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Ancient geographic maps are valuable witnesses to our history, offering a unique perspective on the spatial dynamics of centuries past. Analyzing and comparing these maps with current digital data is a complex and time-consuming task but one that enables us to retrace the evolution of land occupation since the Middle Ages. Thanks to an innovative approach exploiting artificial intelligence (AI), particularly deep learning, new, easier-to-implement techniques are opening up to researchers in the field.

## Cassini map analysis using an AI approach

In a recent article [1], an IFPEN research team presented a method based on CNNs<sup>1</sup> to analyze the **Cassini map** (example in figure 1), created in the 18th century, and **compare its characteristics**

**with current land cover data.** The objective is to retrace the evolution of land occupation over time with a view **to quantifying the effects of human activity on the use of space** via the evolution of agricultural and pastoral activities, among others. The way land is occupied has a significant impact on its fragility; for example, land is far more susceptible to erosion when vegetation cover (particularly forests) is destroyed, thereby promoting rainwater run-off. The approach employed, known as semantic segmentation, made it possible to extract different categories of land use, such as forests, heaths, arboriculture and hydrological networks, from the Cassini map.



Figure 1 : Cassini Map of the Rhône Delta area

<sup>1</sup> Convolutional Neural Networks

## General principle of the methodology

The methodology employed uses a convolutional neural network<sup>2</sup> specifically adapted to **the semantic segmentation of images** (figure 2). In this approach, once a pre-defined class has been assigned to each image pixel, a new image is created in which the pixels indicate these different affiliation classes.

The convolutional neural network was trained via a data set comprising 30 small images extracted from the Cassini map, each one labeled with the different land use categories.

This semantic segmentation methodology links encoder and decoder sub-networks via dense embedded convolution blocks. The encoder plays a crucial role, capturing the characteristics of an image to varying degrees of abstraction. The decoder is then used to reconstruct the output image via a series of scaling operations and deconvolution layers.

<sup>2</sup> U-Net++

In more detail, the main steps in the methodology used are:

- 1) Convolution ( $3 \times 3$ ) with batch standardization and ReLU<sup>3</sup> activation extracts the characteristics.
- 2) Max pooling ( $2 \times 2$ ) reduces the input image to capture fine details and reduce the number of parameters.
- 3) A transposed convolution enlarges feature maps.
- 4) Concatenation merges the feature maps of the two sub-networks.
- 5) The first convolution on the input image uses 32 filters. The number of channels doubles after each convolution, while their size reduces by half after each max pooling.

<sup>3</sup> Rectified Linear Unit : function applied to the network's neurons and used to filter negative values

