

I. Introduction

- Two distinct classes of supernova (Type I & II)
- Upcoming surveys will observe many supernova; classification needs to be robust and logical
- Spectroscopic classification lacks consistency and clarity (especially sub-typing)

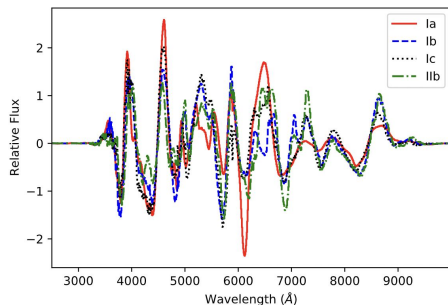


Figure 1. The mean continuum-subtracted flux values of four of the supernova classes.

II. Tagging and Classifying

- Assign spectra tags based on their features
- Sets of tags can be used to define classes
- Not template fitting; only concerned with features in a given spectrum
- Tags assigned using logistic regression
- Tags are passed to a neural network which associates certain tags with certain classes
- Neural network trained and tested on template spectra

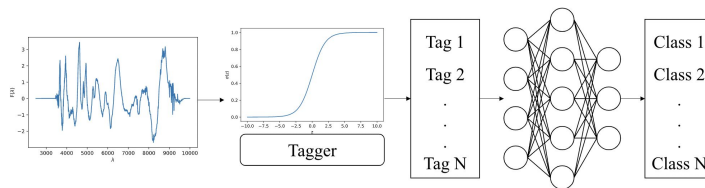


Figure 2. A simplified diagrammatic overview of the general architecture used by STag.

III. Results

- First tested against unseen template data
- Fig. 3 shows that STag is capable of accurately predicting classes

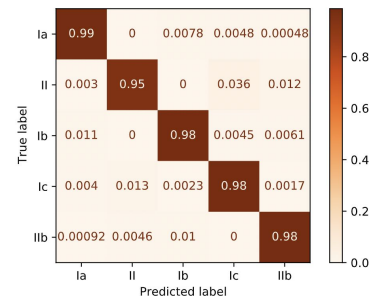


Figure 3. The normalised confusion matrix comparing the predicted class made by STag to the 'true' class of the supernova.

- Tested against previously machine classified OzDES spectra, achieving 81% agreement
- Inclusion of more tags, sub-types, and other transients/non-supernova entities

IV. Further Reading

- <https://github.com/wdavison909/STag>
- <https://arxiv.org/abs/2108.10497>



Follow this QR code to listen to a brief description of STag.