

# Deep Learning For Time Series Classification

Ali Ismail-Fawaz, Maxime Devanne, Jonathan Weber, **Germain Forestier**

IRIMAS, Université de Haute-Alsace



*DALL-E (OpenAI) illustration of Inception Time*

January 2024

Tutorial E-EGC 2024 - Dijon

# **Outline**

---

**1. Introduction**

**2. Taxonomy of methods**

**3. Why Deep Learning for TSC ?**

**4. Applications**

# Contents

---

- 1. Introduction**
2. Taxonomy of methods
3. Why Deep Learning for TSC ?
4. Applications

# The dream team

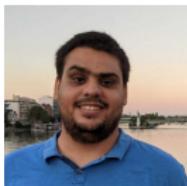
---



Maxime Devanne



Germain Forestier



Ali Ismail-Fawaz



Jonathan Weber

- **Goal of the tutorial:** discover the field of deep learning for time series classification
- **Outline:** a short presentation to introduce the field followed by a practical session on Google Colab
- Deep Learning for Time Series Classification is an active field of research
- slides: <https://germain-forestier.info/e-egc.pdf>

**Acknowledgements :** The providers of the UCR archive, Mésocentre for providing access to their cluster and the Agence National de Recherche for the Delegation Project.

# Open position in our team in 2024

---

Open position of Associate Professor (Maitre de conférences) in 2014 in our team!

**Research:** AI / time series (IRIMAS lab, MSD team)

**Teaching:** Computer Science Engineering School (ENSISA)



*Lac Blanc (massif des Vosges) - MSD team annual hike*

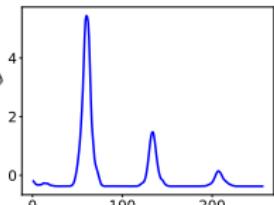
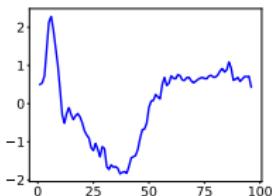
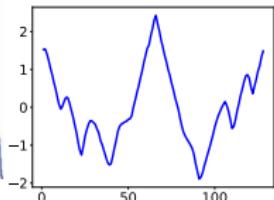
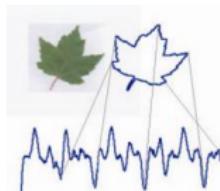
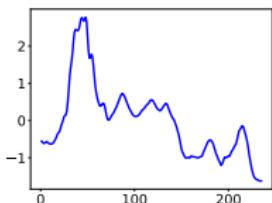
**"Mulhouse en tête de liste des villes où concilier travail et logement"** (29/11/23)

<https://www.leparisien.fr/economie/mulhouse-en-tete-de-liste-des-villes-où-concilier-travail-et-logement-il-y-a-une-energie-et-un-potentiel-enormes-29-11-2023-C5M5YY3J5ZCSLKN7GREIA6KQHM.php>

# Time series are ubiquitous

---

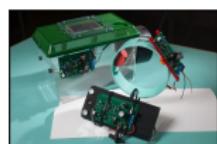
- Type of data present in numerous applicative domains
- Allow to study the evolution in time of a process, a behavior, etc.



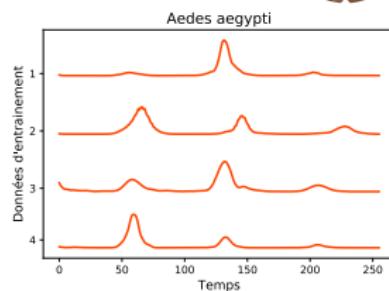
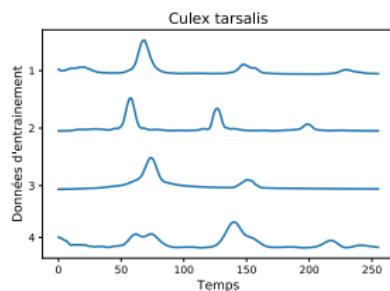
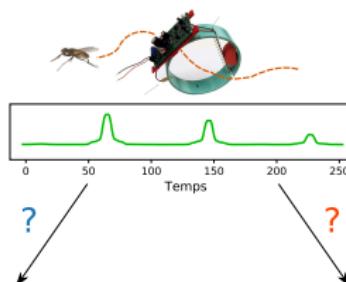
source : <http://timeseriesclassification.com/>

