

Application of Machine Learning to ATLAS Physics Analysis

BDP(Bachelor's degree project) case of study

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Overview

What is about Machine Learning and models?

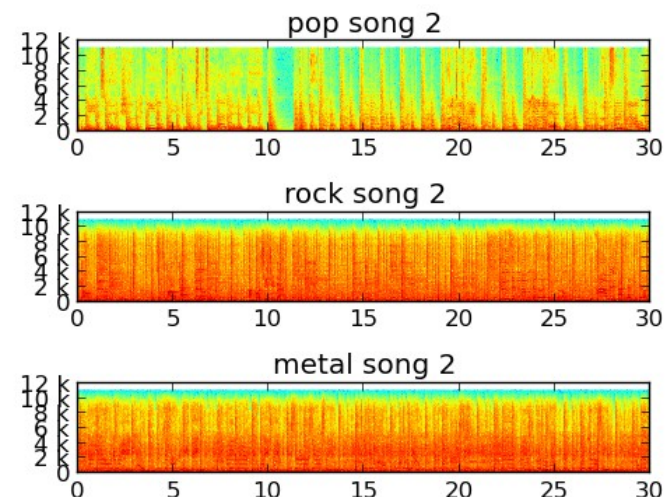
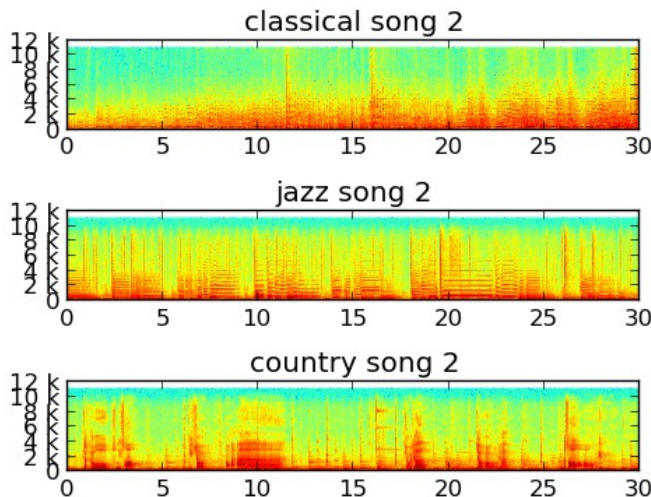
How is related data and model for choosing wisely?

Which are the Machine Learning analysis options?

Machine Learning?

To start:

- **Recognise complex patterns**



- **Ask to data and wait for an understandable answer**



What is a model?

LEARNING

Data

```
100100011101000000101000110111010110
100100111101110000001111100110100100
100001101101111101010011100001101001
111111010000110111001010111100001011
11001111101111111100100001110110110
010000110100110110000110000100010000
010101110011001111011001110100010111
001000010101100101000001000010011110
0111010011111100101110101010111100
100010000101100010101101010111000101
010010000100101011110011100001010000
010110000010011101010010101110110001
011011111010111100010100010100010000
011010011011011010001000101111001101
000101000001100110001100100010010110
100101010100010011100101010101111101
```

