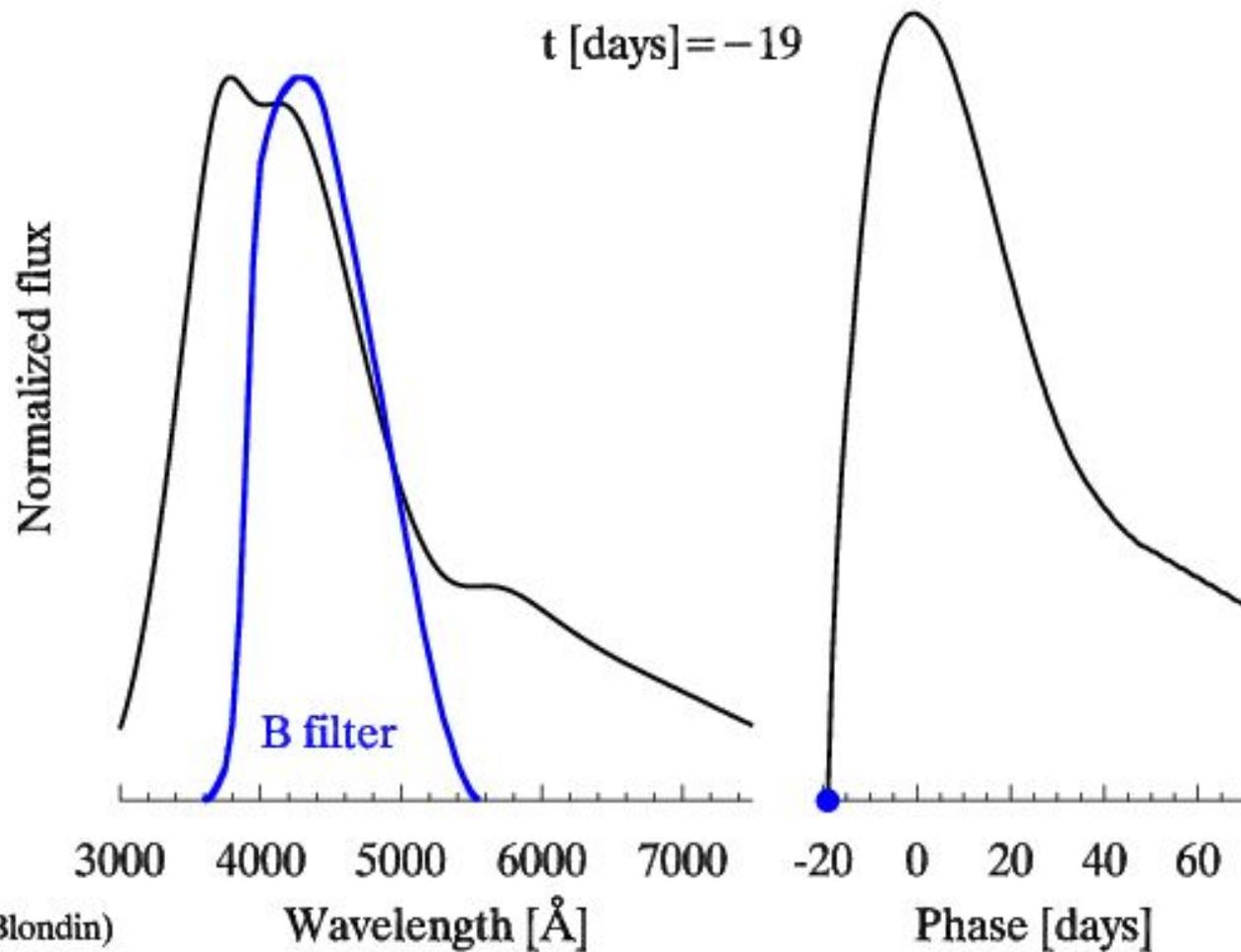


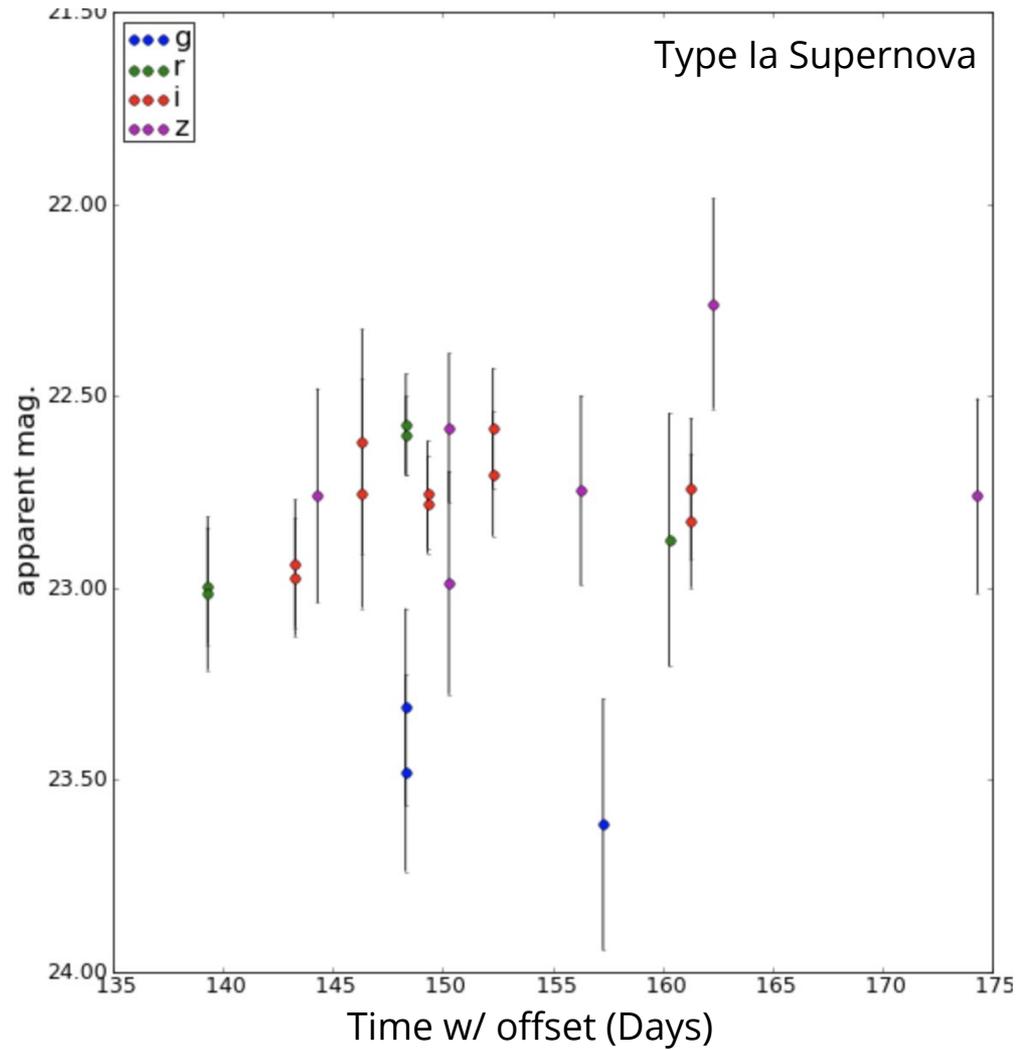
Soares-Santos+2017
L. Singer

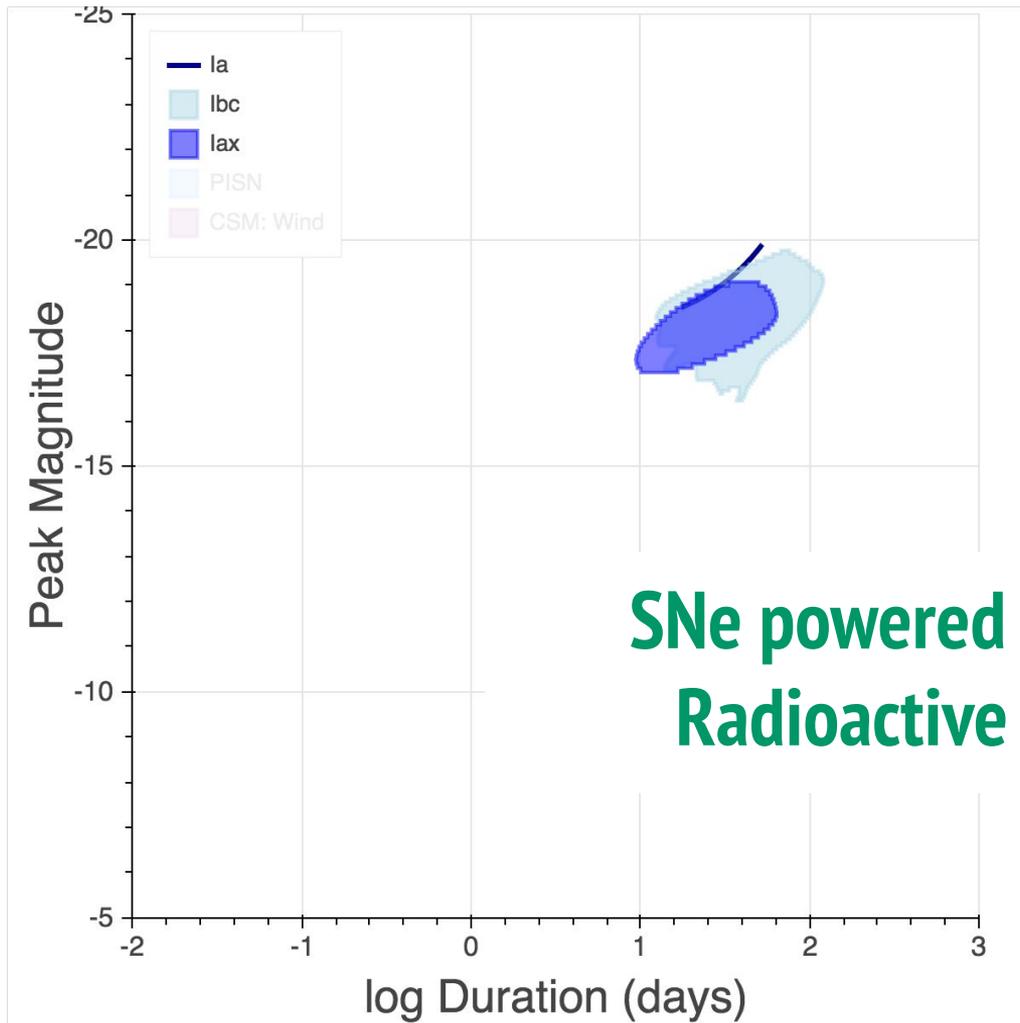
How does the zoo of observed transients connect with the underlying (astro)physics?