# Time series classification

- Time series are regrouped in classes (e.g. Culex / Aedes)
- The goal is to affect a class to new time series



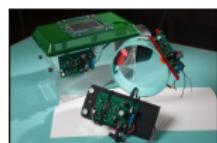
UNIVERSITY OF CALIFORNIA  
RIVERSIDE



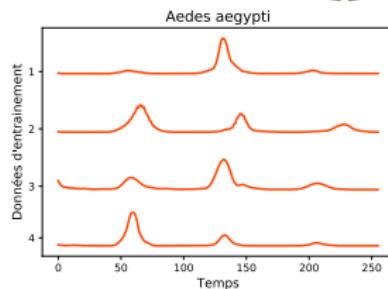
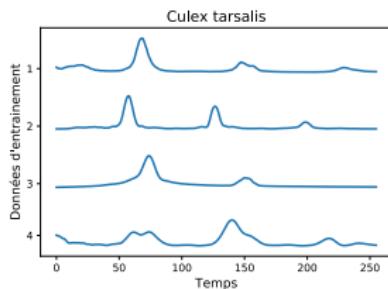
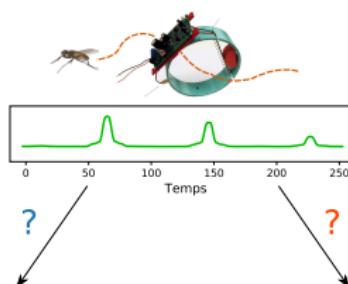
Petitjean, F., Forestier, G., Webb, G. I., Nicholson, A. E., Chen, Y., & Keogh, E. (2014). Dynamic time warping averaging of time series allows faster and more accurate classification. In **IEEE International Conference on Data Mining** (pp. 470-479)  
→ [ICDM 2023 10-year highest-impact paper Award](#)

# Time series classification

- Time series are regrouped in classes (e.g. Culex / Aedes)
- The goal is to affect a class to new time series



UNIVERSITY OF CALIFORNIA  
RIVERSIDE



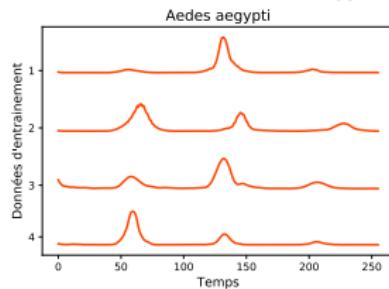
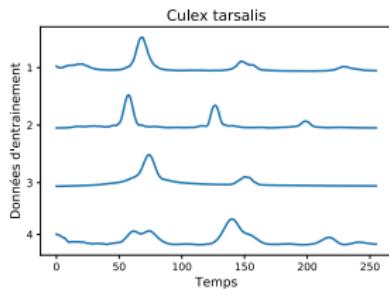
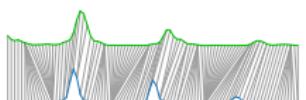
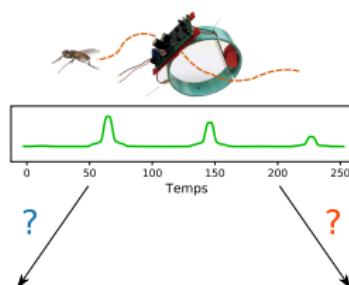
Petitjean, F., Forestier, G., Webb, G. I., Nicholson, A. E., Chen, Y., & Keogh, E. (2014). Dynamic time warping averaging of time series allows faster and more accurate classification. In *IEEE International Conference on Data Mining* (pp. 470-479)  
→ ICDM 2023 10-year highest-impact paper Award

# Time series classification

- Time series are regrouped in classes (e.g. Culex / Aedes)
- The goal is to affect a class to new time series



