



Eeg features extraction

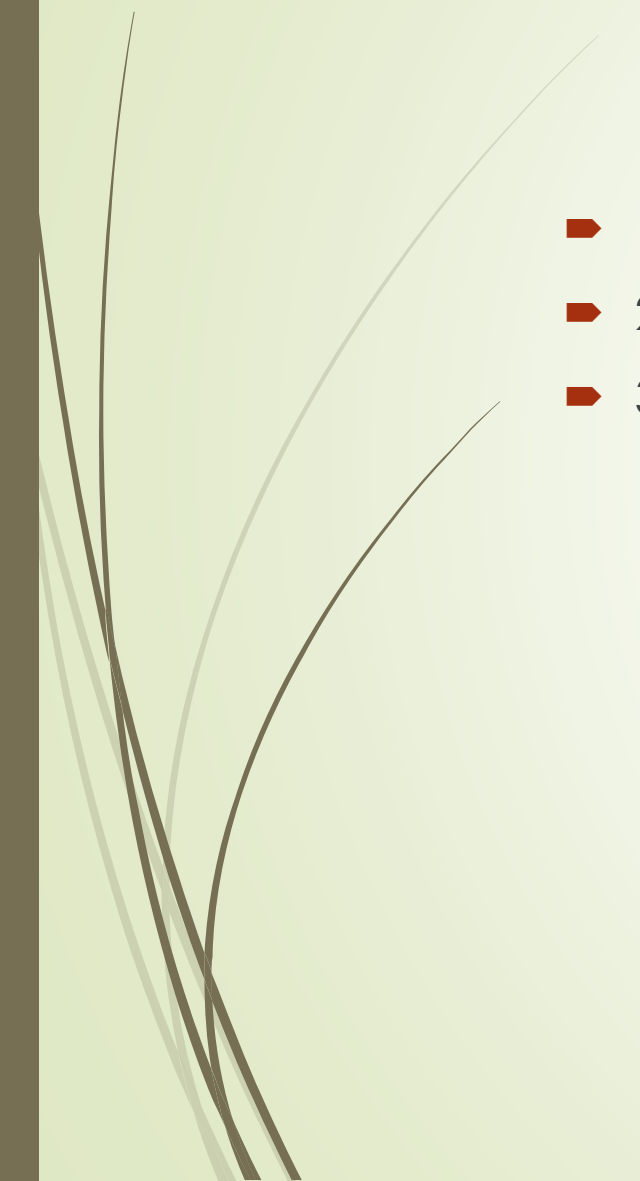
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Summary

- 1. Introduction
 - 2. Problem explanation
 - 3. Features extraction
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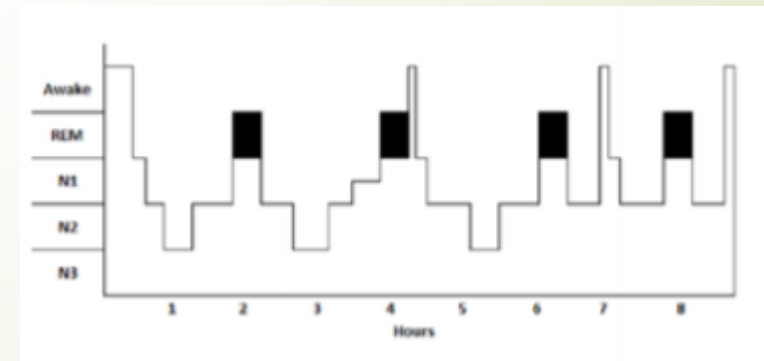
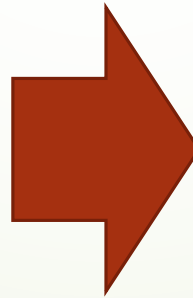
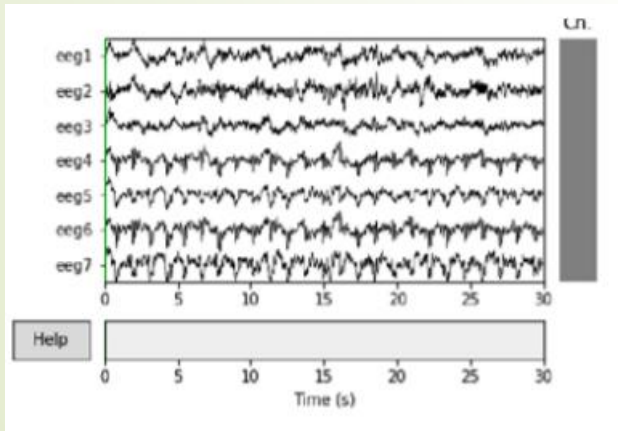
It's important but complex to classify sleep stage

- Sleep can be classified into different category : wake, light sleep, deep sleep, REM
- Monitoring sleep stage is beneficial for diagnosing sleep disorders.
- It is a long and difficult work to score it manually based on EEG signal reading

Our goal : to develop an algorithm of sleep staging able to differentiate between the different sleep stages.