SN Ia spectrum

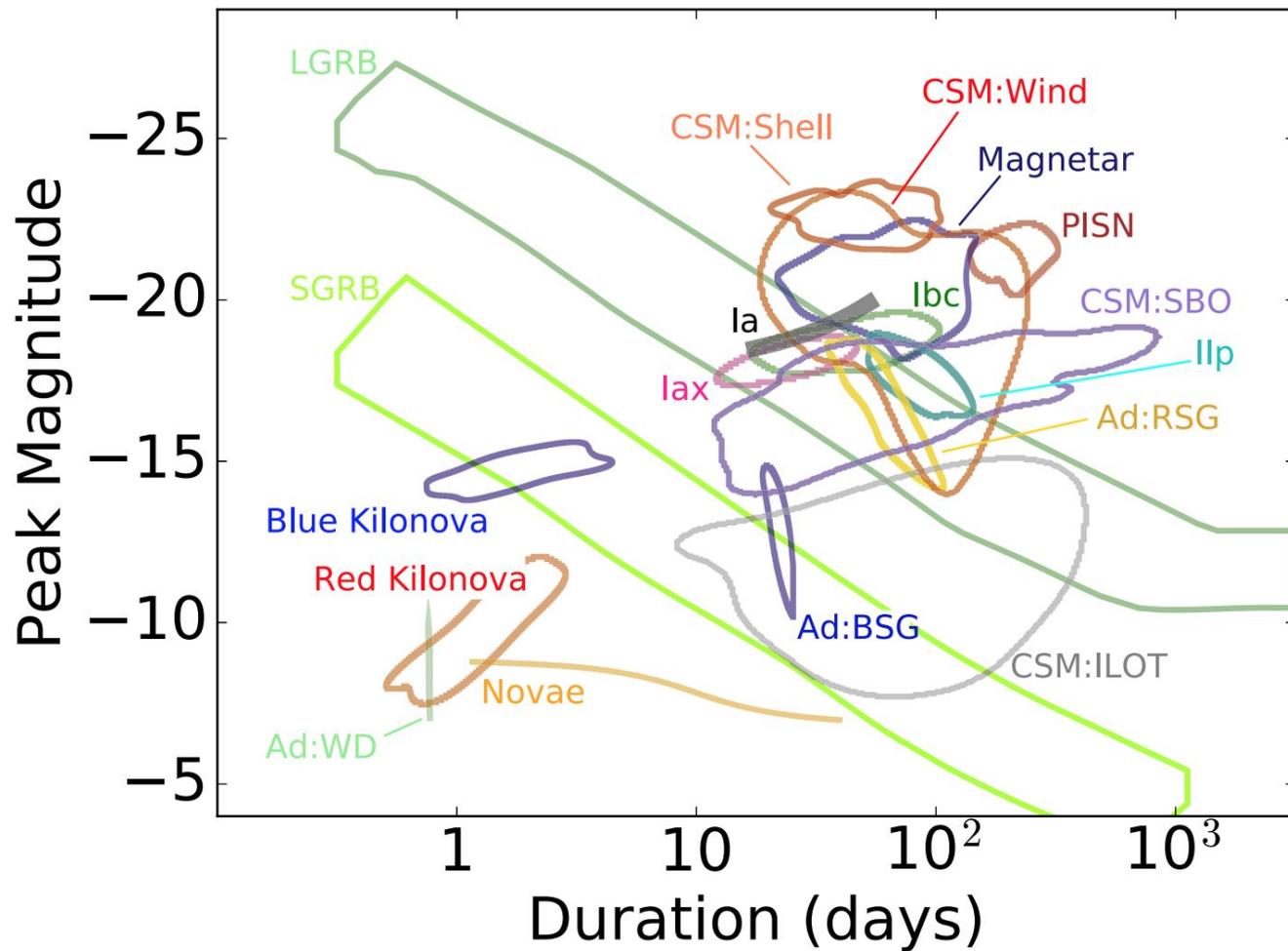
SN Ia light curve



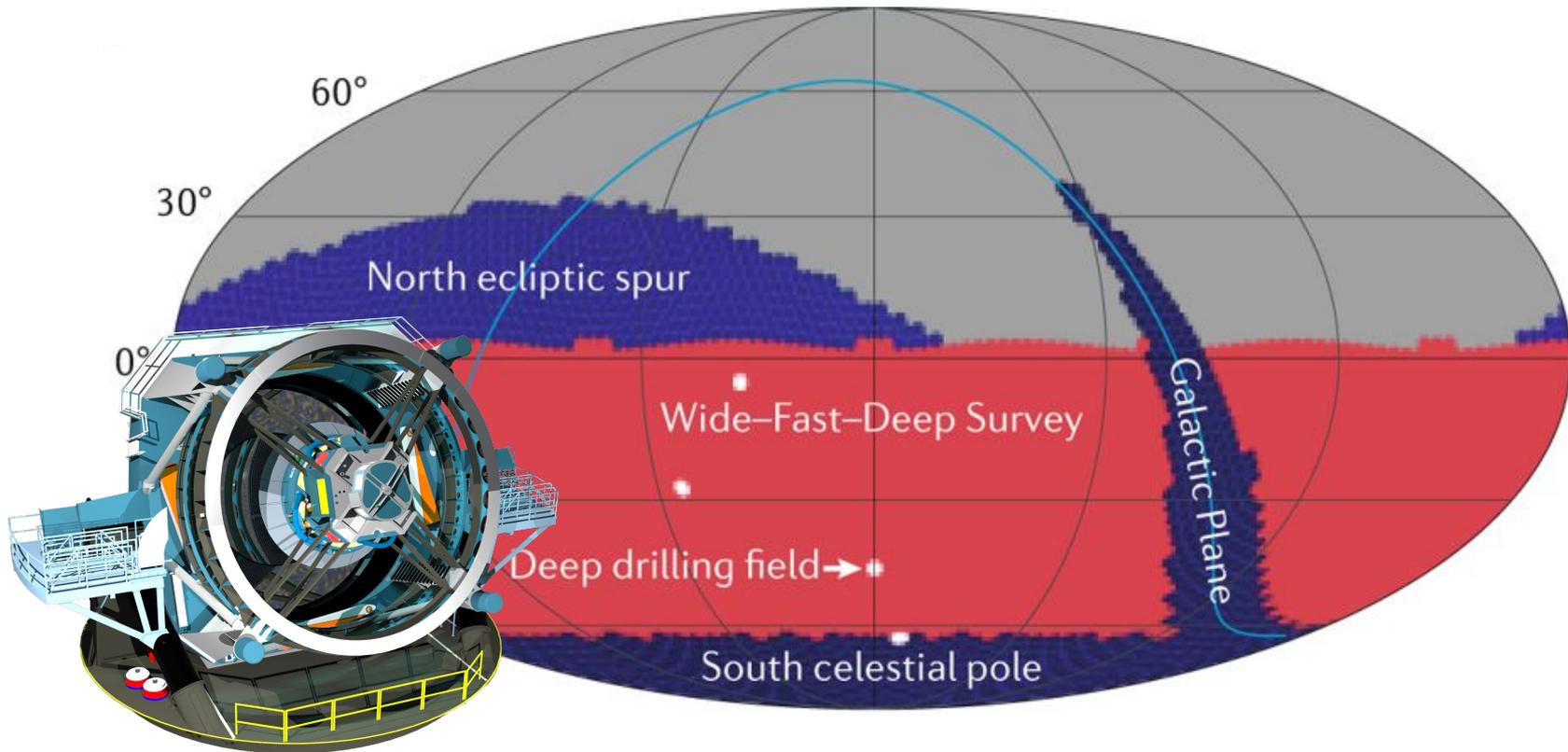




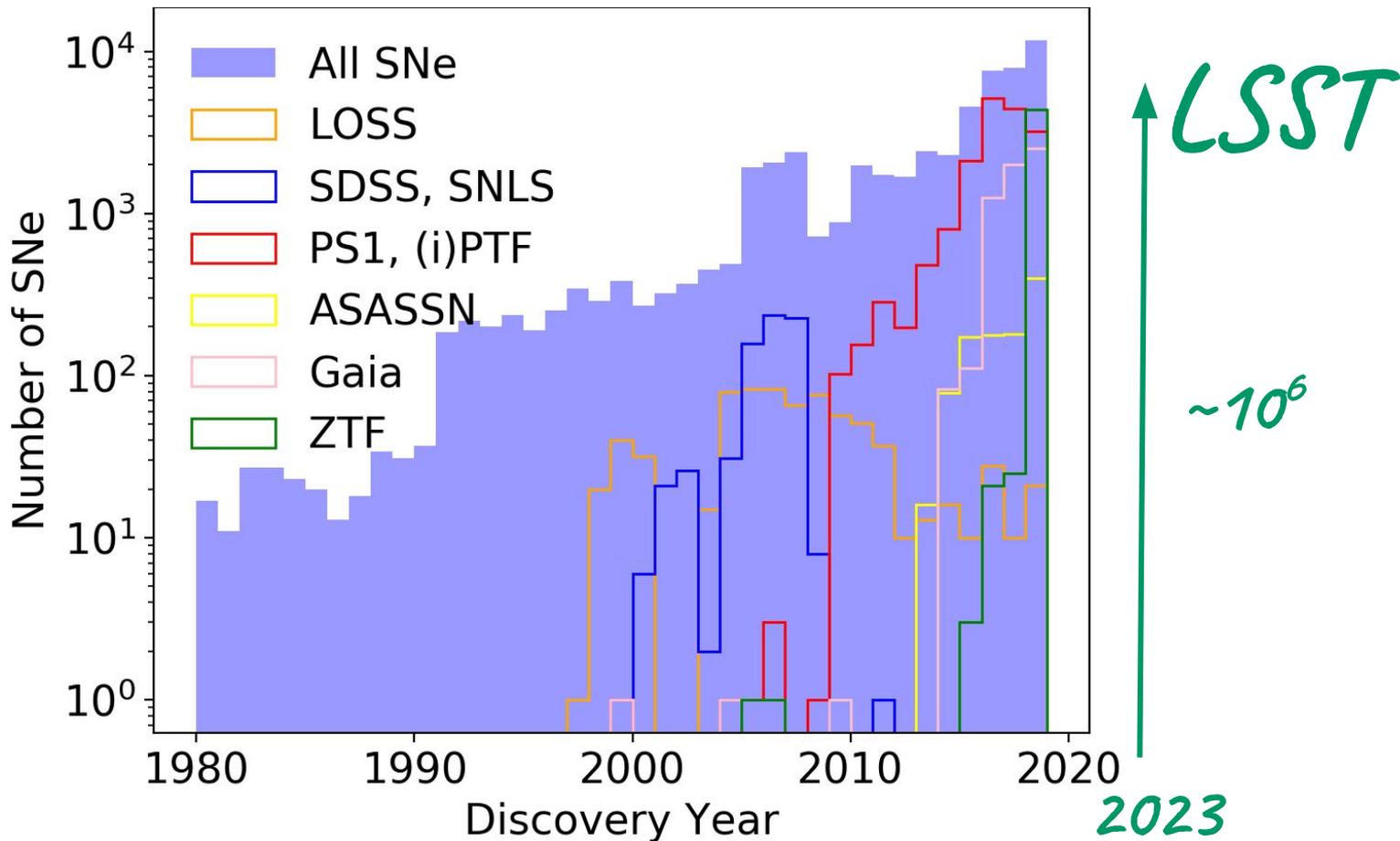
**SNe powered by ^{56}Ni
Radioactive Decay**



