

# Visión por computador para las geociencias

a.k.a. Computer Vision for Geosciences (CV4GS)

2026-02-04

Sébastien Valade



1. Lecturers

2. Lecture goals

3. Lecture material

4. Lecture content

5. Points to discuss

## Main lecturer:

- **Sébastien Valade** ([valade@igeofisica.unam.mx](mailto:valade@igeofisica.unam.mx))
  - ⇒ Researcher at UNAM (Instituto de Geofísica, Departamento de Vulcanología)
  - ⇒ Volcanologist, research focus on Remote Sensing & Geophysics (<https://svalade.github.io>)

## Assistant:

- **Santiago Piñon Juárez** ([santiagopinonj@gmail.com](mailto:santiagopinonj@gmail.com))
  - ⇒ PhD student at UNAM (Instituto de Geofísica, Departamento de Vulcanología)
  - ⇒ Physicist, research topic on remote sensing of volcanic ash plumes from ground and space

## Special guest:

- **Sophie Giffard-Roisin**
  - ⇒ Researcher at IRD (France)
  - ⇒ Computer Scientist, research focus on Computer Vision and Deep Learning techniques for natural hazards (<https://sophiegif.github.io/>)
  - ⇒ will co-supervise the projects

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- acquire theoretical & practical knowledge to process digital images (*photographs, satellite imagery, microscope imagery, video, etc.*)
- progression from classical computer vision (CV) methods, to advanced artificial intelligence (AI) methods, using the Python programming language
- examples and applications inspired from geoscience problems

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- Programming language: Python
- Programming environment: Jupyter notebooks

Note: although the course is built around Python / Jupyter, you can use the programming language and IDE of your choice.

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1. Overview

2. Python/Jupyter crash course

3. Digital Image Processing

4. Machine Learning Methods

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## Overview (previous editions)

- Previous editions (2021, 2024) can be found here:
  - ⇒ <https://svalade.github.io/cv4gs/>
  - ⇒ 40% CV, 35% ML, 25% DL
- This edition intentions:
  - ⇒ *less CV, more DL !*
  - ⇒ *less lectures, more time for group projects !*

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## 1. Crash course to Python

- Introduction to the main features of Python  
⇒ *data types, operators, containers, control flow statements, functions/classes, modules/packages*
- Introduction to the programming tools used during the course  
⇒ *Jupyter notebooks*

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## 2. Digital Image Processing

- Digital image  
⇒ *what is a digital image and how to manipulate it*
- Image Filtering  
⇒ *spatial- and frequency-domain filtering, learn to code a spatial filter & use standard image processing libraries*
- Image Morphology & Segmentation  
⇒ *erosion, dilation, opening and closing operators; why segment an image and how to do so*
- Image Homography  
⇒ *image transformation matrices, image stitching & perspective correction*
- Features & Motion Estimation  
⇒ *image features, Digital Image Correlation & Optical Flow*

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### 3. Machine Learning Methods

- Introduction to the basic concepts of statistical learning
  - supervised / unsupervised learning
  - input / output domain
  - test / training / validation datasets
  - overfitting / underfitting
  - features and performance evaluation
- Machine Learning
  - Dimensionality reduction (PCA)
  - Regression algorithms (linear, polynomial, logistic, softmax)
  - Classification algorithms (kNN, SVM, RF)
  - ⇒ introduction to Python's Scikit-learn library
- Deep Learning
  - MLP (MultiLayer Perceptron) to CNN (Convolutional Neural Networks)
  - network architectures, transfer learning,
  - ⇒ introduction to Python's TensorFlow / Keras library

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- Lectures date/time: Fridays 09:00-13:00 ? Presencial o virtual?
- Key dates:
  - 2026-02-03 = start of semester
  - 2026-02-04 = first meeting (today)
  - 2026-05-29 = end of semester
  - 2026-06-12 = end of exam period
- Evaluation:
  - 1 partial exam
  - 1 group project
  - ⇒ Calificación total = promedio de los 2
  - ⇒ 80% de presencia para validar
- Experience with Python? With Jupyter notebooks? Personal Computer?
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