Our dataset

- Dataset of 38k entries
- Each entry : 7 eeg signals of 30s sampled at 50hz
- Each entry is labeled with a sleep stage (1-4)
 - A dataset of about 5 GO



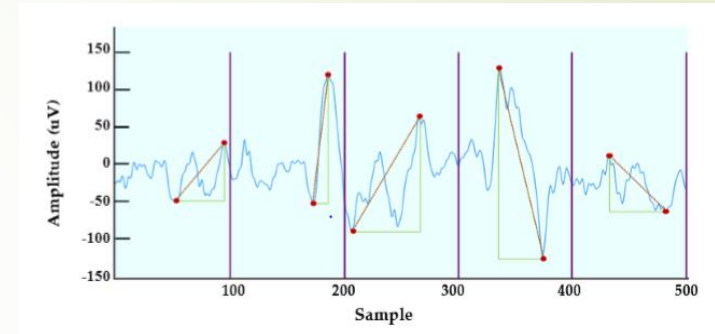


Our approach

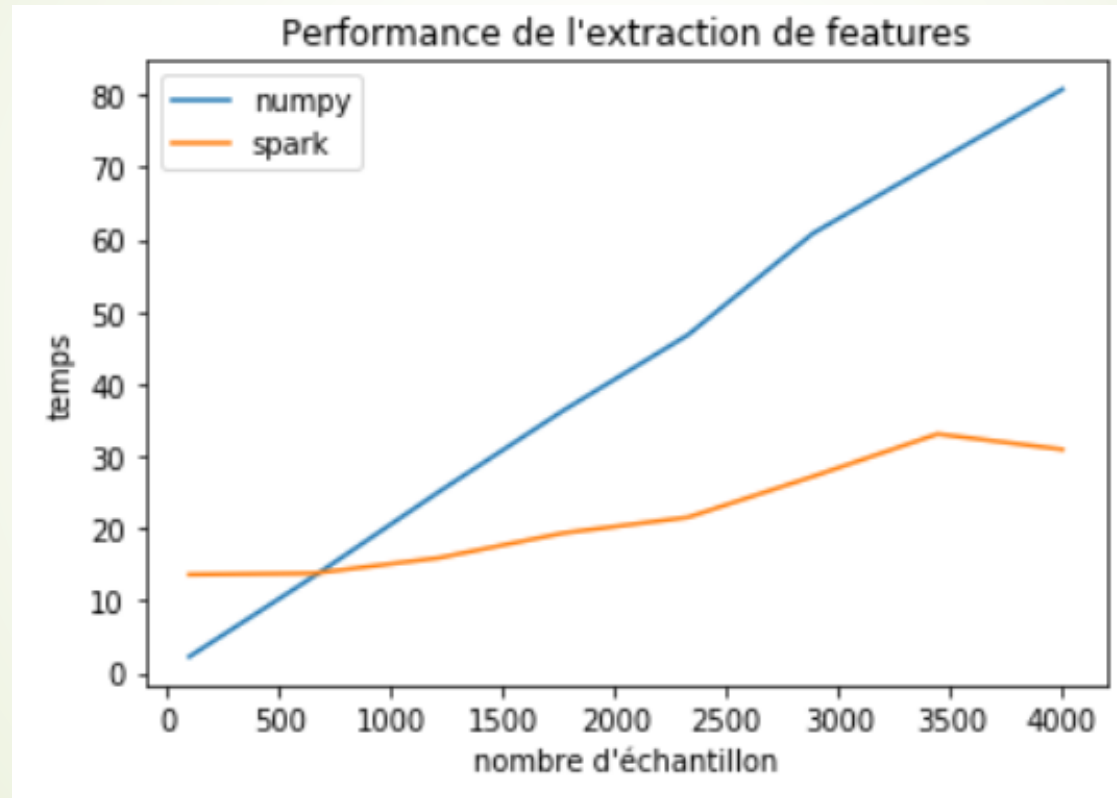
- 1) Extracting different features from the signals
 - In a traditional manner
 - Using a pyspark architecture
- 2) Using a classifying algorithm in order to predict f1 score

Our features

- Mean
- Standard deviation
- Min-Max distance
- Entropy



Pyspark vs traditionnal extraction





Classification and evaluation

- Model : A simple random tree forest
- F1 : 0,61

K-means Map-Reduce pseudo-code

- 1. Choose k centers: C_1, \dots, C_k
- 2. Apply the map method and associate each x_i to its closest center
- 3. Apply the reduce method until all the points corresponding to each class are merged, we then have a list of k tuples:
 $[(1, (y_1, c_1)), \dots, (k, (y_k, c_k))]$
- 4. Compute the new centers as follow $\tilde{C}_l = \frac{y_l}{c_l}$
- 5. Go back to step 2 until the algorithm converge

K-means Map-Reduce methods

Class Mapper

method map((C_1, \dots, C_k, x_i)):

$$l = \operatorname{argmin}_{j \in \llbracket 1, k \rrbracket} (\|C_j - x_i\|)$$

emit ($l, (x_i, 1)$)

Class Reducer

method reduce($(l, [(x_i, \text{card}_i), \dots, (x_j, \text{card}_j)])$):

emit ($l, [(y = \sum_{p=i}^j x_p, c = \sum_{p=i}^j \text{card}_p)]$)