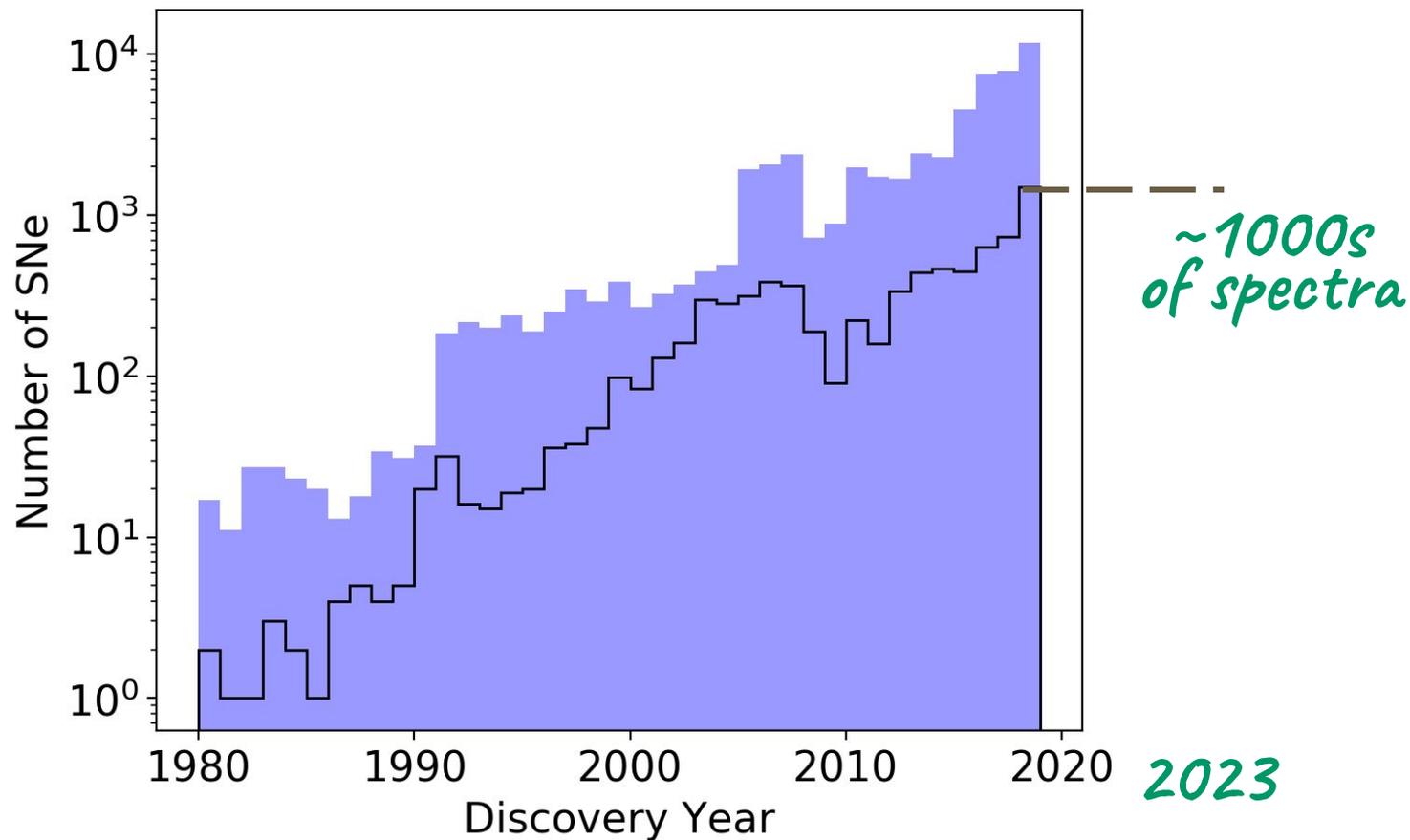
The Large Synoptic Survey Telescope



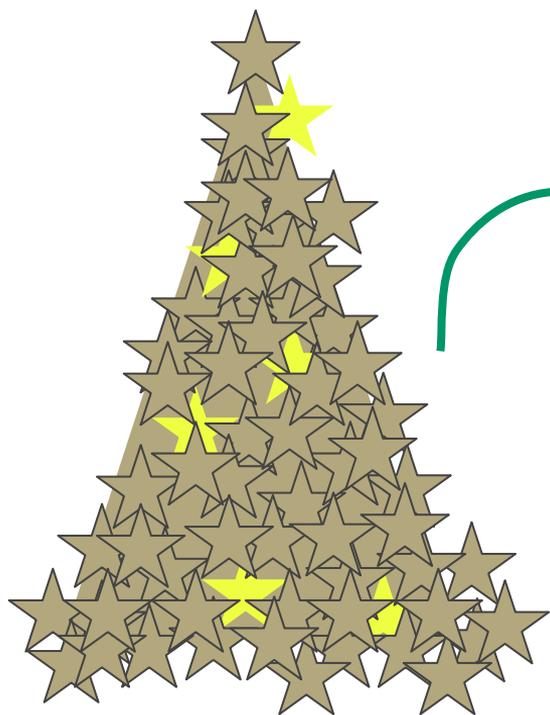
LSST will discover >1 million supernovae annually!



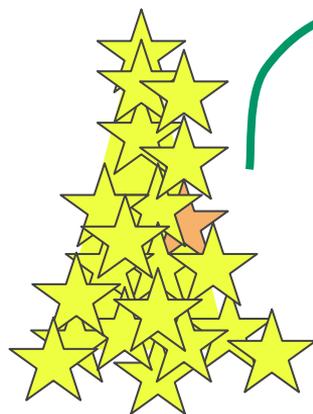
We will follow up *some* of these supernovae



The LSST Needles & the Haystack



~Million SNe / Year



~1000s / Year

With spec. classification



~100 SNe we actively follow
with other resources

A Christmas list for SN classification:

1. Meaningful feature extraction which can handle noisy, sparse data
2. Feature extraction which can utilize unclassified data
3. Classification which can work on incoming data