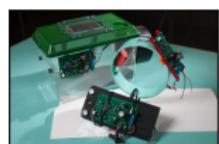
UNIVERSITY OF CALIFORNIA  
RIVERSIDE



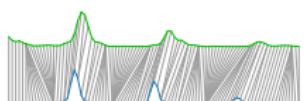
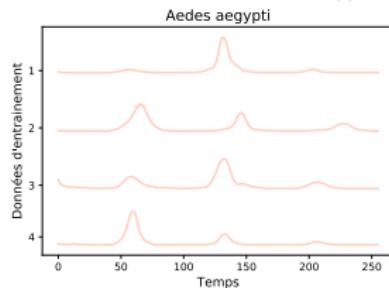
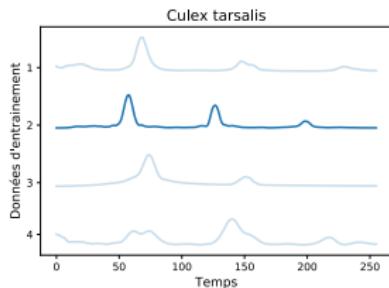
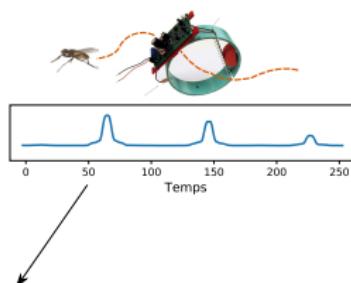
Petitjean, F., Forestier, G., Webb, G. I., Nicholson, A. E., Chen, Y., & Keogh, E. (2014). Dynamic time warping averaging of time series allows faster and more accurate classification. In **IEEE International Conference on Data Mining** (pp. 470-479)  
→ **ICDM 2023 10-year highest-impact paper Award**

# Time series classification

- Time series are regrouped in classes (e.g. Culex / Aedes)
- The goal is to affect a class to new time series



UNIVERSITY OF CALIFORNIA  
RIVERSIDE



Alignement avec  
Dynamic Time Warping (DTW)



Petitjean, F., Forestier, G., Webb, G. I., Nicholson, A. E., Chen, Y., & Keogh, E. (2014). Dynamic time warping averaging of time series allows faster and more accurate classification. In **IEEE International Conference on Data Mining** (pp. 470-479)  
→ [ICDM 2023 10-year highest-impact paper Award](#)