Algorithm



Model

$f(\mathbf{x})$

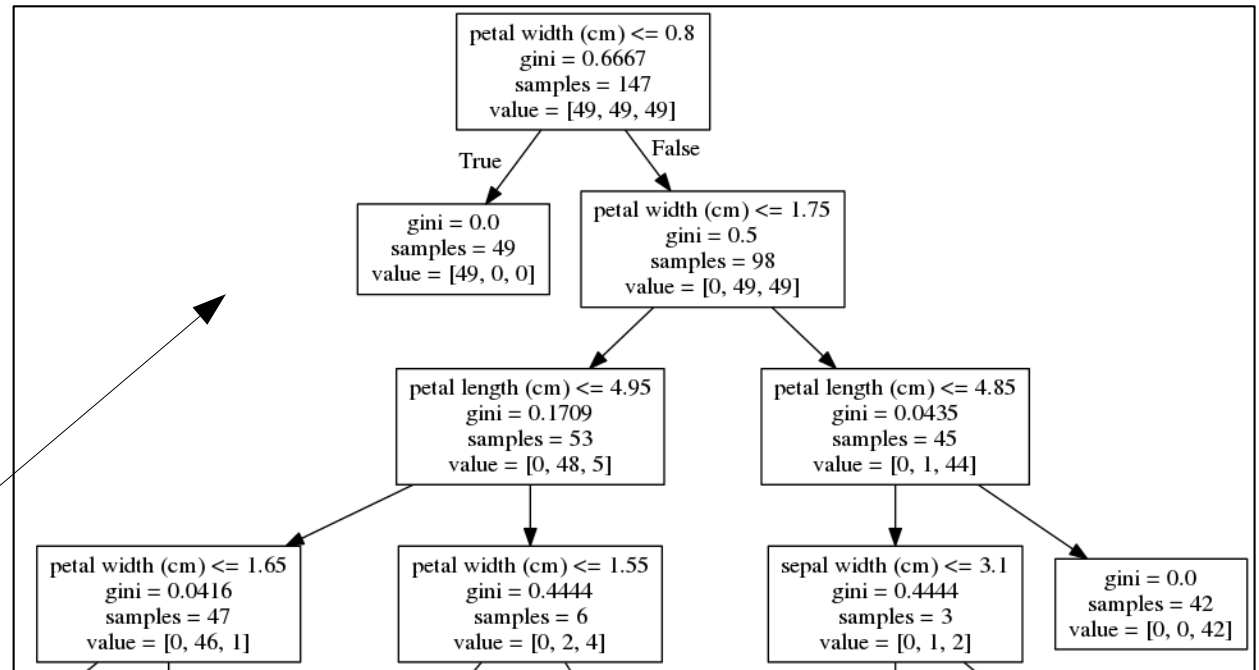
$$P(Y|X) = \theta_1 x_1 + \theta_2 x_2 + \dots$$

What makes a model?

ANALYSIS

```
100100011101000000101000110111010110
100100111101110000001111100110100100
100001101101111101010011100001101001
111111010000110111001010111100001011
11001111110111111100100001110110110
010000110100110110000110000100010000
010101110011001111011001110100010111
001000010101100101000001000010011110
011101001111110010111010101010111100
100010000101100010101101010111000101
010010000100101011110011100001010000
010110000010011101010010101110110001
011011111010111100010100010100010000
011010011011011010001000101111001101
000101000001100110001100100010010110
10010101010001001110010101010111101
```

$f(\mathbf{x})$



Which model should I choose?

Depends on:

- Which **data type** you have
- What **information** you want to extract

Data types

Label Data

sepal length(cm)	sepal width(cm)	petal length(cm)	petal width (cm)	type
5,0	3,3	1,4	0,2	Iris-setosa
7,0	3,2	4,7	1,4	Iris-versicolor
5,9	3,0	5,1	1,8	Iris-virginica