4. *A method which can search for needles in real time*

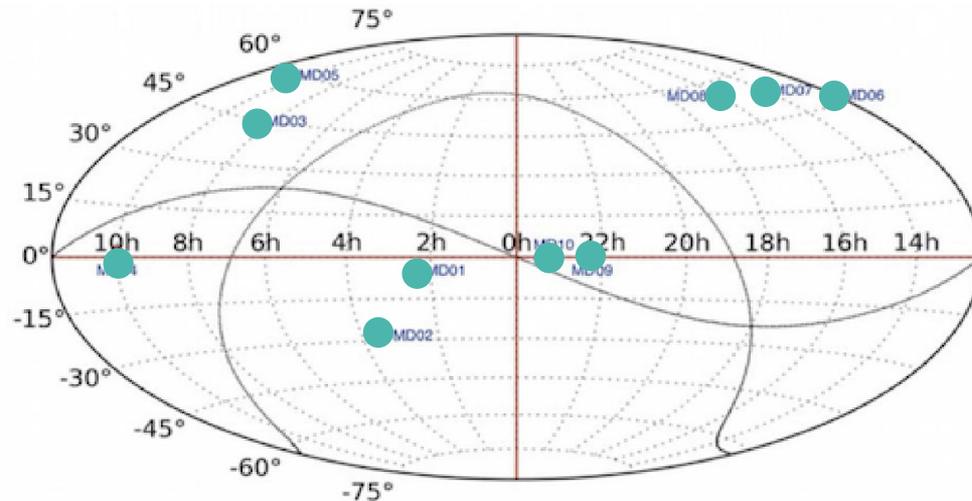
A Christmas list for SN classification:

1. Meaningful feature extraction which can handle noisy, sparse data
2. Feature extraction which can utilize unclassified data
3. Classification which can work on incoming data
4. *A method which can search for needles in real time*

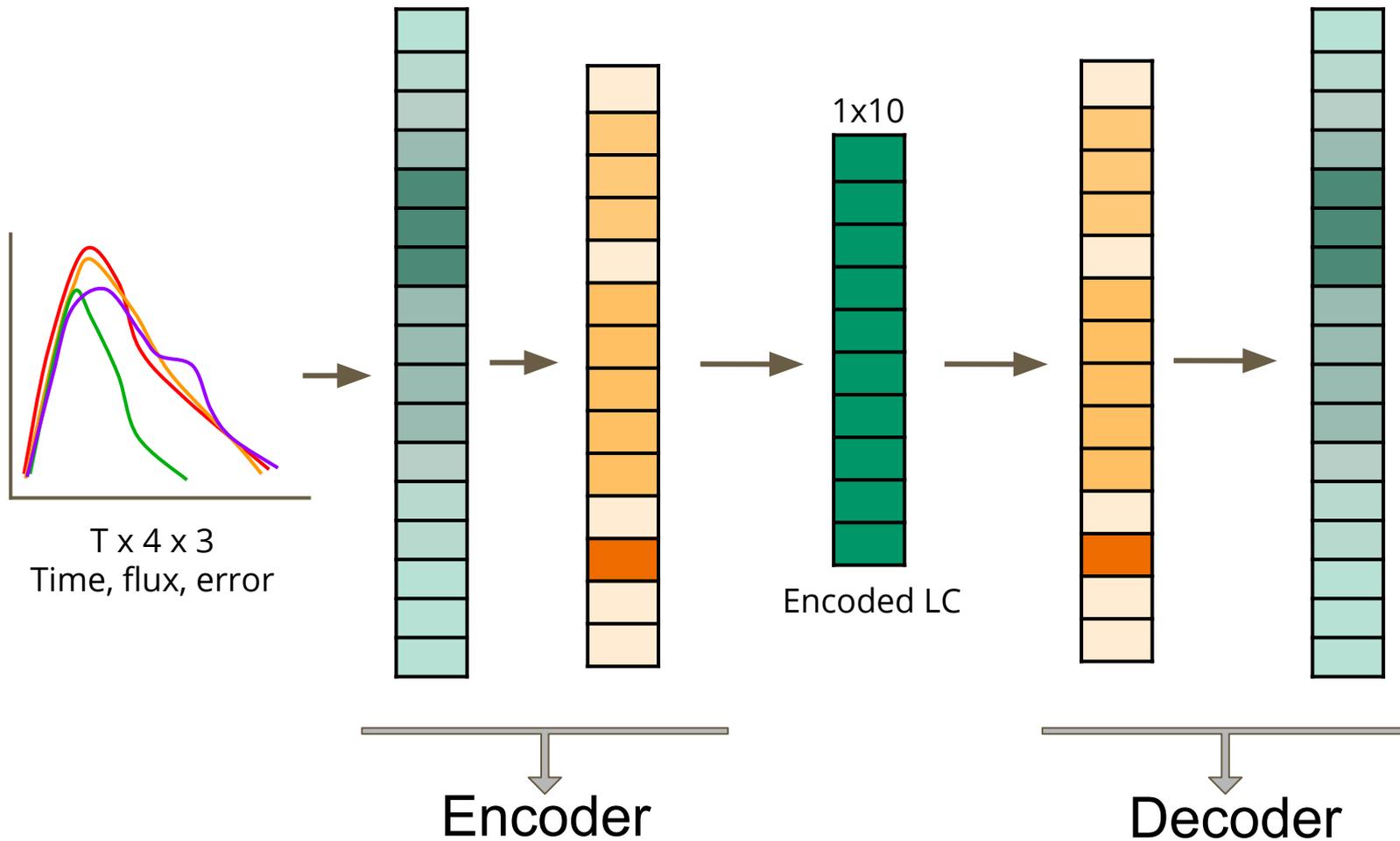
Recurrent neuron-based autoencoder

Pan-STARRS Medium Deep Survey is a milliLSST

- ~5200 SNe-like transients in PS1 MDS (Jones+2017)
- ~**3200 SNe** have host redshift measurements
- ~**520 SNe** are spectroscopically classified with host redshift measurements

