Amine Hadj-Youcef



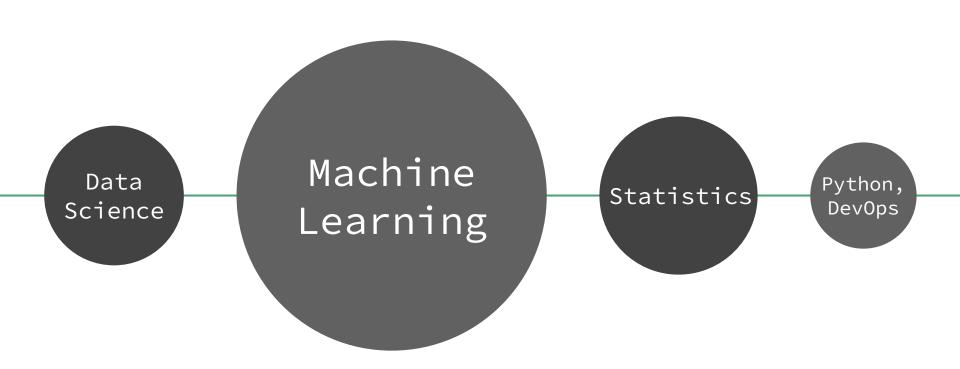
Senior Data Scientist

About me

I am passionate about solving problem to make our life better. I build solutions by processing data and developing computer algorithms. I have over 5 years of cross-sector experience in R&D of AI solution for startups and world's biggest brand - spatial, aerospatial, IoT, EdTech.

I am an avid runner, and I'm excited to partner with you!

Skills & expertise



Career highlights

Mentor Data Scientist

Openclassrooms, Paris September 2020 - Present

Data Scientist / Data Engineer

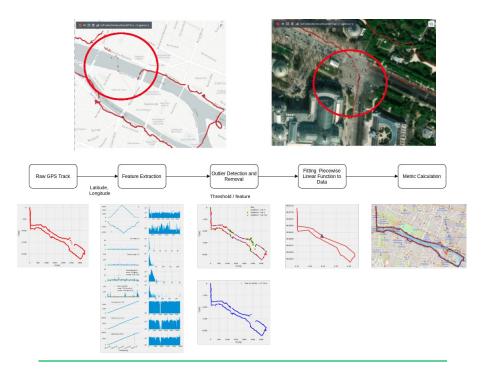
TAG Heuer, Paris

Marsh 2020 - September 2020

Data Scientist

Drone Volt, Villepinte
January 2019 - December 2019

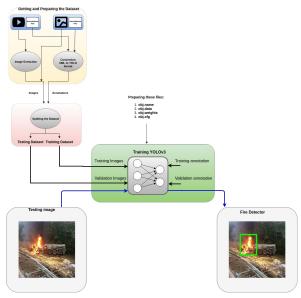
Portfolio



Anomaly Detection in GPS Data

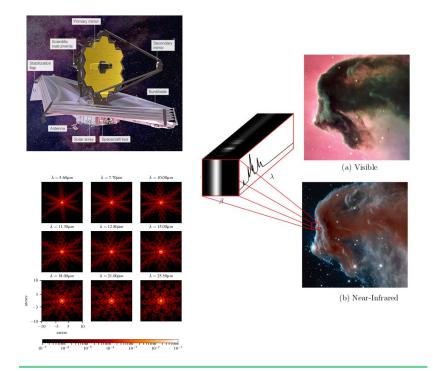
Data Processing Pipeline development for Time series data processing and anomaly detection using machine learning

Training YOLOv3 for Fire Detection



Fire Detection using Deep Learning Deployed on PENSAR

Deep learning model for the detection of fire in a video. The algorithm is embedded in the edge camera and runs in real time. Collecting and annotating data to Training and testing a deep learning model.



Spatio-spectral reconstruction from multispectral data. Application to the Mid-Infrared instrument of the JWST

In charge of innovating a data processing solution for the JWST / NASA space telescope (14 countries, budget> \$ 10 billion)

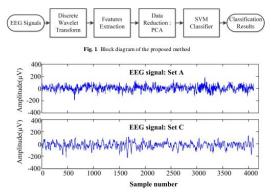
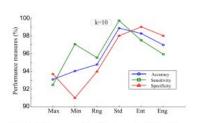


Fig. 2. Example of EEG signal from set A (Healthy) and set C (Epileptic).



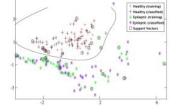


Fig. 8. Performance of classification using different features with 10-fold cross-validation

Fig. 7. Example of classification using SVM classifier and standard deviation feature (std) with 10-fold crossvalidation.

Detection of epileptics during seizure free periods

The goal is to anticipate the treatment of the epilepsy and protect the individual suffering from the pathology in case of epileptic seizures.