# Contents

---

1. Introduction

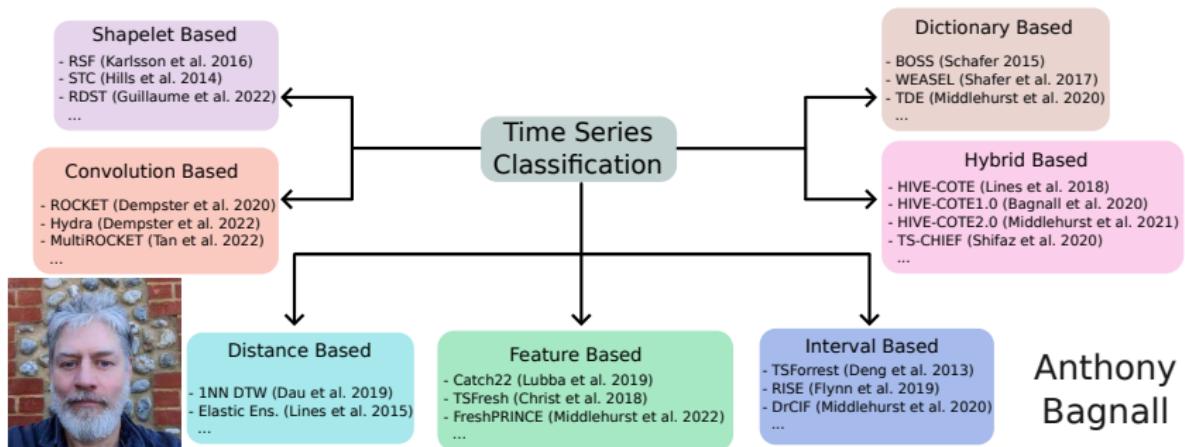
**2. Taxonomy of methods**

3. Why Deep Learning for TSC ?

4. Applications

# Taxonomy of methods

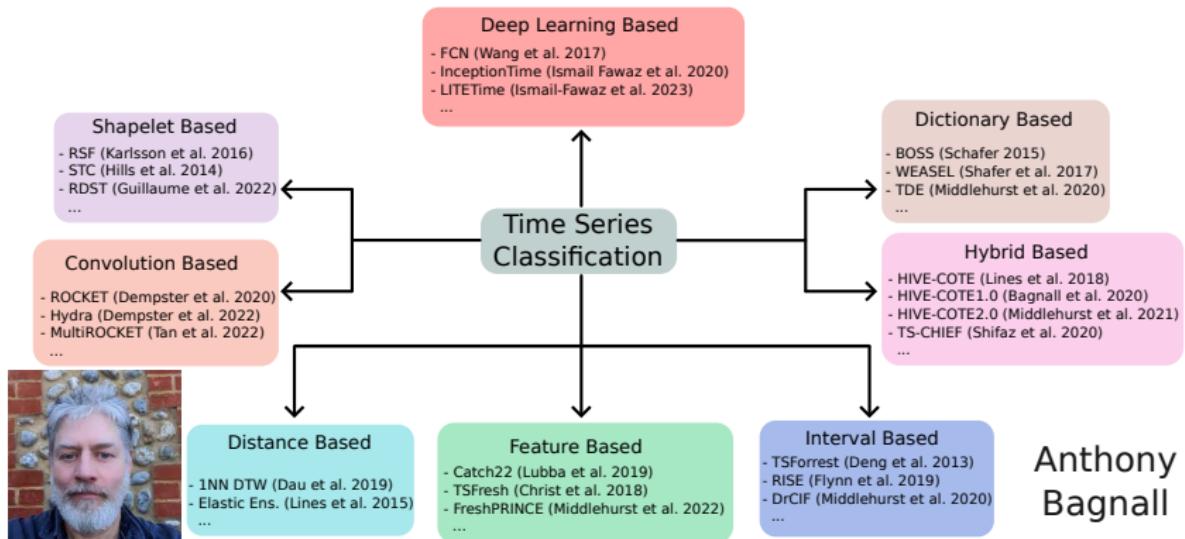
- Time Series Data can be used for various tasks including:
  - Clustering, representation learning, self-supervised learning, etc.
  - Extrinsic regression, forecasting, imputation, generation, etc.
  - Anomaly detection, similarity search, etc.
- In this talk, we focus on **classification**:



Middlehurst, M., Schäfer, P., & Bagnall, A. (2023). Bake off redux: a review and experimental evaluation of recent time series classification algorithms. *arXiv preprint arXiv:2304.13029*.

# Taxonomy of methods

- Time Series Data can be used for various tasks including:
  - Clustering, representation learning, self-supervised learning, etc.
  - Extrinsic regression, forecasting, imputation, generation, etc.
  - Anomaly detection, similarity search, etc.
- In this talk, we focus on **classification**:



Middlehurst, M., Schäfer, P., & Bagnall, A. (2023). Bake off redux: a review and experimental evaluation of recent time series classification algorithms. *arXiv preprint arXiv:2304.13029*.

