

**Biometric Security (IT499)**  
**Lab Assignment 2**  
**Biometric System Errors and Metrics**  
**Part B: Identification Framework**

Note: Use Python libraries such as numpy, pandas, matplotlib, and sklearn for data handling and plotting.

**1. Dataset Description**

LFW [dataset](#)

Dataset split (Assuming N subjects having  $t_i$  samples of  $i$ th subject)

**Enrollment dataset:** consider first  $t_i/2$  samples of the first  $N/2$  subjects for enrolment to constitute the gallery set.

**2. Preprocessing**

- A. Face detection and alignment
- B. Resizing the image as per the pretrained model input
- C. Normalize the image as per the model's requirement

**3. Feature extraction (classical)**

- A. Local Binary Pattern:
    - i. Use `skimage.feature import local_binary_pattern`
    - ii. Parameters: `radius = 3` `n_points = 8 * radius` `lbp_bins = 256`
    - iii. Generate 256 dimensional feature embeddings
  - B. Pretrained FaceNet model to extract 512 dimensional feature embeddings.
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**I. Closed set identification**

**A. Subject-dependent evaluation protocol:**

- i. Dataset
  - 1. Enrollment/Gallery set: same as discussed above
  - 2. Probe set: remaining  $t_i/2$  images of the **first  $N/2$  subjects (same as in gallery)**
  - 3. Feature extraction: **Fine-tune pretrained model** on the gallery set (show the loss curve as a function of epochs)
  - 4. Matching: Use cosine similarity to compute the comparison scores.
- ii. Based on the 1:N comparison scores:
  - 1. Plot the CMC curve showing rank-k identification rate in % (TPIR) for ranks  $k=1$  to 30.
  - 2. Is a threshold for comparison scores considered in this scenario? Why or why not?
- iii. Which feature (model) shows better identification performance?
- iv. Visualize the features by plotting the T-SNE plots.

**B. Subject-independent scenario- evaluation protocol**

- i. Dataset
  - 1. Enrollment/Gallery set: same as discussed above
  - 2. Probe set: remaining  $t_i/2$  images of the **first  $N/2$  subjects**
  - 3. Feature extraction: **Do not fine-tune pretrained model** on the gallery set. Extract features
  - 4. Matching: Use cosine similarity to compute the comparison scores between gallery and probe set embeddings.
- ii. Based on the 1:N comparison scores,

1. Plot the CMC curve showing rank-k identification rate in % (TPIR) for ranks  $k = 1$  to 30.
2. Is a threshold for comparison scores considered in this scenario? Why or why not?
- iii. Which feature (model) shows better identification performance?
- iv. Visualize the features by plotting the T-SNE plots.

## II. Open set identification

### A. Subject-dependent scenario: evaluation protocol:

- i. Dataset
  1. Enrollment/Gallery set: same as discussed above
  2. Probe set: remaining  $t_i/2$  images of the **total N subjects**
  3. Feature extraction: Fine-tune pretrained model on the gallery set (show the loss curve as a function of epochs)
  4. Matching: Use Euclidean distance to compute the comparison scores.
- ii. Based on the 1:N comparison scores,
  1. Is a threshold for comparison scores considered in this scenario? Why or why not?
  2. From the unenrolled entities, use a validation set of 30 users and select decision thresholds such that:
    - a. FPIR = 0.01
    - b. FPIR = 0.0001
    - c. FPIR = 0.000001
- iii. Since FPIR and FNIR are functions of the decision threshold  $d$ , calculate the False Positive Identification Rate (FPIR) and False Negative Identification Rate (FNIR) for threshold values determined above and plot separate ROC curve for each  $d$ . Report TPIR@FPIR of  $10^{-2}$ ,  $10^{-4}$  and  $10^{-6}$ .
- iv. Which feature (model) shows better identification performance?
- v. Visualize the features by plotting the T-SNE plots.

### B. Subject-independent scenario- evaluation protocol

- i. Dataset
  1. Enrollment/Gallery set: same as discussed above
  2. Probe set: remaining  $t_i/2$  images of the **total N subjects**
  3. Feature extraction: **Do not Fine-tune pretrained model** on the gallery set. Extract features.
  4. Matching: Use Euclidean distance to compute the comparison scores between gallery and probe set embeddings.
- ii. Based on the 1:N comparison scores,
  1. Is a threshold for comparison scores considered in this scenario? Why or why not?
  2. From the unenrolled entities, use a validation set of 30 users and select decision thresholds such that:
    - a. FPIR = 0.01
    - b. FPIR = 0.0001

- c.  $\text{FPIR} = 0.000001$
- iii. Since FPIR and FNIR are functions of the decision threshold  $d$ , calculate the False Positive Identification Rate (FPIR) and False Negative Identification Rate (FNIR) for threshold values determined above and plot separate ROC curve for each  $d$ . Report  $\text{TPIR} @ \text{FPIR}$  of  $10^{-2}$ ,  $10^{-4}$  and  $10^{-6}$ .
- iv. Which feature (model) shows better identification performance?
- v. Visualize the features by plotting the T-SNE plots.

## Submission Requirements:

- Submit your Jupyter/Colab notebook with the complete code and outputs.
- **Include a written report summarizing your observations and conclusions.**