

#### Faculty of Engineering – Cairo University Credit Hours System – Senior Level Spring 2024



# CMPS450 – Pattern Recognition and Artificial Neural Networks

## **Project Report**

## **Submitted by:**

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#### **Due Date:**

15 May 2024

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## **Problem Description**

## **Objective:**

Develop a system to classify Arabic paragraphs into four font categories based on input images.

#### **Challenges:**

Characterizing unique font features; handling variations in writing styles, spacing, and noise.

| Font Code | Font Name            |
|-----------|----------------------|
| 0         | Scheherazade New     |
| 1         | Marhey               |
| 2         | Lemonada             |
| 3         | IBM Plex Sans Arabic |

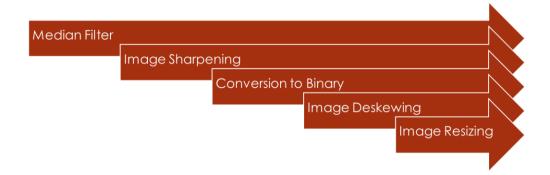
## **Project Pipeline**

Data Loading & Preprocessing Feature Selection Model Development Tuning Prediction

## **Data Loading & Splitting**

- Load the dataset into memory.
- Split the dataset into training and validation sets to evaluate model performance.
- The validation set is used to tune model hyperparameters to find the best model state.

## **Preprocessing Module**



#### **Steps:**

- 1. Median blur filter: to remove salt and pepper noise.
- 2. Filter2D: to sharpen the image.
- 3. cv2 Threshold: to convert the image to binary.
- 4. Image deskewing:
  - We want to detect any skew present in the image, which means if the image is slanted or tilted.
  - We try different angles  $(0^{\circ}, 45^{\circ}, 90^{\circ}, \text{ etc.})$  to check for skew.
  - For each angle tested, we compute a score based on how much the histogram of the image changes.
  - A higher score means more skew corrections are needed.
  - We choose the best angle that gives us the highest score, indicating the most significant skew.
  - With the best angle identified, we rotate the image in the opposite direction to counteract the skew.
  - The rotation fills any remaining empty spaces with a white color to maintain the image's rectangular shape.
- 5. Image resizing.

#### **Feature Extraction and Selection Modules**



#### **Steps:**

• Histogram of Oriented Gradients (HOG):

Extract features from images to describe their shapes and textures.

- Scale-Invariant Feature Transform (SIFT):
- ✓ Detect and describe key points in images to help recognize objects or scenes.
- ✓ Pad SIFT descriptors to make sure all SIFT feature sets are of the same length for consistency in analysis.
  - Standardization:

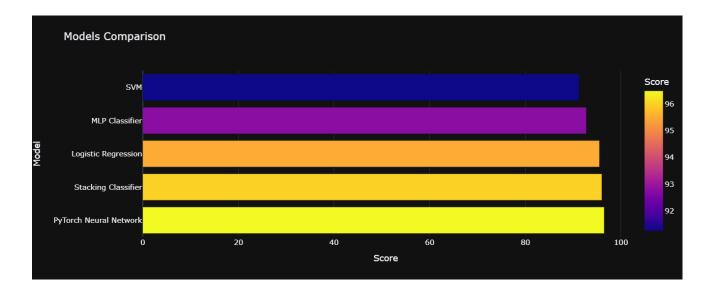
Adjusts the scale of features to ensure fair comparison.

• Principal Component Analysis (PCA):

Transform the data into a lower-dimensional space while preserving the most important information.

## **Performance Analysis**

| Model                  | Accuracy |
|------------------------|----------|
| PyTorch Neural Network | 96.5%    |
| Stacking Classifier    | 96%      |
| Logistic Regression    | 95.5%    |
| MLP Classifier         | 92.75%   |
| SVM                    | 91.25%   |



## **Model Architecture**

- Input layer processes feature vectors.
- 2 hidden layers learn complex patterns with ReLU activation.
- Output layer generates class probabilities with softmax activation.

**Accuracy obtained:** 96.5%

#### **Unsuccessful Trials**

#### A. Edge Direction Matrix (EDM):

#### **Features from EDM1 & EDM2:**

- Edges Direction
- Homogeneity
- Pixel Regularity
- Edges Regularity

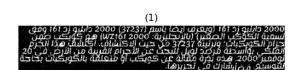
**Accuracy obtained: 85%** 

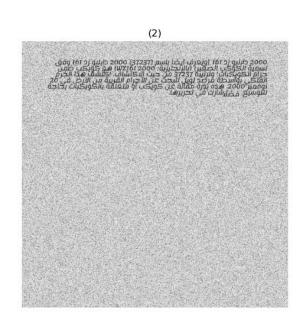


|      | 135° | 90°  | 45°  | _       |
|------|------|------|------|---------|
|      | 4    | 2    | 2    | ]       |
| 180° | 4    | 12   | 4    | 0°      |
|      | 2    | 2    | 4    | ]       |
|      | 225° | 270° | 315° | $EDM_1$ |

|      | 135° | 90°  | 45°  |         |
|------|------|------|------|---------|
|      | 2    | 0    | 1    | ]       |
| 180° | 3    | 12   | 4    | 0°      |
|      | 1    | 0    | 1    | ]       |
| ,    | 225° | 270° | 315° | $EDM_2$ |

## **B.** Segmentation technique (inspired by Variance Threshold by scikit-learn)





## **Workload Distribution**

| Team member | Tasks                                 |
|-------------|---------------------------------------|
| Ahmed Emad  | Preprocessing, Model Development, API |
| Hla Hany    | EDM, Model Development, Deployment    |
| Nada Tarek  | HOG, SIFT, PCA                        |
| Ziad Ahmed  | Preprocessing, Model Development, API |