Mathematical characterization of galaxies using IFS data: I - analysis of spectral information

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INAOE, 01/09/2014







Outline

The traditional approach

2 The astrostatistical approach

- Spectra as multi-dimensional data
- Classification

3 What does it have to do with astronomy?

- The ASK spectral classification
- A 'spectroscopic sequence' of galaxies?



The traditional approach

Characterization of spectra

Physical modelling

- Mean stellar age/SFH
- Stellar/gas metallicity
- Ionization parameter
- Presence/contribution of AGN/shocks/...

• ...

Feature extraction

- Broadband magnitudes/colours
- Emission lines/ratios
- Lick indices
- Redshift/velocity/moments

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Spectra as multi-dimensional data Classification



The astrostatistical approach

Spectra as multi-dimensional data Classification

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Spectra as multi-dimensional data



Spectra as multi-dimensional data Classification

Principal Component Analysis (PCA)

Find "principal axes" of the distribution

- Compute centre
- 2 Compute covariance matrix
- Find (largest) orthogonal eigenvalues/vectors

What for?

- Dimensionality reduction
- Physical interpretation?



Spectra as multi-dimensional data Classification

Principal Component Analysis (PCA)

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What for?

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- Physical interpretation?



Spectra as multi-dimensional data Classification

Principal Component Analysis (PCA)

Advantages

- Simple
- Fast
- Scalable

Disadvantages

- Uncorrelated ≠ independent
- Non-linear manifolds
- Metric-dependent

Spectra as multi-dimensional data Classification

Independent Component Analysis (ICA)

x and y are...

- uncorrelated $\Leftrightarrow \operatorname{Cov}(x, y) = 0$
- independent $\Leftrightarrow p(x, y) = p(x)p(y)$



Spectra as multi-dimensional data Classification

Independent Component Analysis (ICA)



Spectra as multi-dimensional data Classification

Independent Component Analysis (ICA)



Spectra as multi-dimensional data Classification

Manifold learning



Spectra as multi-dimensional data Classification

How to define a metric?



Spectra as multi-dimensional data Classification

Field Estimator for Arbitrary Spaces



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Spectra as multi-dimensional data Classification

Minimal Spanning Tree (MST)



Spectra as multi-dimensional data Classification

Minimal Spanning Tree (MST)



Spectra as multi-dimensional data Classification

Summary

What to take home?

- A spectrum is a N_{λ} -dimensional vector
- Global distribution (e.g. PCA, ICA)
- Local distribution (e.g. FiEstAS, MST)

Spectra as multi-dimensional data Classification

Classification



Spectra as multi-dimensional data Classification

Classification





Classification

Spectra as multi-dimensional data Classification

Supervised classification

- Nearest neighbour(s)
- Naive Bayes

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- Decision trees/forests
- Support Vector Machines
- Artificial Neural Networks

Unsupervised clustering

- ...
- k-means
- Expectation-Maximization
- HOT+FiEstAS

Spectra as multi-dimensional data Classification





Spectra as multi-dimensional data Classification





Spectra as multi-dimensional data Classification





Spectra as multi-dimensional data Classification



Spectra as multi-dimensional data Classification

k-means



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Spectra as multi-dimensional data Classification



k-means

Advantages

- Simple
- Fast
- Scalable

Spectra as multi-dimensional data Classification

Disadvantages

- Metric-dependent
- Initialization
 - number of classes
 - random seed
- Similar class sizes

Spectra as multi-dimensional data Classification

Expectation-Maximization

Hard classification
 $\rho_A(x) > \rho_B(x) \Rightarrow x \in A$ Gaussian mixture modelSoft classification
 $p(x \in A) = \frac{\rho_A(x)}{\rho_A(x) + \rho_B(x)}$ $e^{-\frac{(x-x_A)^2}{2\sigma_A^2}}$ E-M algorithm
Image: Second seco

Iterate to convergence

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Kernel density estimation



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Spectra as multi-dimensional data Classification

Hierarchical Overdensity Tree



Spectra as multi-dimensional data Classification

Summary

Classification

- Supervised:
 - for boring things that you don't want to do by hand

Unsupervised:

- for boring things that you don't want to do by hand
- for scientific analysis

The ASK spectral classification A 'spectroscopic sequence' of galaxies?



What does it have to do with astronomy?

The ASK spectral classification A 'spectroscopic sequence' of galaxies?

Spectroscopic classes



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The ASK spectral classification A 'spectroscopic sequence' of galaxies?

Automatic Spectroscopic K-means (ASK)

Definition of ASK classes

- SDSS galaxy spectra
- 2 Rest-frame corrected
- Ormalized to g band
- Feature selection
- k-means

The ASK spectral classification A 'spectroscopic sequence' of galaxies?

Automatic Spectroscopic K-means (ASK)



Sánchez-Almeida et al. (2010)

The ASK spectral classification A 'spectroscopic sequence' of galaxies?

Automatic Spectroscopic K-means (ASK)



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Distribution in spectral space



What do you think?

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The ASK spectral classification A 'spectroscopic sequence' of galaxies?

The MST of ASK classes



Ascasibar & Sánchez-Almeida (2011)

The ASK spectral classification A 'spectroscopic sequence' of galaxies?

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The ASK spectral classification A 'spectroscopic sequence' of galaxies?

A spectroscopic sequence?



Ascasibar & Sánchez-Almeida (2011)

The ASK spectral classification A 'spectroscopic sequence' of galaxies?

More on k-means



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Physical interpretation

Normal galaxies

Normal galaxies form a well-defined, one-parameter sequence in spectral space

Active galaxies

Active galaxies form a separate branch that intersects the main sequence at the green valley

Phases of galactic evolution?

- Star formation
- Optical activity
- Red and dead

The ASK spectral classification A 'spectroscopic sequence' of galaxies?

Summary

What to take home?

- There is a lot of **physical** information in the way data are distributed in multi-dimensional space
- There are several tools to retrieve it

What next?

• Learn more, e.g. http://scikit-learn.org/stable/user_guide.html

http://www.astroml.org/sklearn_tutorial/

• Apply this approach to your problem