# Contents

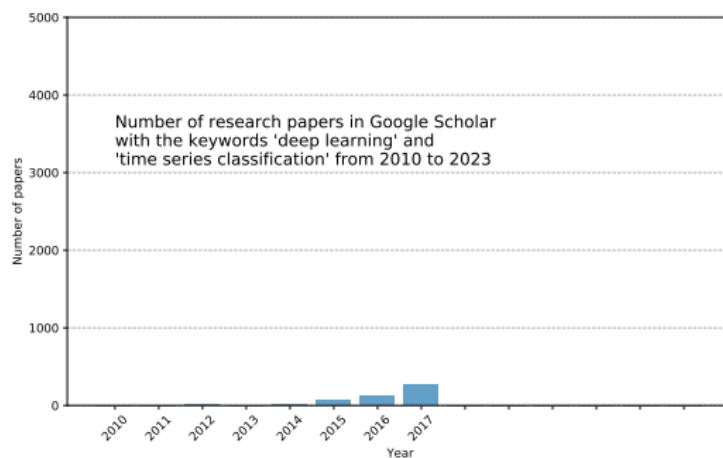
---

1. Introduction
2. Taxonomy of methods
- 3. Why Deep Learning for TSC ?**
4. Applications

# Why Deep Learning ?

---

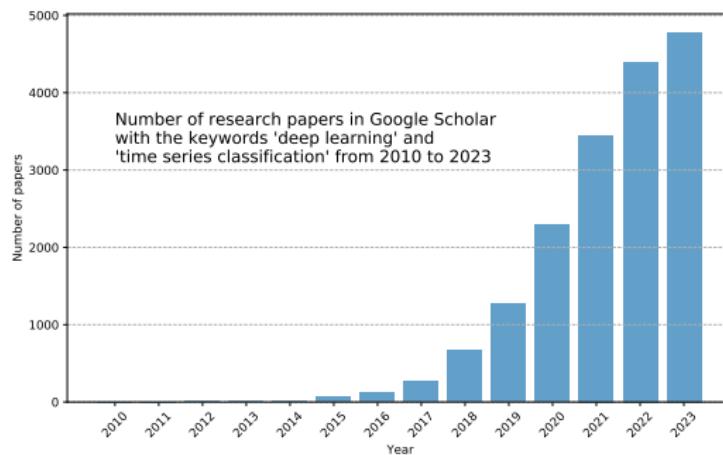
- In 2017, reviewers began to question the potential performance of deep learning for TSC while assessing papers on non-deep learning TSC methods.
- Deep learning has achieved great success in other data types, such as computer vision and natural language processing (NLP), so why not with time series?
- Deep Learning models learn to extract features and perform classification with one set of parameters (end-to-end)



# Why Deep Learning ?

---

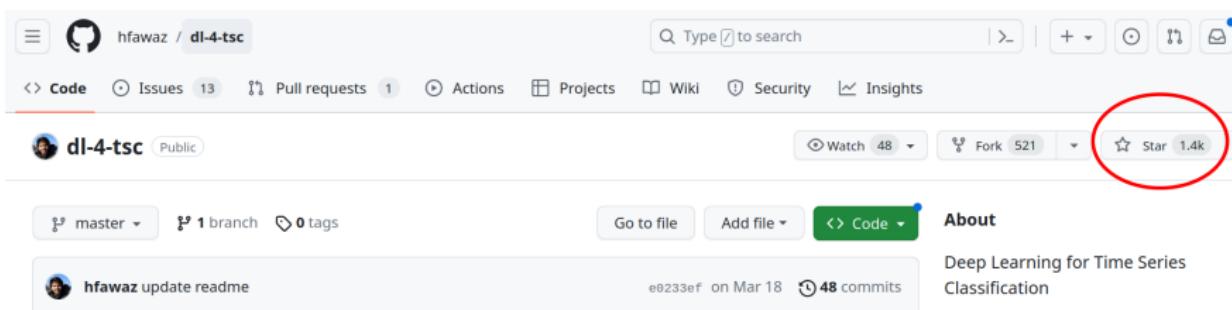
- In 2017, reviewers began to question the potential performance of deep learning for TSC while assessing papers on non-deep learning TSC methods.
- Deep learning has achieved great success in other data types, such as computer vision and natural language processing (NLP), so why not with time series?
- Deep Learning models learn to extract features and perform classification with one set of parameters (end-to-end)



# DL4TSC - 2019

In 2019, we presented a study of **Deep Learning for Time Series Classification** (cited more than 2.7K times (GoogleScholar)) [1].

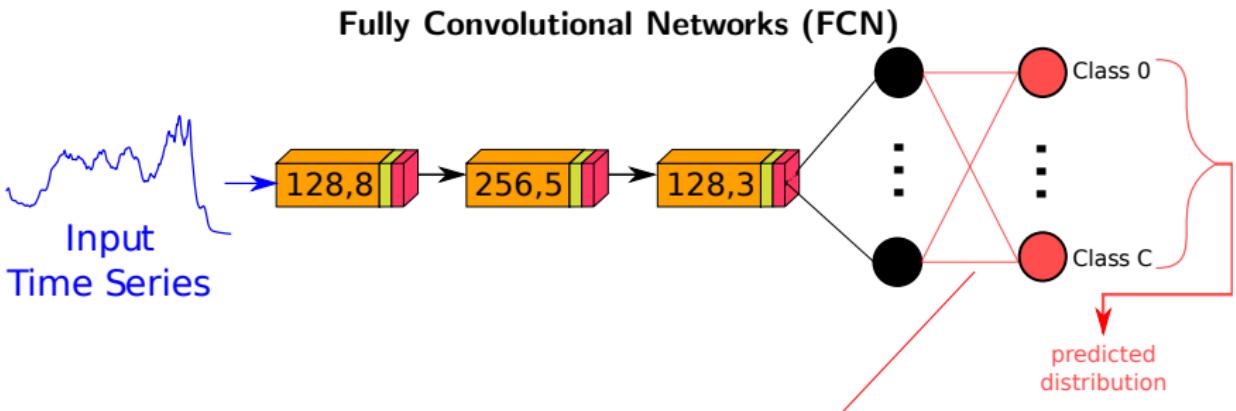
- We selected models with enough details (or available code) to reproduce the model's architecture
- We benchmarked all the models on the UCR archive [2]
- We published the code on Github for reproducibility and got very positive feedback (>1.4K stars)