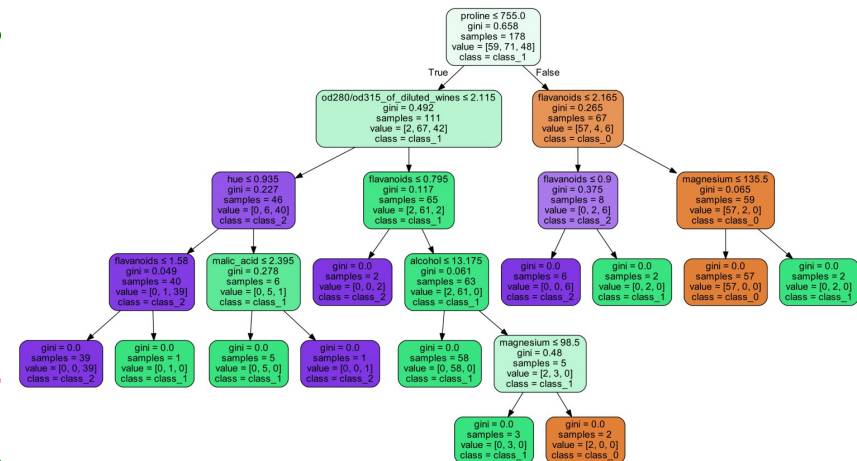
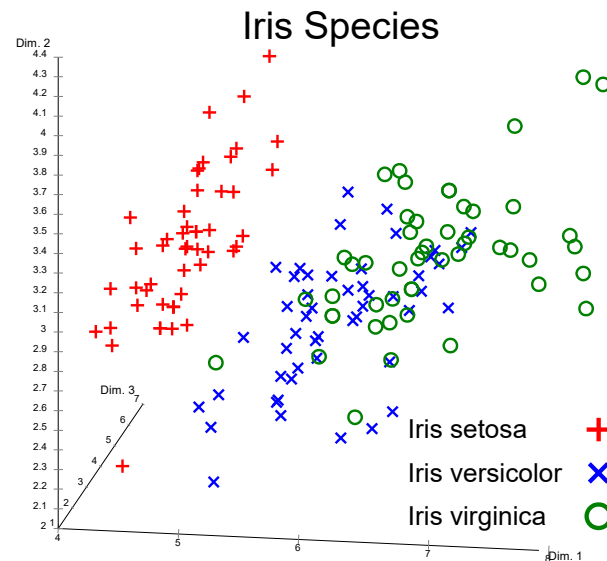
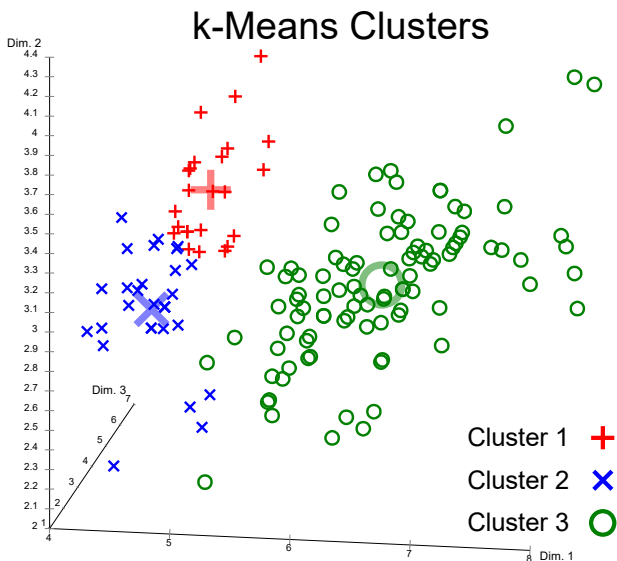
Unlabel Data

Height (cm)	Weight (kg)	Age (years)	Buys (€)
172	80	14	20
150	43	50	400