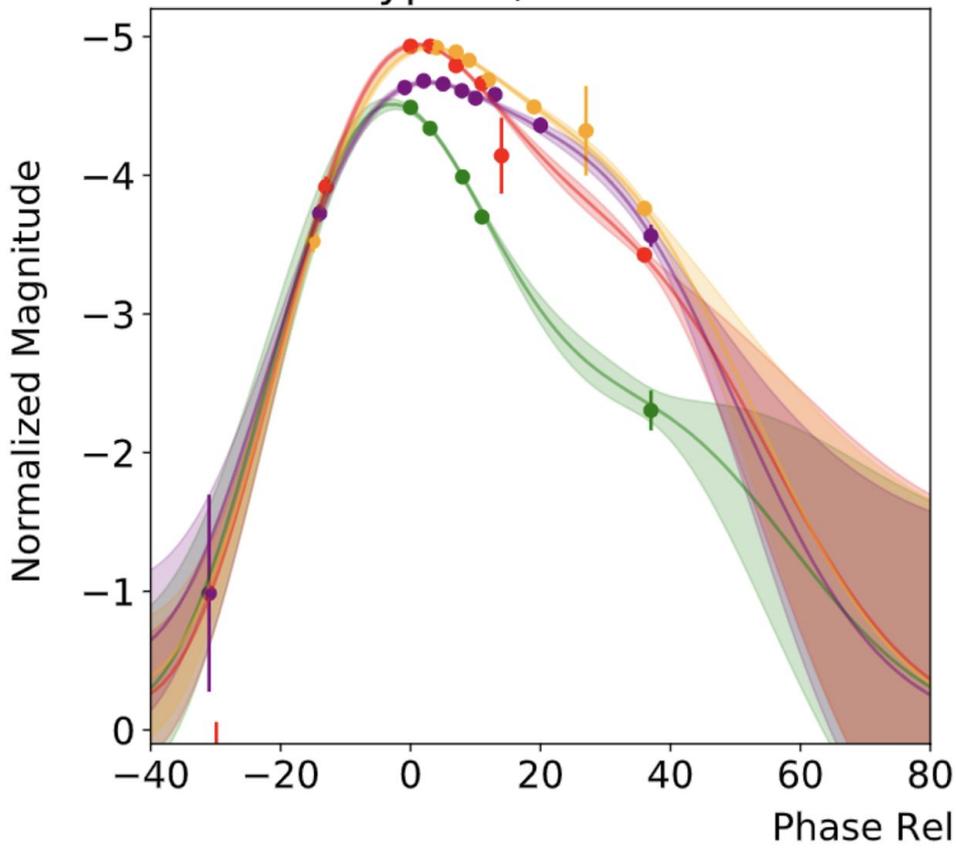


A semi-supervised method to encode/classify SNe

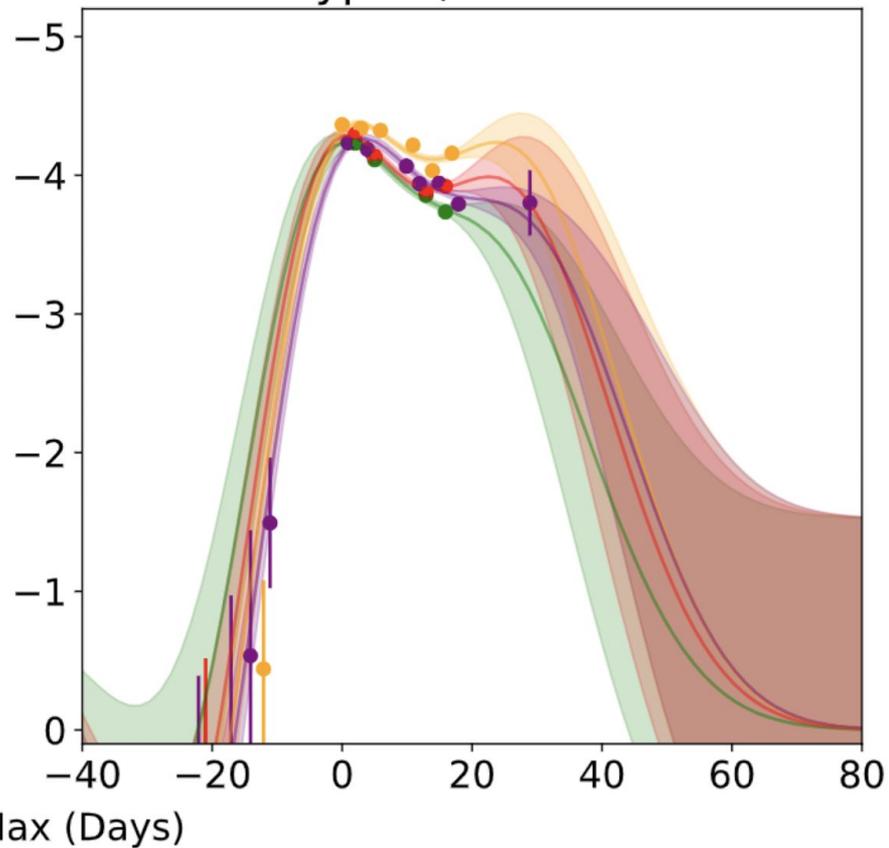


Use a GP to deal with uneven sampling in filters

Type Ic; $z=0.119$

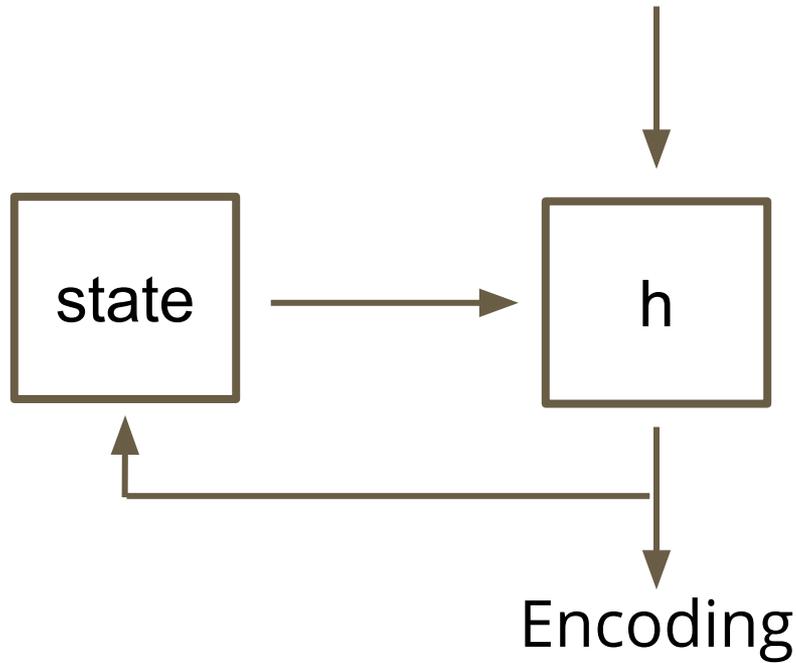


Type II; $z=0.108$

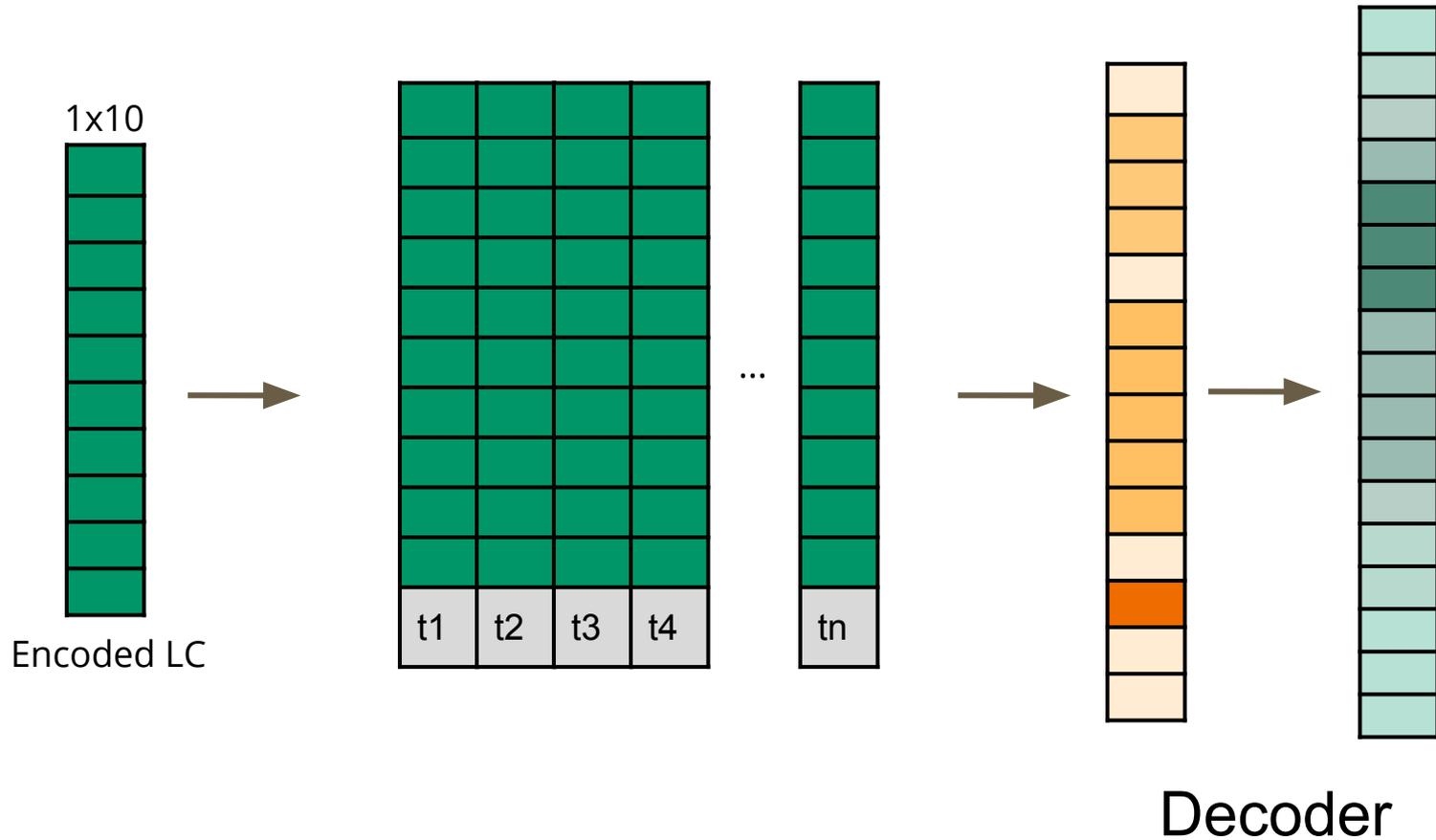


