

# Computer Science approach to the stellar fabric of violent starforming regions in AGN

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**Abstract.** In order to analyse the large numbers of Seyfert galaxy spectra available at present, we are testing new techniques to derive their physical parameters fastly and accurately.

We present an experiment on such a new technique to segregate old and young stellar populations in galactic spectra using machine learning methods. We used an ensemble of classifiers, each classifier in the ensemble specializes in young or old populations and was trained with locally weighted regression and tested using ten-fold cross-validation. Since the relevant information concentrates in certain regions of the spectra we used the method of sequential floating backward selection offline for feature selection.

Very interestingly, the application to Seyfert galaxies proved that this technique is very insensitive to the dilution by the Active Galactic Nucleus (AGN) continuum. Comparing with exhaustive search we concluded that both methods are similar in terms of accuracy but the machine learning method is faster by about two orders of magnitude.

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## 1. Introduction

In the last decade we have witnessed a world wide boom in projects that provide clues to elaborate a picture to describe the evolution of the Universe. They are mostly large observational projects that generate huge amounts of data (2dF, 6dF, SDSS, 2MASS, VIRMOS, DEEP2... INAOE/UMASS will also contribute after the expected first light of the Large Millimeter Telescope LMT/GTM in the near future). We are confronted with a new situation in Astrophysics for which, as a community, we are inadequately prepared.

The large number of new facilities coupled to technological advances both in sensors and storage and in automation of data acquisition, plus a new generation of telescopes covering from mm to X-Rays and also having multiplexing capabilities, create an unprecedented challenge: the need to unify data, obtained both from ground and from space forming MEGADATASETS, which lack standardization and, even worse, each database uses different access software.

There is a revolution in the way we do Astrophysics in that we need to search for and develop applications to automatically (and intelligently) manage, query, visualize, analyse, etc. the whole space and variety of large datasets (including artificial or synthetic ones). This is particularly relevant for places like Mexico, with restricted access to large observing facilities.

Recent spectroscopic surveys of nearby AGN have proven that a large fraction show high-order hydrogen Balmer absorption lines in the near-UV. These features are characteristic of young stars and therefore represent strong evidence of recent star formation in the nuclear regions of these galaxies.

From a theoretical point of view, it is very important to determine the age of these starbursts, in order to understand the nature of the starburst-AGN connection and galaxy formation and evolution. The characterization of the nuclear star forming region (its age and mass) is very difficult to achieve in AGN, due to the contamination of the nuclear stellar absorption lines by the AGN component itself. The recent release of high-resolution spectra of a large number of galaxies by the Sloan Digital Sky Survey (SDSS) consortium allows spectroscopic studies to be performed now on thousands of galaxies with active nuclei.

As members of an European/Mexican ‘Violent star formation’ network, we have started a project to determine the composition of the stellar population of galaxies from their spectra (Solorio *et al.* 2004) and in particular, the nuclear stellar population of AGN using machine learning methods, that will allow us to segregate old and young population in spectra, and apply the method to large number of objects (e.g. SDSS) at no prohibitive computational cost.

## 2. The method

The method applied here creates and uses an ensemble of classifiers (each one specializes in “young” or “old” stellar populations) to train the system via Locally Weighted Regression, which is much faster than neural-networks). The fact that relevant information concentrates in certain regions of the spectra (high order Balmer absorptions for the young and CaII K lines for the old populations) helps to choose a fast method for future selection. The algorithms are briefly described in § 2.1, 2.2 and 2.3.

### 2.1. *Sequential Floating Backward Selection (SFBS)*

This is a feature selection algorithm that allows to work with non-monotonic data. It constructs in parallel the feature sets of all dimensionalities up to a specified threshold and consists of applying after each feature exclusion a number of features inclusion as long as the resulting subsets are better than those previously evaluated at that level. It makes a dynamically controlled number of iterations and achieves good results without static parameters (Pudil 1994).

### 2.2. *Locally Weighted Regression (LWR)*

This is an instance based learning method; it assumes that instances can be represented as points in an Euclidean space (Schneider & Moore 2000). Its training consists of explicitly retaining the training data and using them each time a prediction needs to be made. LWR performs a regression around a point of interest using only a local region around that point. Locally weighted regression can fit complex functions in an accurate way and data modifications have little impact on the training.