List of Publication

Journal

- Hadj-Youcef, M. A., Orieux, F., Abergel, A., & Fraysse, A. (2020). Fast Joint Multiband Reconstruction From Wideband Images Based on Low-Rank Approximation. *IEEE Transactions on Computational Imaging*, 6, 922–933.
- Boucaud, A., Bocchio, M., Abergel, A., Orieux, F., Dole, H., & Hadj-Youcef, M. A. (2016). *Convolution kernels for multi-wavelength imaging*. *October*, 7. https://doi.org/10.1051/0004-6361/201629080

International conference

- Hadj-Youcef, M. A., Bousbia-Salah, A., & Adnane, M. (2018). Feature selection applied to wavelet packet transform for an efficient EEG signal classification. *2018 International Conference on Electrical Sciences and Technologies in Maghreb (CISTEM*), 1–6.
- Hadj-youcef, A. (2014). Reconstruction des Images par des méthodes d'optimisation convexes et non lisses.
- Hadj-youcef, M. A., & Abergel, A. (2017). Restauration d'objets astrophysiques à partir de données multispectrales floues et une réponse instrument non stationnaire. *Gdr Isis*, 123(908), 2017.
- Orieux, F., Fraysse, A., & Abergel, A. (2018). Restoration from Multispectral Blurred Data with Non-Stationary Instrument Response. ii, 1–5.
- Hadj-Youcef, M. E. (2018). Spatio spectral reconstruction from low resolution multispectral data: application to the Mid-Infrared instrument of the James Webb Space Telescope. Université Paris-Saclay (ComUE).
- Hadj-Youcef, M. A., Orieux, F., Fraysse, A., & Abergel, A. (2018). Spatio-spectral multichannel reconstruction from few low-resolution multispectral data. *2018 26th European Signal Processing Conference (EUSIPCO)*, 1980–1984.
- Hadj-Youcef, M. A., Adnane, M., & Bousbia-Salah, A. (2013). Detection of epileptics during seizure free periods. 2013 8th International Workshop on Systems, Signal Processing and Their Applications, **WoSSPA** 2013, May, 209–213. https://doi.org/10.1109/WoSSPA.2013.6602363

Posters

Restauration d'objets astrophysiques à partir de données multispectrales floues et d'une réponse instrument non stationnaire

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Contexte

Caltraval traits is restauration d'objet astrontrusique à partir de données multispectrales acquises pa l'imageur moyen-infrarauge (MRI), à bord du prochair télescope spatial James Webb (JWST) [jwst.nasa.gov].



L'obje d'intérêt à l'entrée du système optique est un flux lumineux avant deux a varyon writtens a remove our système optique est un flux lumineux ayant deux disconsions spotiales et une dimension spottrale. Durant le processus d'observation, l'objet est modifié par la réponse de l'instrument, fournissant une sortie discréte 2D pour chaque fitre [1].

Problématique

Limitation de la résolution spatiale de données à cause de la diffration de la lumière [2] Variation spectrale importante de la réponse optique entre [5 - 18;im] Intégration spectrale de l'objet sur de larges bandes

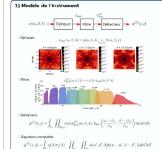
Mayora's échantillograge spertrale de données, seulement 9 filtre pour l'imaneur MIRI

3) Modèle directe multi-longueur d'onde En combinant les modèles de l'instrument et de l'objet nous un seul filtre $y^{(f)}(i, j) = \sum_{i=1}^{n_b} h_{int}^{(f,b)}(i, j) \times_{\alpha} x^{(b)}(i, j)$ $b_{int}^{(f,h)}(i,j) = \int_{a} q(\lambda)\tau_f(\lambda) \left(g_+^{(h+1)}(\lambda) + g_-^{(h)}(\lambda)\right)h(i,j,\lambda) d\lambda.$ $\hat{x} = \operatorname{argmin} J(x) = ||y - Hx||_2^2 + \frac{1}{2}$ - Solution du problème

 $\dot{\boldsymbol{x}} = (\boldsymbol{H}^t\boldsymbol{H} + \boldsymbol{C}^t\boldsymbol{C})^{-1}\boldsymbol{H}^t\boldsymbol{y}$

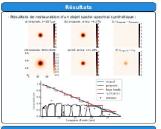
Reconstruction de l'abjet atrophysique spatio-spectral original en exploitant l'ensemble de dannées à différentes bandes spectrales de l'instrument

Méthodologie



2) Modèle de l'objet

La distribution spectral e de l'objet au pixel (
$$i,j$$
 est mod di sée par des fenctions linéaire par morceaux subst $\phi_{i,j}(\lambda) = \sum_{k=1}^{m_i} x_i^{(k)} y_j^{(k)}(\lambda) + x_i^{(k-1)} y_j^{(k)}(\lambda)$ $y_j^{(k)}(\lambda)$ est $y_j^{(k)}(\lambda)$ est une fonction contenant la variation spectral linéaire