Recurrent neurons update the encoded light curve

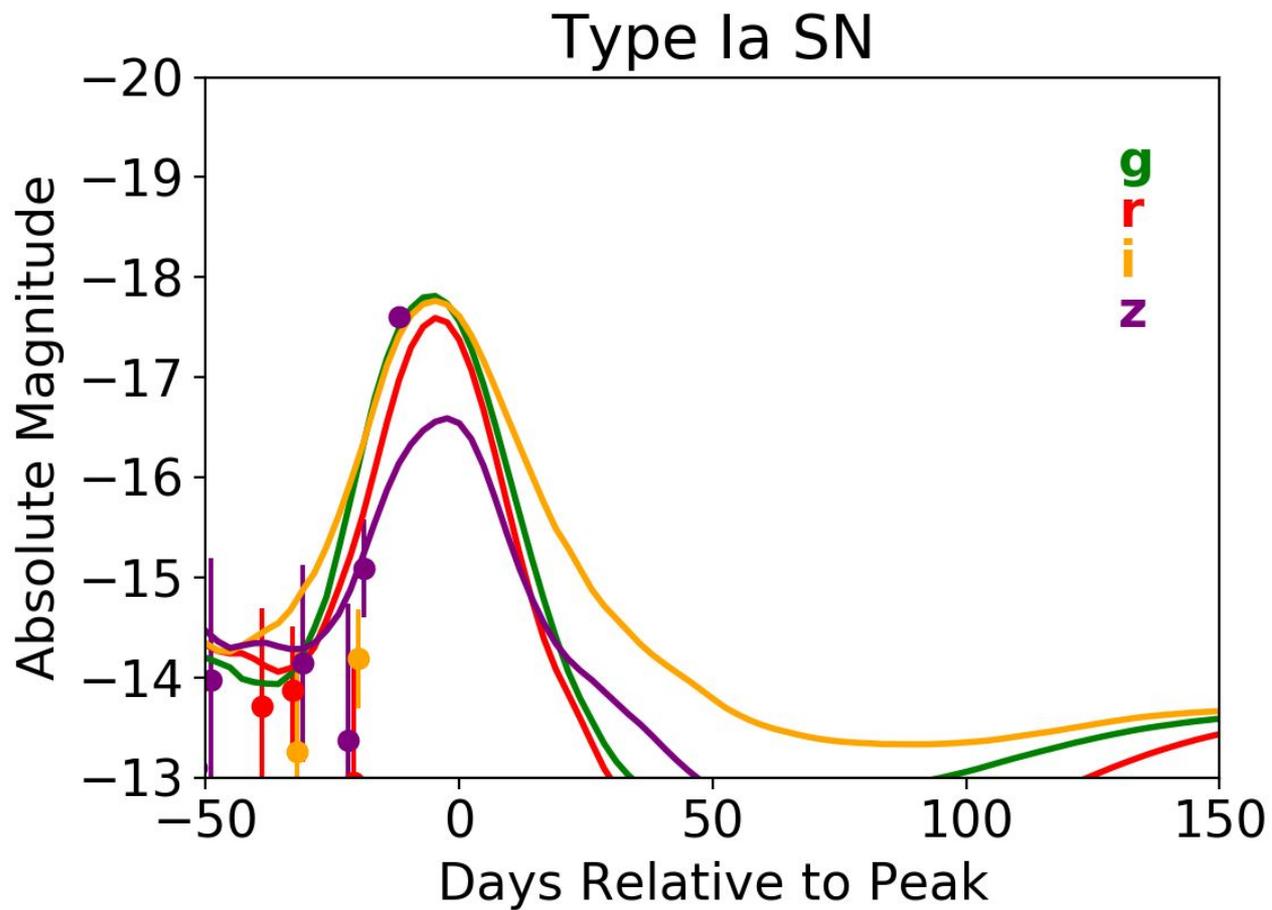
Input: $[T, F_{g'}, F_{r'}, F_{i'}, F_{z'}, \sigma_{g'}, \sigma_{r'}, \sigma_{i'}, \sigma_{z'}]$



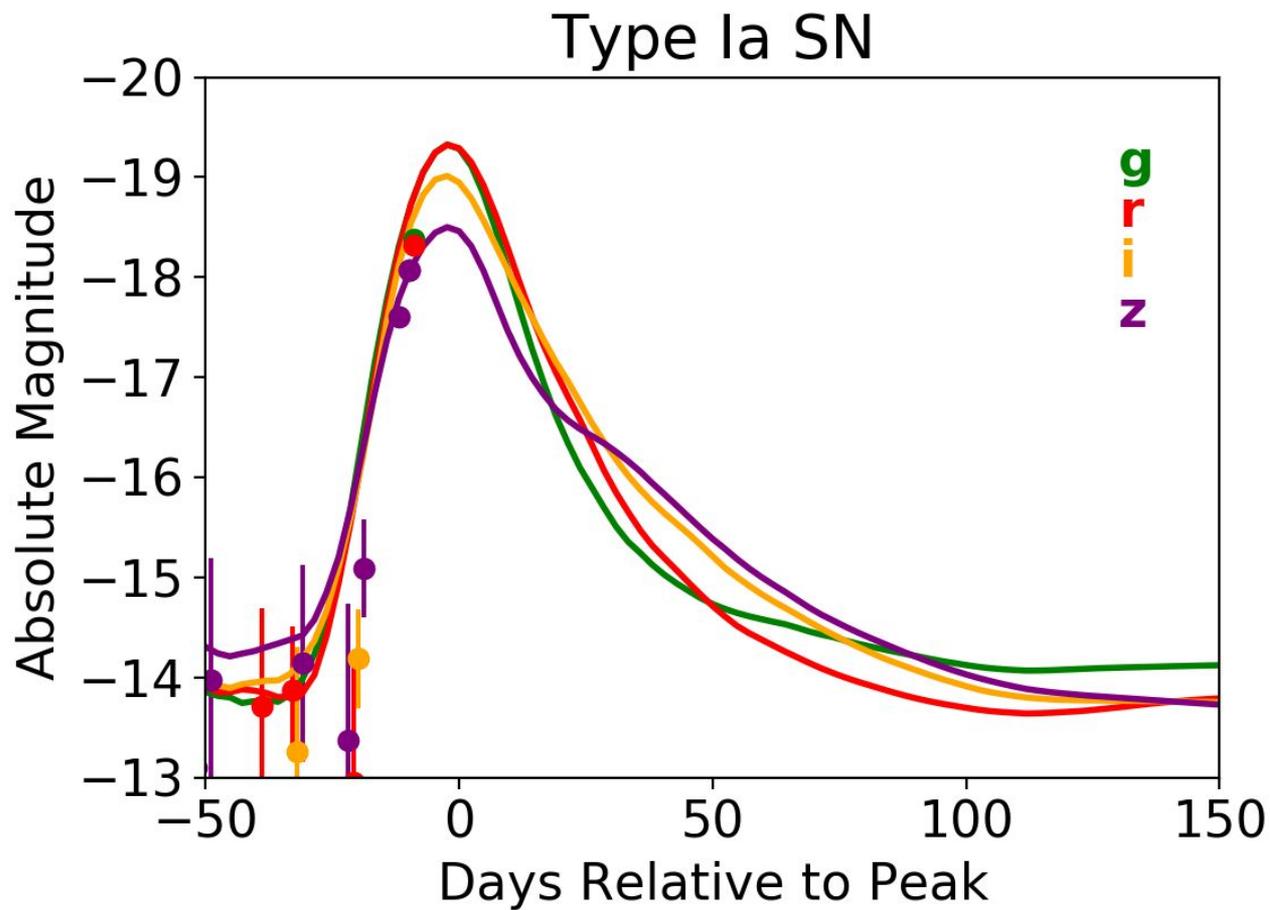
Repeat encoded LC with a new set of times