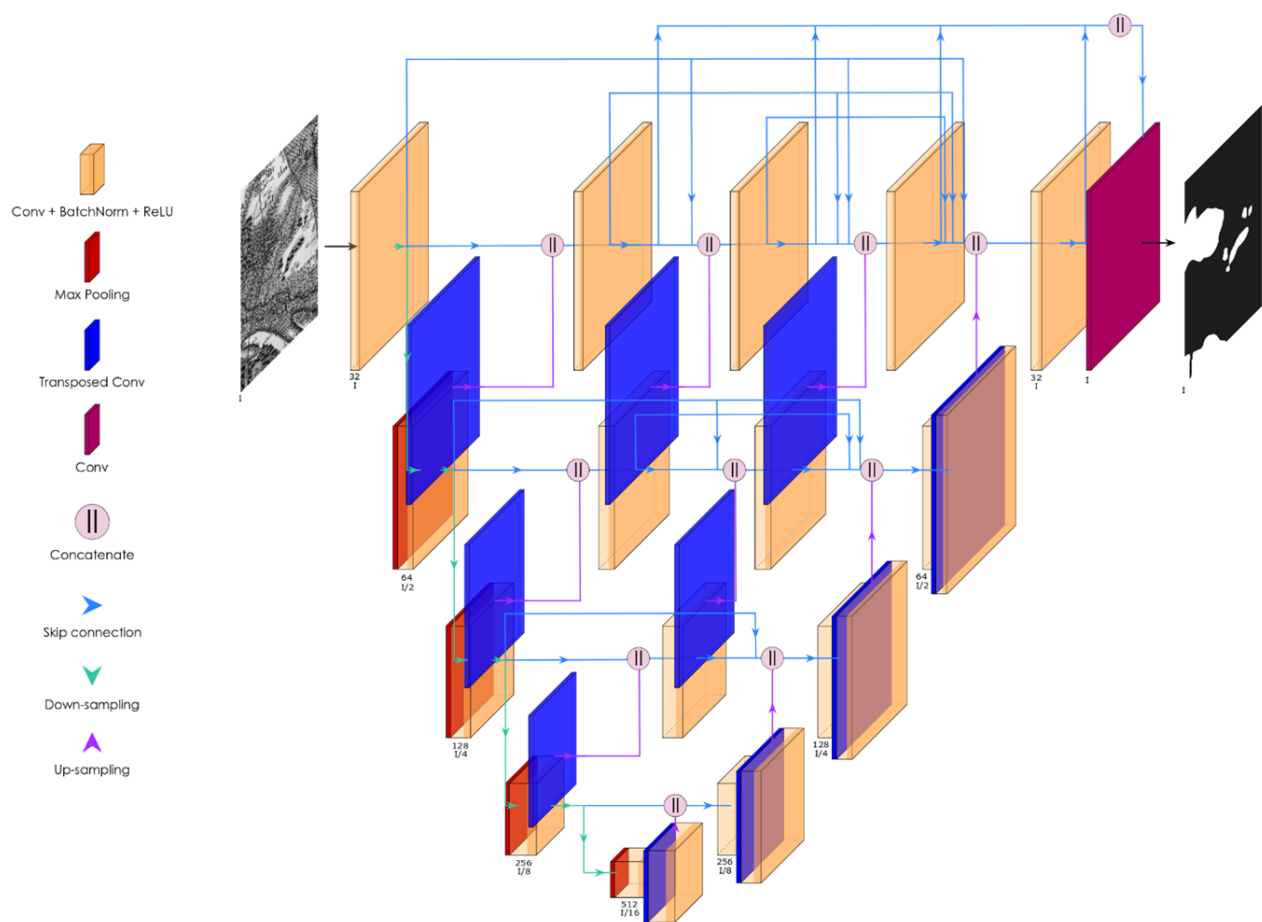


Figure 2 : Overview of U-Net++ architecture for our specific output

## Maps that speak volumes about landscape evolution

The approach adopted delivered good segmentation results with an average precision of 90% (between automatic and manual segmentations, the latter being considered to be our ground truth). **Segmentation results revealed significant differences between the Cassini map and current land cover data.** For example, forests currently cover a surface area more than 2.5 times greater than

on the Cassini map, while the hydrological network occupies a surface area more than five times greater on the old map. For the zones considered, these results provide clues about the major changes related to climate and human activities that have taken place over the past two centuries (figure 3).



# Deep learning ancient map segmentation to assess historical landscape changes

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## Location of the study area



## Legend :

### Land cover :

Forest

Heaths

Arboriculture

Hydrology

### Boundary :

Rhone basin boundary

Sub-basin boundary

Ellipsoid : WGS 84

Projection : World Geodetic System 1984

(EPSG : 4326)