162	46	34	1000
160	65	89	10

Data types – Learning model type

Label data → Supervised learning

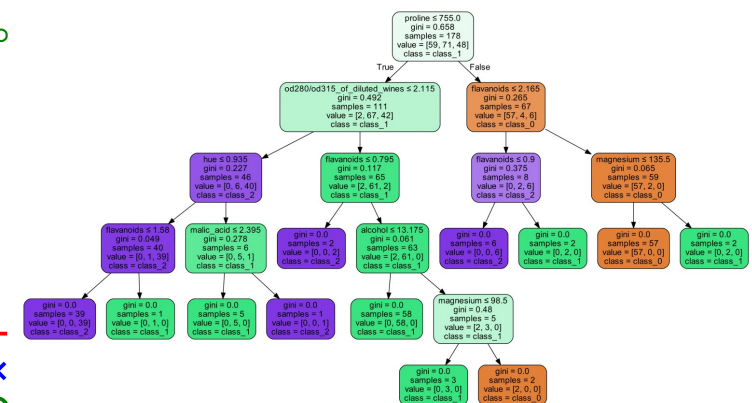
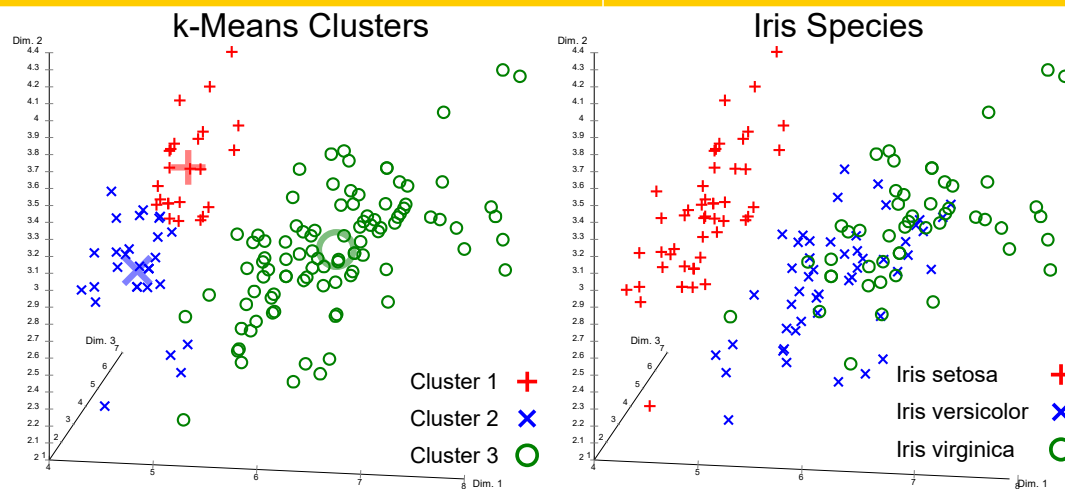
Unlabel data → Unsupervised learning