Conclusion

Restauration du contenu spatio-spectral de l'obiet original à partir de données nultispectral acquises par l'imageur

Madélisation de la réponse instrumentale décrivant la physique de l'instrument. Elle prend en compte la variation en longueur d'onde de la PSF et l'intégration spectrale sur

Dévelopmement d'un modèle direct linéaire muiti Jonqueux d'ande en modélicant le spectre de l'objet par une fonction continue, choisie linéaire par morreaux régularisés per l'algorithme du gradient conjugué Augmentation significative de résolution spatiale et une meilleure reconstruction de la

distribution spectrale par rapport aux approches conventionnelles utilisant une PSF 2D

[1] M.A. Hadi-bucet: F. Orieux: A. Fraysse: A. Abergel: Restoration from Multispectra Blurred Data with Non-Stationary Instrument Response, EUSIPCO 2017
[2] J. W. Goodman, Introduction to Fourier optics, Raberts and Company Pub., 2005. [3] G. Aniano, B. Draine, K. Gordon, and K. Sandstrom, "Common-resolution convolution sernels for space-and ground-based telescopes," Publications of the Astronomical Societ of the Pocific, vol. 123, no. 988, p. 1218, 2011.

XXVIe Colloque GRETSI, Palais des Congrés, Juan-Les-pins, France, 5-8 Semptebre 2017

DETECTION OF EPILEPTICS DURING SEIZURE FREE PERIODS

M.A. Hadj-Youcef¹, M. Adnane¹, A. Bousbia-Salah²



INTRODUCTION

In this paper the problematic of epileptic detection is treated. The difference with conventional methods is the use of free seizure epileptic records. The raw EEG signal is decomposed using discrete wavelet transform (DWT). principal component analysis (PCA) is used to reduce the dimensionality of the Data. Several features are extracted and used as input of support vector machine (SVM) classifier. Results show satisfactory classification accuracy comparable or better than those reported in literature.

PROBLEMATIC

Detection of epileptic during seizure free periods. Classification of EEG signal into two classes: Healthy and Enilentic.

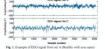
PROPOSED METHOD

Development of an automatic system for EEG signal classification Choose of an efficient method of EEG signal analysis and

classification.

DATA AND METHOD

. A publicly available EEG database [1] was used. A complete description of the database is available in [2]. Example given in Figure 1 shows two EEG signals one from set A (Healthy with eyes open) and one from set C (Epileptic during seizure free periods).



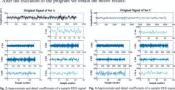
The two Sets of EEG signal are decomposed using Discrete Wavelet Transform (DWT), with level '6' and wavelet 'db2'.

REFERENCES

[1] EEG time series data, Department of Epileptology University of Bonn, http://www.meb.unibonn.de/epileptologie/science/ physik/eegdata.html (accessed: 21 December 2012). [2] R. G. Andrzejak, K. Lehnertz, F. Mormann, C. Rieke, P. David, C. E. Elger, Indications of nonlinear deterministic and finite-dimensional structures in time series of brain electrical activity: Dependence on recording region and brain state. Phys. Rev. E. 64 (6) (2001) 61907.

RESULTS

After the execution of the program we obtain the below results:



seament taken from a healthy patient with even open (set A).

Several feature are extracted : Maximum s-(Max), Minimum (Min), Range (Rng), Standard deviation (Std), Energy (Eng), Entropy of shannon (Ent).

These features are reduced using Principal Component Analysis (PCA) and used like an input to the support vector machine (SVM) classifier, the figure below show the classification using SVM classifier using standard deviation feature (Std) with Fig. 4.Classification of EEG data using Std with 10-fel cross-10-fold cross-validation.



segment taken from a epileptic patient during seizure free period

Method is performed using three parameter measure : Accuracy, Sensibility, Specificity. The figure below show the performance of classification using different feature



Best results (Table 1) are obtained when combining features standard deviation (Std) and entropy (Ent). This result can not be commented without deep investigation about feature selection and feature comparison which is our next research work

with 10-fold cross-validation:

cross-validation Performance K=2 K=5 K=10 ACC (%) 98,84 98,99 99 SE (%) 98,97 99 99 SP (%) 98,89 99 99

CONCLUSION

In this paper, method for EEG signal classification is described. The two classes' scheme was chosen such as class 1 is healthy EEG records and, class 2 is the epileptic EEG records with no seizures. The method uses the decomposition of the EEG signals using DWT. Then features are extracted. Moreover, PCA is used so vectors of features obtained in this operation are transformed into a new data vectors. The dimension of new data vectors is reduced in comparison to initial data. Finally, Training and Classification are applied to data from mean of SVM classifier. The results are very satisfactory which means that there is possibility to detect epilepsy at an early stage of the pathology using the developed method.

The 8th International Workshop on Systems, Signal Processing and their Applications (WOSSPA), 12-15 May, 2013, Mazafran, Algiers, Algeria

Gallery



Contact

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