- [1] Ismail Fawaz, H., Forestier, G., Weber, J., Idoumghar, L., & Muller, P. A. (2019). Deep learning for time series classification: a review. *Data mining and knowledge discovery*, 33(4), 917-963.
- [2] Dau, H. A., Bagnall, A., Kamgar, K., Yeh, C. C. M., Zhu, Y., Gharghabi, S. & Keogh, E. (2019). The UCR time series archive. *IEEE/CAA Journal of Automatica Sinica*

# DL4TSC - Some Architectures

 : 1D convolution layer with n filters of size k.  
 : batch normalization.  
 : activation  
— : fully Connected.  
— : 1D global average pooling



<https://msd-irimas.github.io/pages/dl4tsc/>

-  Wang, Z., Yan, W., & Oates, T. (2017, May). Time series classification from scratch with deep neural networks: A strong baseline. In 2017 International joint conference on neural networks (IJCNN) (pp. 1578-1585). IEEE.

# Why Deep Learning ?

## Convolutions on Images vs Time Series



The result of applying an edge detection convolution on an image



Convolution Result

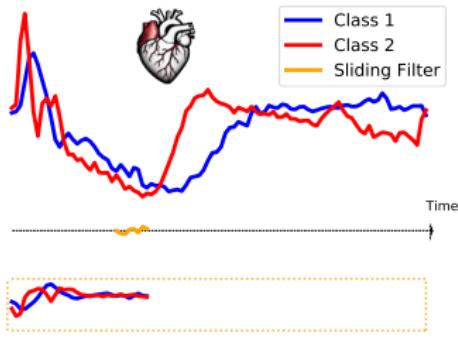
The result of applying a learned discriminative convolution on the ECG200 dataset

# Why Deep Learning ?

## Convolutions on Images vs Time Series



The result of applying an edge detection convolution on an image



Convolution Result

The result of applying a learned discriminative convolution on the ECG200 dataset

# Why Deep Learning ?

## Convolutions on Images vs Time Series



The result of applying an edge detection convolution on an image



The result of applying a learned discriminative convolution on the ECG200 dataset

# Why Deep Learning ?

## Convolutions on Images vs Time Series



The result of applying an edge detection convolution on an image



Convolution Result

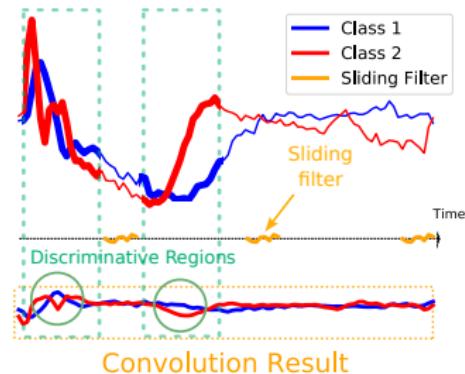
The result of applying a learned discriminative convolution on the ECG200 dataset

# Why Deep Learning ?

## Convolutions on Images vs Time Series



The result of applying an edge detection convolution on an image



The result of applying a learned discriminative convolution on the ECG200 dataset

# **Contents**

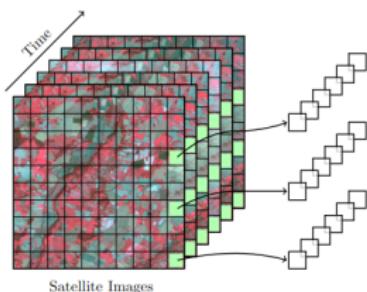
---

- 1. Introduction**
- 2. Taxonomy of methods**
- 3. Why Deep Learning for TSC ?**
- 4. Applications**