0 10 Kilometers

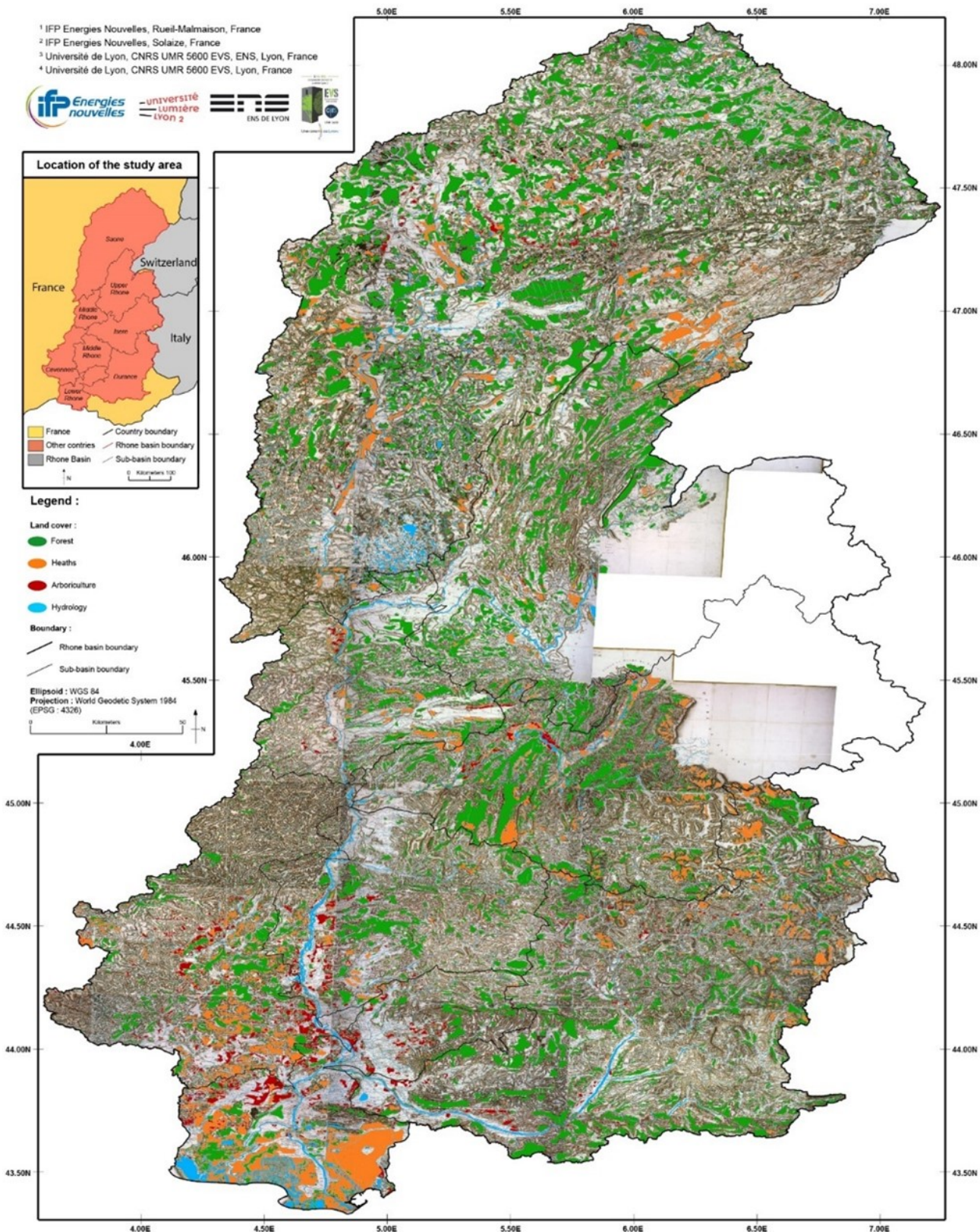


Figure 3 : Result of semantic segmentation on the Cassini map of the Rhône Basin [1] with different land occupation types (forest, heaths, arboriculture, hydrology) superimposed on the Cassini base map.

## The role of AI in the study of ancient maps

It turns out that AI has a crucial role to play in analyzing ancient maps for several reasons:

- **Time savings and efficiency:** automated semantic segmentation significantly reduces the time required to analyze ancient maps, thereby making it possible to explore large volumes of data rapidly.
- **Precision and reliability:** AI techniques deliver greater precision in the identification of land use classes, leading to reliable and robust analyses.
- **Enhanced historic understanding:** the combination of data extracted from ancient maps thanks to AI with historic and geographic knowledge opens up opportunities for a better understanding of the concomitant evolution of landscapes and societies over the course of time.

## Prospects for a better understanding of the past

This study underlines **the potential of AI in the analysis of ancient geographic maps** and paves the way for further such research in the field. Similar approaches could be conducted for other geographic regions and periods of history, thereby enabling a more in-depth exploration of our past. This represents a significant and promising methodological advance insofar as **it opens up new opportunities for researchers in the fields of history, geography and environmental sciences**. It was used in the PhD research conducted by T. Martinez [2] to quantify sediment flows resulting from erosion in the Rhône catchment area transiting towards the delta since the Middle Ages, with a view to calibrating numerical modeling parameters and anticipating the evolution of the Rhône delta as a function of various land occupation scenarios. Finally, this study illustrates the important role that digital technologies can play in the preservation and exploitation of our mapping heritage.

## References

[1] T. Martinez, A. Hammoumi, G. Ducret, M. Moreaud, R. Deschamps, H. Piegay, J.-F. Berger, (2023) Deep learning ancient map segmentation to assess historical landscape changes, Journal of Maps, 19:1. <https://doi.org/10.1080/17445647.2023.2225071>

[2] T. Martinez, Impacts des paramètres hydroclimatiques & anthropiques sur la sédimentation dans le système deltaïque du Rhône à la transition Holocène/Anthropocène (Impacts of hydroclimatic and human activity parameters on sedimentation in the Rhône delta system in the Holocene-Anthropocene transition), Université de Lyon 2, 2024

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14 May 2024

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