Unsupervised and Supervised Learning

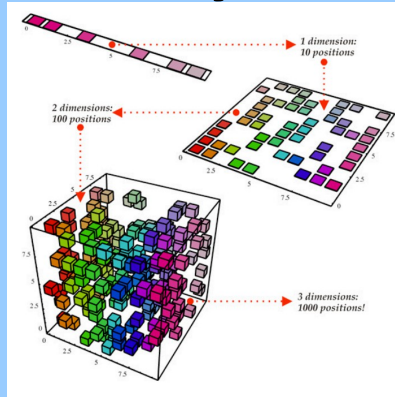
Unsupervised: *interpreting and processing models*
Supervised: *predictive and descriptive models*

	Unsupervised Learning	Supervised Learning
CONCLUSION	-	+
TEST	-	+
FEATURES ANALYSIS	+	-
PROCESS DATA	+	-
CONTROL	-	+

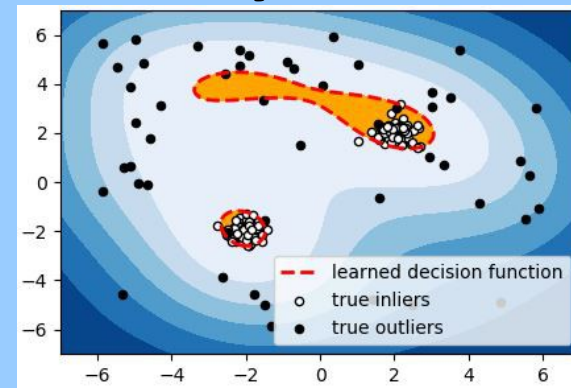


Information from data by ML

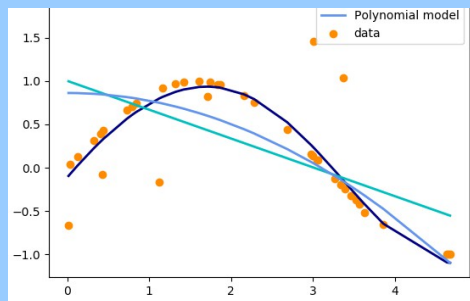
Dimensionality Reduction



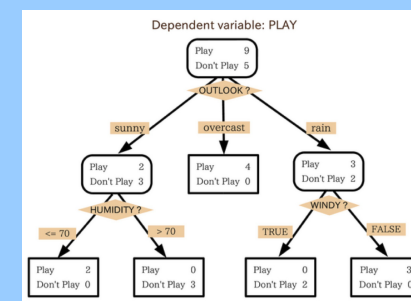
Anomaly detection



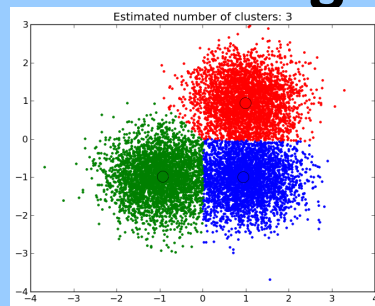
Regression



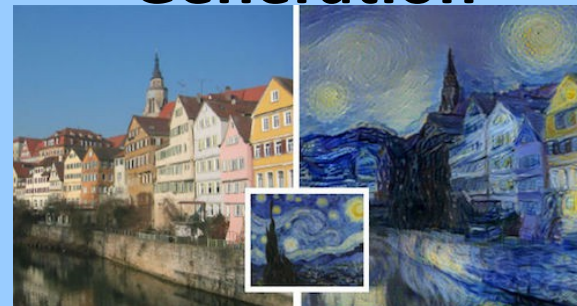
Classification



Clustering



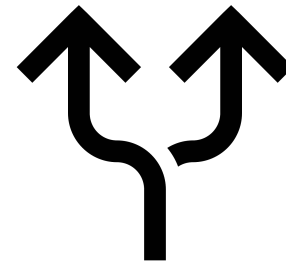
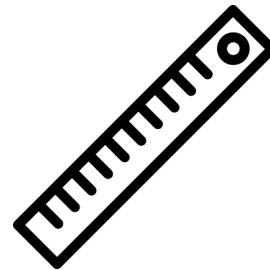
Generation



Machine Learning Workflow

Gather Data > Prepare Data > Choose Model (Train > Evaluate > Hyperparameter Tuning) > Prediction

Prepare Data: *Clean, scale and train/test split*



Choose Model: *Train, evaluate and maximize accuracy*

Our case of study

Pseudorapidity	Missing Momentum	Angle	Invariant mass	Model
2,1245	1,5123	-2,7362	1500	1
0,5787	0,1913	-2,4253	500	1
-1,4872	0,1008	2,9731	750	0
-0,4587	-2,0445	2,8722	1000	1
1,0124	-0,6498	-2,5371	1250	0
0,7652	-1,0118	2,2763	1000	1
-0,1241	1,6875	2,7352	1000	1

Models we have used:

Classification: *Neural Network, Decision Tree, Logistic Regression, etc.*

Models we are going to use:

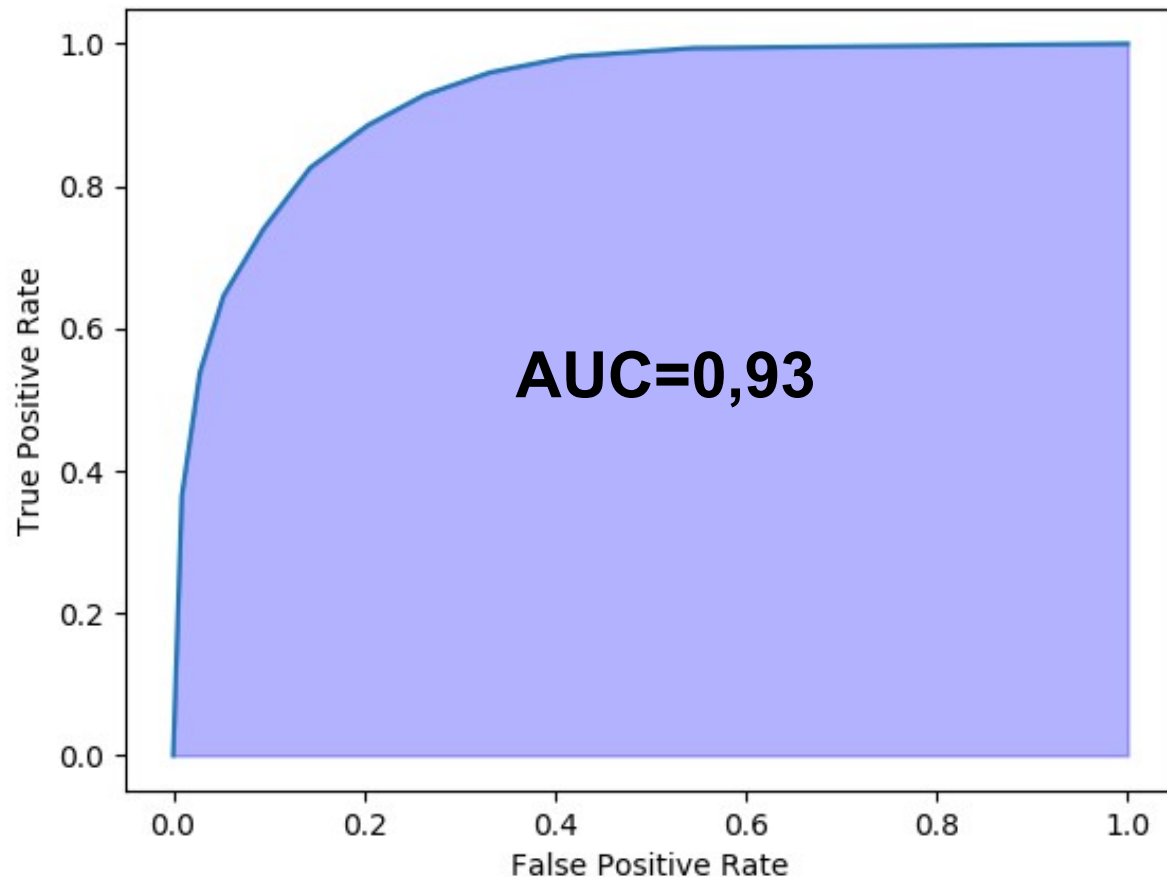
Clustering: *K-means*

Generation: Generative Neural Network

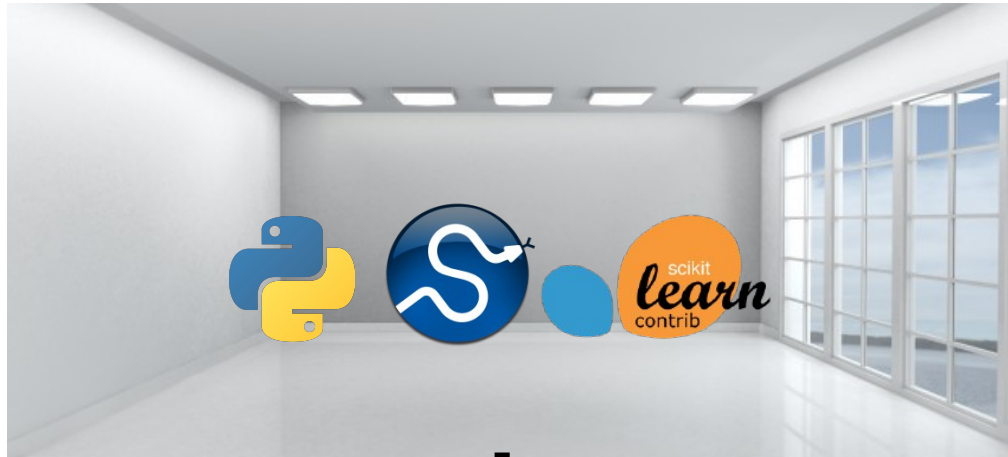
Showcase of Machine Learning

- 28 features to analyse with.
- More than 7,000,000 events.
- 2 Labels.

Classification for 2 labels among 196,000,000 non-integer numbers



Code Laboratories: Enviroments and Dockers



Local computer



Shared enviroment



Companion local computer

Code Laboratories

Anaconda



Enviroments which can be copied and shared,
but you **need some installation**.

Easiest and more common

Docker



docker

Containers that can be copied and shared, and
you have **NO need of installing** nothing at all.

Lightest and more powerful

Open Fronts

Nowadays our project has these open fronts:

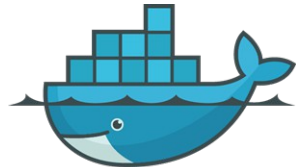


- Make real use of **SWAN** on our study



- Put together ML models trained in **Python and ROOT**

- Implement **Docker** as a common tool on our



docker

development



NVIDIA

- Improve our analysis on **Neural Network** using **GPU's**

To sum up:

Machine Learning is used for **pattern analysis** and **prediction**

Docker and **Anaconda** are really good friends for **Science research** using Machine Learning

References to learn more...

Machine Learning explanations and examples:

<https://machinelearningmastery.com/>

<http://scikit-learn.org>

<https://stackoverflow.com/>

Datasets to practise:

<https://archive.ics.uci.edu/ml/index.php>

<https://www.kaggle.com/datasets>

**Thanks for
your attention!**