

# Homework Instructions

**Release Date:** 17.11.2025

**Submission Deadline:** 01.12.2025 (23:59)

**Late submissions will *not* be accepted under any circumstances.**

## Submission Format

- You may submit the homework **individually** or in **pairs (group of two)**.
- All solutions must be written and executed using a **Jupyter Notebook** (IPython) or **Google Colab**.
- Submit **one notebook file (.ipynb)** for each question containing:
  - The code
  - The output/results
  - Explanations and comments where needed
  - The images you used also in each question

# Code Quality Requirements

- Write **clean, readable, and well-structured code**.
- Use **meaningful variable and function names**.
- Add **clear comments** explaining your logic.
- Avoid unnecessary complexity.
- Make sure every code cell runs **without errors**.
- Ensure the notebook is organized into sections using markdown headings.

## Notebook Structure (Recommended)

### 1. Title and Student Information

- a. Full name(s)
- b. ID number(s)

### 2. Environment Setup

- a. Import all needed libraries

### 3. Task Breakdown

- a. Each question starts with a markdown explanation
- b. Code cells placed directly under the corresponding question

### 4. Results and Discussion

- a. Summaries, comments, or graphs if required

### 5. Final Conclusions (optional)

## Additional Notes

- Test your notebook before submission to ensure all outputs appear correctly.
- If using Google Colab, make sure to download the .ipynb file before submitting.
- When working in pairs, only **one student** uploads the assignment, but **both names and IDs** must appear clearly.
- Keep your answers concise, but demonstrate understanding.
- If the homework includes working with files or datasets, include clear instructions on how to run the notebook.
- If you have any questions, please contact the instructor **before** the submission deadline.

## Question 1 — Image Reading and Matrix Fundamentals

In this question, you will work with images as **matrices** and perform basic operations that are fundamental in image processing.

### Tasks

#### 1. Reading Images (Basics)

- a. Write Python code (using OpenCV or Matplotlib) to read an image (your choice).
- b. Display:
  - i. The original image
  - ii. The image shape (height, width, channels)
  - iii. The image data type
- c. Convert the image to **grayscale** and display it.

#### 2. Image as a Matrix

- a. Print a small section of the image matrix (e.g., a 5×5 block of pixel values).
- b. Explain what each number represents.
- c. Explain the difference between:
  - i. RGB image matrix
  - ii. Grayscale matrix
  - iii. Binary (thresholded) matrix

#### 3. Manual Pixel Manipulation

- a. Change some pixel values manually (e.g., set a region to white or black).
- b. Explain how direct manipulation of pixel values affects the displayed image.

#### 4. Creating a Matrix Manually

- a. Create a Python matrix (NumPy array) of size  $10 \times 10$  with specific patterns:
    - i. A diagonal line
    - ii. A filled square
    - iii. A gradient from 0 to 255
  - b. Display these matrices as images.
5. **Matrix Operations on Images** Perform the following operations using NumPy:
- a. Add a constant value to all pixels (brightness increase)
  - b. Multiply all pixels by a constant (contrast effect)
  - c. Clip values to valid range [0, 255]
  - d. Explain what happens mathematically.

## Question 2 — Histogram and Histogram Equalization

In this question, you will work with [image histograms](#), understand their meaning, and apply **histogram equalization** to enhance image contrast.

### Tasks

#### 1. Compute the Histogram of a Grayscale Image

- a. Read an image ( your choice ) and convert it to **grayscale**.
- b. Compute the histogram manually using NumPy:
  - i. Count how many pixels have value 0
  - ii. Count how many have value 1
  - iii. ...
  - iv. Count how many have value 255

- c. You may also use OpenCV's built-in function for comparison.

## **2. Plot the Histogram**

- a. Use matplotlib.pyplot to plot:
  - i. x-axis: pixel intensity (0–255)
  - ii. y-axis: number of pixels
- b. Label the axes clearly.
- c. Add a title such as "**Histogram of the Original Image**".

## **3. Interpretation Questions Answer:**

- a. What does it mean when the histogram is concentrated on the left side?
- b. What does it mean when the histogram is stretched across all values?
- c. What do high peaks represent?

## **4. Histogram Equalization**

- a. Apply histogram equalization to the grayscale image.
- b. You may implement it manually (CDF method) or use OpenCV: `cv2.equalizeHist()`.
- c. Display:
  - i. Original image
  - ii. Equalized image
  - iii. Original histogram
  - iv. Equalized histogram
- d. Explain how equalization changes the distribution of pixel intensities.

## **5. Comparison**

- a. Write a short comparison:
  - i. When does histogram equalization improve images?

- ii. When does it create noise or distort colors?
- iii. Why is it often used on medical or low-light images?

## Question 3 — Coin Detection in Mario Maps

You are given several **Mario map images** that contain multiple **coins** scattered in the scene. Your task is to detect these coins using **template matching** techniques.

### Tasks

#### 1. Smart Coin Cropping (Pre-processing)

- a. Select one clear coin from the map.
- b. Crop it **smartly and manually** (tight bounding box, centered, minimal background).
- c. You may apply:
  - i. Contrast enhancement
  - ii. Histogram equalization
  - iii. Smoothing or sharpening filters
- d. Explain why your cropping is effective and how it improves template matching.

#### 2. Template Matching Methods Implement and test the following three similarity measures:

- a. **SAD** — Sum of Absolute Differences
- b. **SSD** — Sum of Squared Differences
- c. **ZNCC** — Zero-mean Normalized Cross-Correlation

For each method, you must:

- d. Apply it on **each Mario map**.
- e. Detect all possible coin locations.
- f. Mark the detected coins visually on the image (e.g., bounding box / circle).
- g. Discuss **positives and negatives** of each method:
  - i. Sensitivity to lighting
  - ii. Sensitivity to contrast
  - iii. When it fails and why

### 3. Coin Detection Evaluation Tables (Per Method & Per Map)

You must create **three separate evaluation tables**, one for each method:

- a. **SAD**
- b. **SSD**
- c. **ZNCC**

Each table should contain **all maps** and show detection performance per map.

Example Format (Repeat for SAD, SSD, ZNCC)

Table: SAD — Coin Detection Results

Map Name	Total Coins (Ground Truth)	Detected	Missed	False Positives	Success Rate (%)
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- a. **Detected:** number of correctly identified coins.
- b. **Missed:** coins that exist but were not detected.
- c. **False Positives:** detections that are not actual coins.
- d. **Success Rate:**

$$\text{Success Rate} = \frac{\text{Detected Coins}}{\text{Total Coins (Ground Truth)}} \times 100\%$$

You must repeat this table for:

- d. **SSD — Coin Detection Results**
- e. **ZNCC — Coin Detection Results**