### 2.3. *Ensembles of Classifiers*

An ensemble of classifiers is a group of classifiers trained independently whose outputs are combined in some way, usually by voting (Mitchell 1997). They are normally more accurate than the individual classifiers that make it up.

Several algorithms tested rendered comparable results regarding precision, but the machine learning used was faster by up to two orders of magnitude.

## 3. Results and Conclusions

The discovery of high order Balmer absorption lines in the near UV, characteristic of young stars has been reported in many spectra of Seyfert galaxies (Gonzalez-Delgado

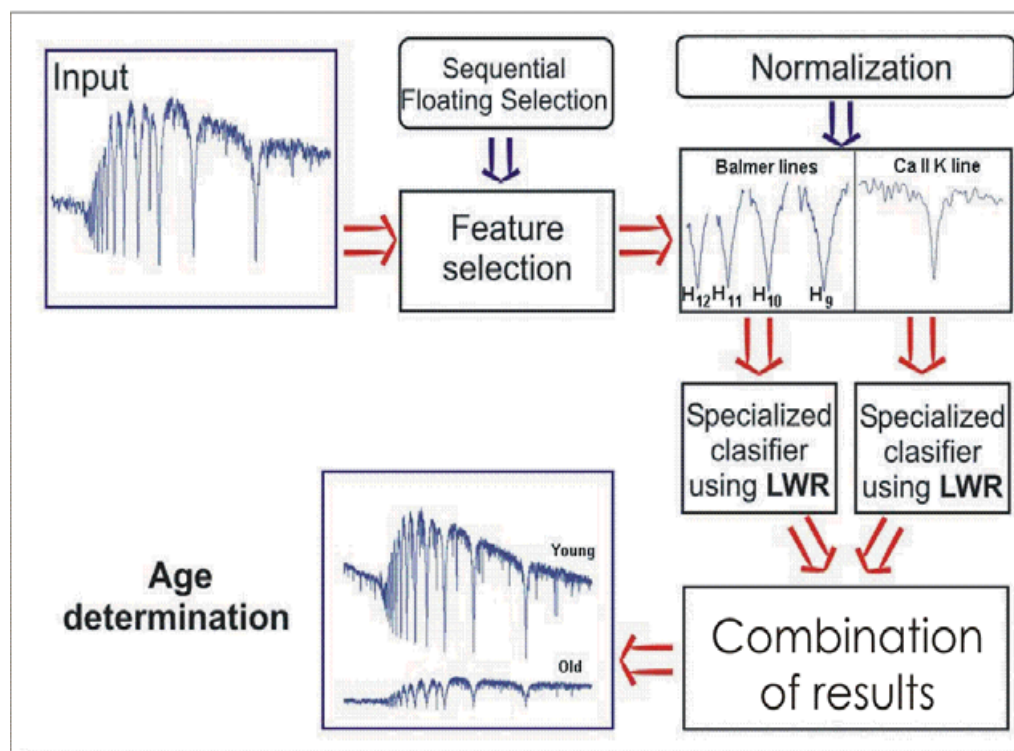


Figure 1. Schematic process of age identification.

et al. 1999; Joguet et al. 2001; Torres-Papaqui et al. 2004). The analysis of these features allows to estimate the age and mass of these nuclear starbursts, task that is otherwise very difficult to achieve due to contamination by the non-stellar component.

Our first step in this investigation was to apply the machine learning method to synthetic data to study its efficiency and validity, following the diagram shown in Figure 1 (Estrada-Piedra et al. 2004). The ‘data’ consisted of fourteen high resolution synthetic spectra (Bressan, Bertone and Rodríguez, private communication) considering ten levels of dilution by a power law and with added Gaussian noise. The age calibration was performed using young starburst ( $10^7 - 10^{8.7}$  yrs) and old bulge ( $10^8 - 10^{9.6}$  yrs) spectra characterized by the Balmer lines and the CaII K line respectively.

The synthetic spectra, with a  $0.2\text{\AA}$  sampling, covers a range of  $3600 - 5300\text{\AA}$ . The accuracy achieved in recovering the input age is  $0.3$  dex in  $\log(\text{age})$ . To our delight, the technique proved to be quite insensitive to dilution by a featureless continuum.

The experimental results show the efficiency of the automatic learning method applied to astronomy. Our next step is to apply the method to a sample of real data and to define the training set with a young, intermediate and old age components instead of only two (Torres-Papaqui et al. 2004).

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