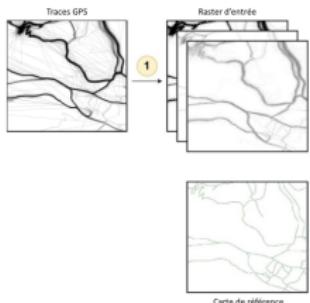
# Applications

---

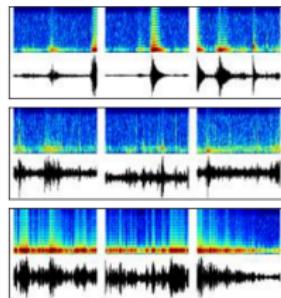
## Applications:



Series of Remote Sensing Images



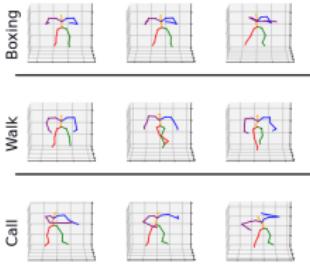
Generating Hiking Maps



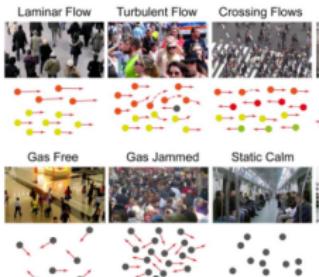
Seismology



Surgical Data Science



Body movements



Crowd movements

# Applications

---

## Crop dataset:

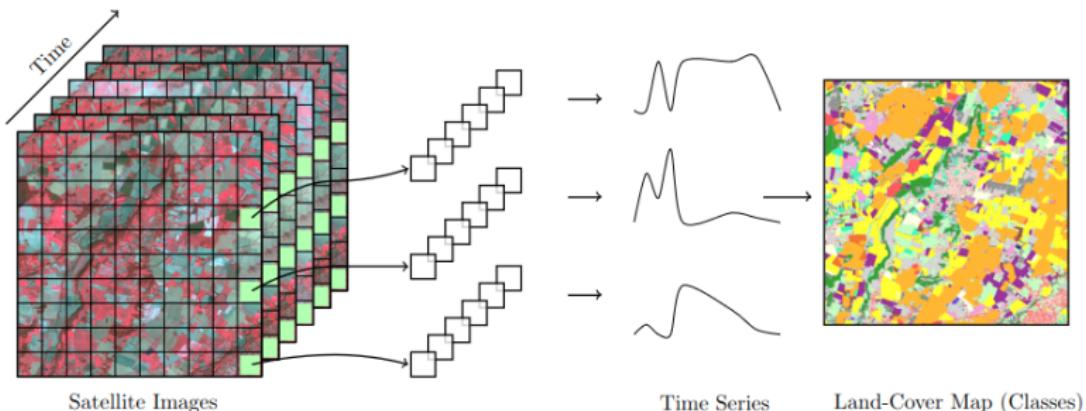


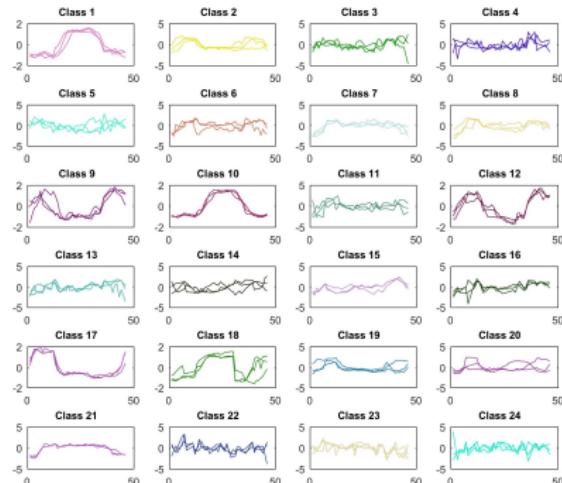
Figure 2: Production of a time series datasets from satellite image series.

Tan, C. W., Webb, G. I., & Petitjean, F. (2017, June). Indexing and classifying gigabytes of time series under time warping. In Proceedings of the 2017 SIAM international conference on data mining (pp. 282-290). Society for Industrial and Applied Mathematics.

# Applications

---

## Crop dataset:



Tan, C. W., Webb, G. I., & Petitjean, F. (2017, June). Indexing and classifying gigabytes of time series under time warping. In Proceedings of the 2017 SIAM international conference on data mining (pp. 282-290). Society for Industrial and Applied Mathematics.

# Applications

## Lets code!