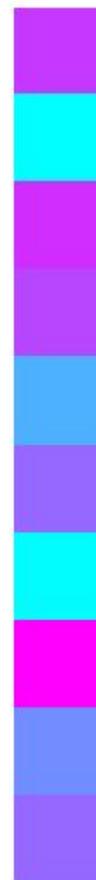
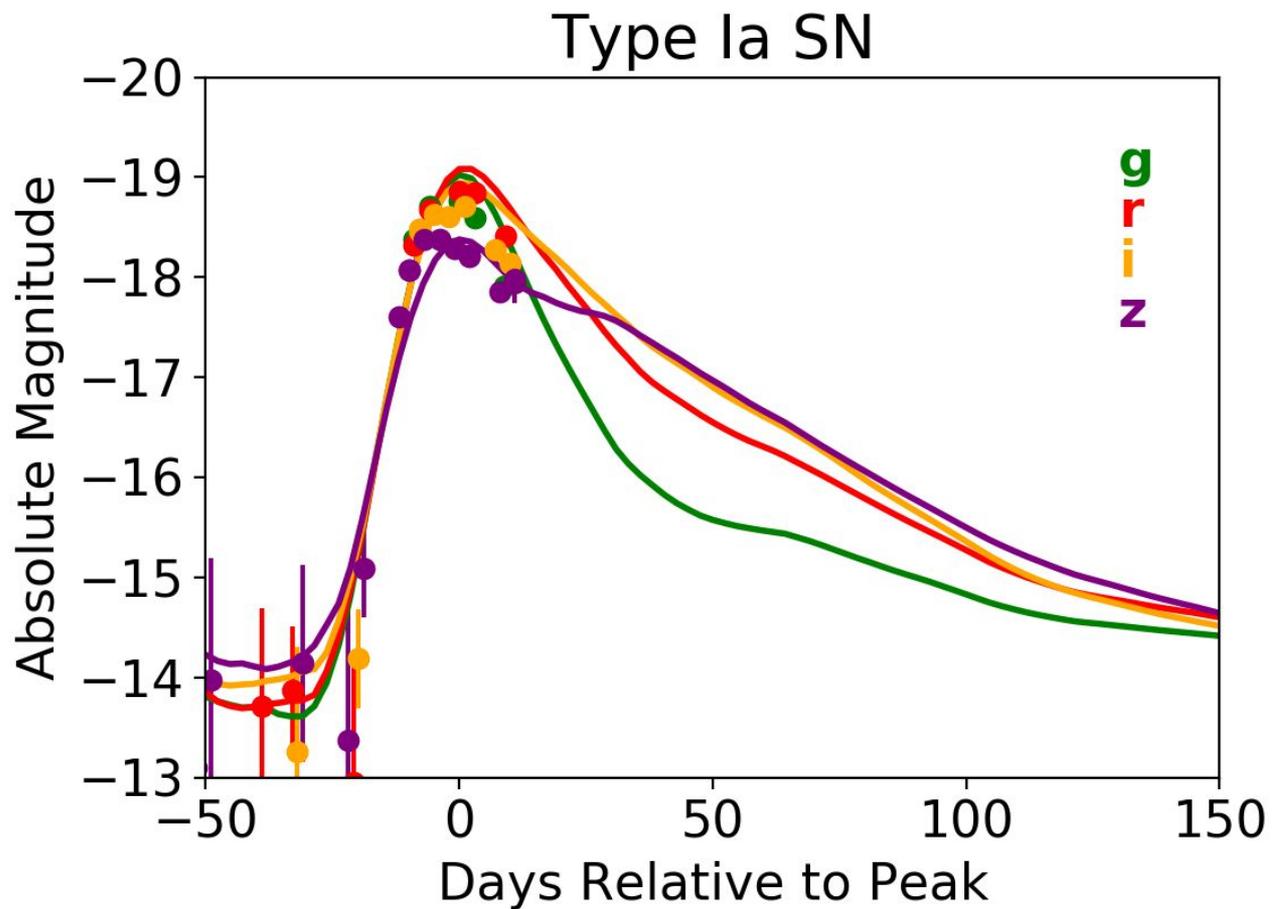
Decoded light curve updated with new data



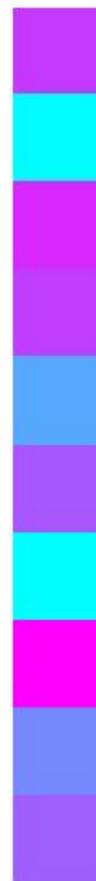
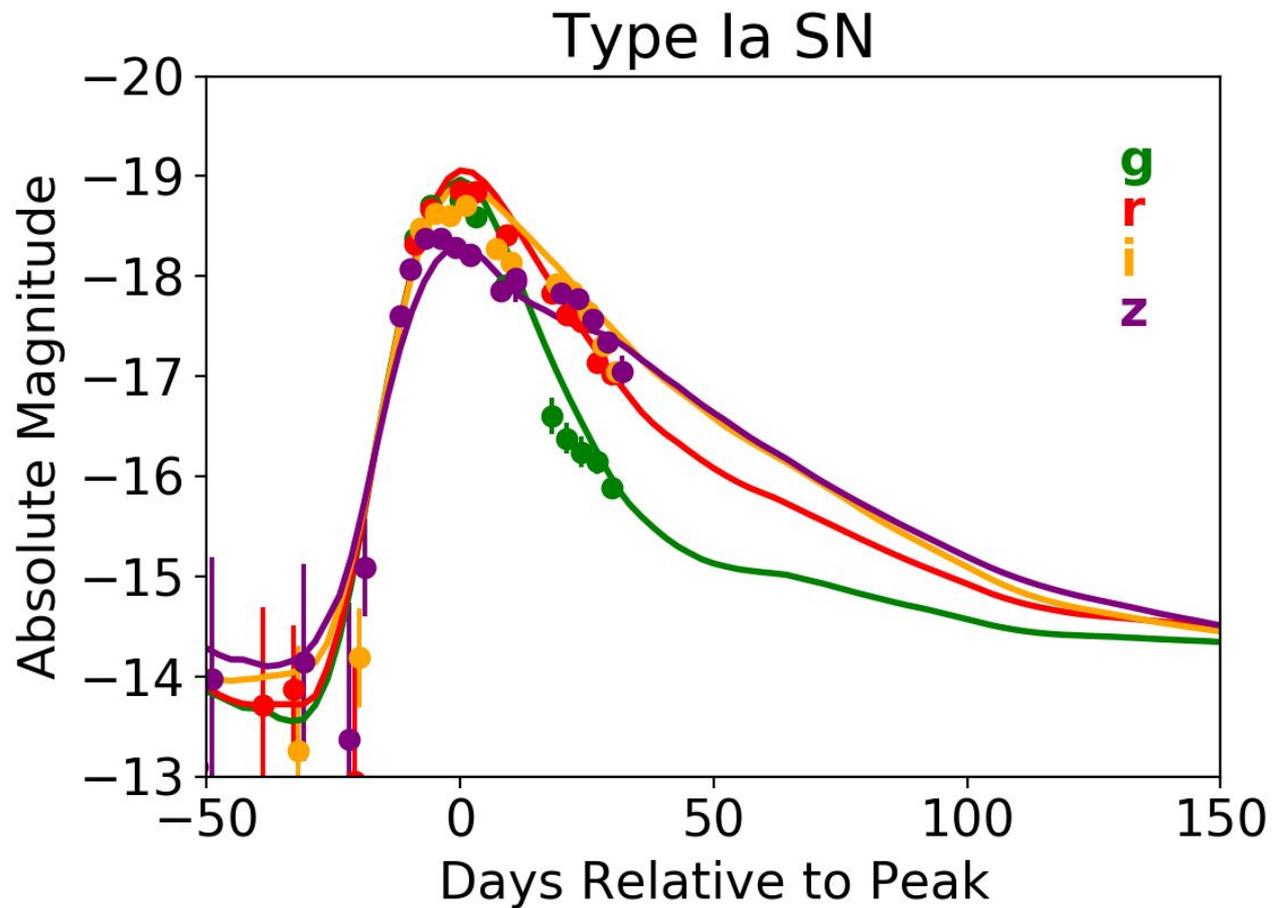
Decoded light curve updated with new data



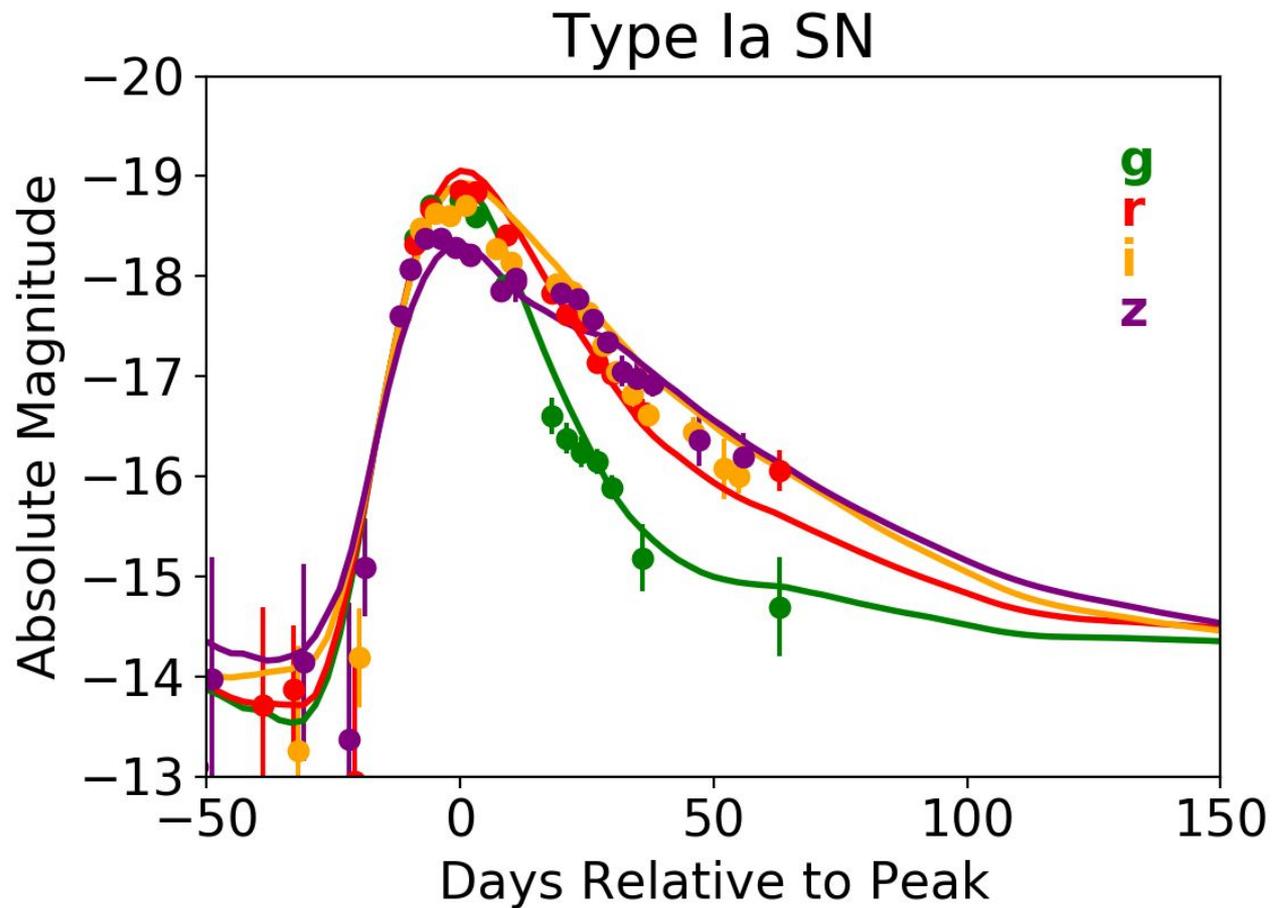
Decoded light curve updated with new data



Decoded light curve updated with new data



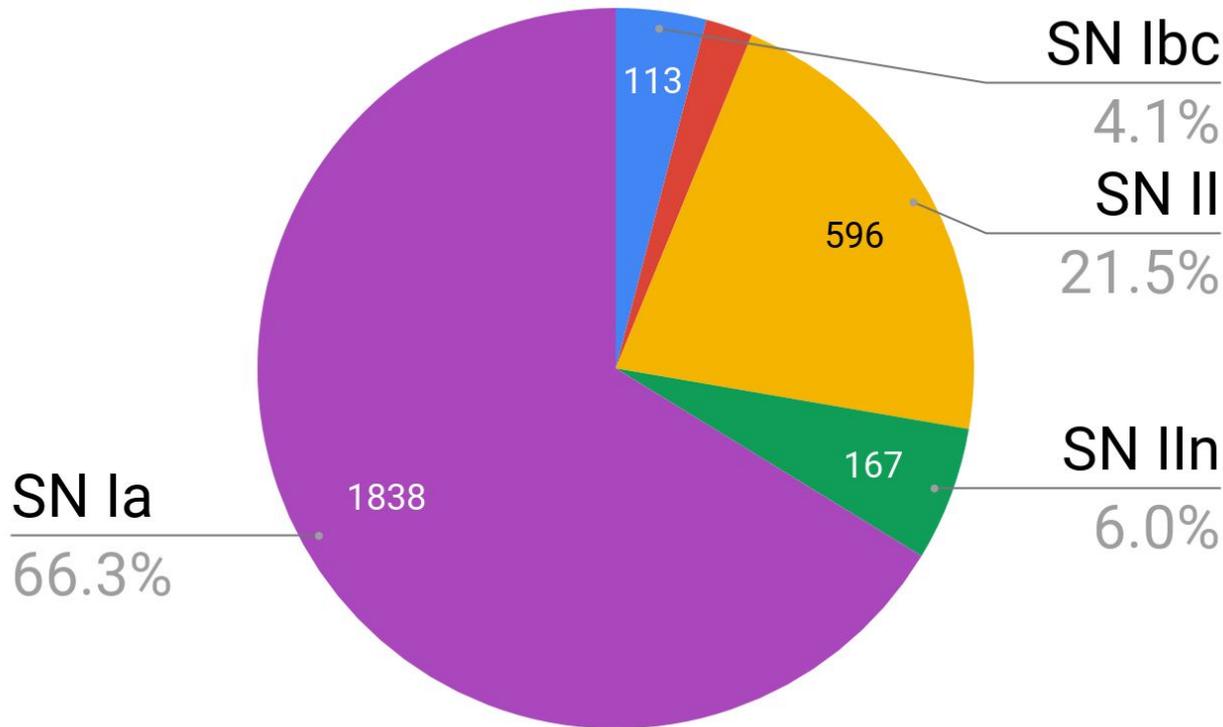
Decoded light curve updated with new data



Why use a RNN autoencoder?

- Semi-supervised methods allow us to use information from *the full dataset*
- We can extract *unique, nonlinear* features directly from the light curves
- Actively makes *forecasting* predictions, which may be used to hunt for anomalies aka **the needles**

Using a random forest classifier, we classify the full sample of 3200 SNe



Time-domain Astrophysics in the Era of Big Data

- LSST will bring TDA into a new era of big data, thanks to both a deep and wide survey strategy
- LSST light curves will be noisy and sparse, but simple features correlate with underlying physics
- RNN-based AEs are a promising strategy to classify SNe in real time
- RNN-based AEs *may* be a promising strategy for real time anomaly detection

CLASSIFICATION

RAPID

PELICAN

Wavelet
decomposition

Online
learning

avocado

High la
purity!

PLAsTiCC

SNPCC

Image-
based CNN

<https://tinyurl.com/transienttable>

Do we have a suitable training set for classification?



PLAsTiCC

Real datastream!

gr(i) filters

Depth ~21 mag

Simulated dataset

LSST filters/cadence

see e.g., Bellm+ 2019; Kessler+ 2019