

# Project Write-up

## Facial key point detection and recognition.

March 2025

### CS 419/619 Computer Vision

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# Objectives

The goal of this project is to predict facial key points on images using computer vision techniques. Each key point is represented by an  $(x, y)$  coordinate in pixel space. The model should accurately detect 15 predefined key points corresponding to facial features, namely `left_eye_center`, `right_eye_center`, `left_eye_inner_corner`, `left_eye_outer_corner`, `right_eye_inner_corner`, `right_eye_outer_corner`, `left_eyebrow_inner_end`, `left_eyebrow_outer_end`, `right_eyebrow_inner_end`, `right_eyebrow_outer_end`, `nose_tip`, `mouth_left_corner`, `mouth_right_corner`, `mouth_center_top_lip`, `mouth_center_bottom_lip`.

## Scope

- Preprocessing image data and handling missing key points.
- Building a model to accurately predict key points.
- Evaluating model performance and optimizing accuracy.
- Applying the trained model to test images for key point detection.

## Dataset Description

The dataset, obtained from Kaggle, consists of grayscale facial images with a resolution of  $96 \times 96$  pixels. The image data is stored in the last column of the dataset as a row-wise sequence of pixel intensity values ranging from 0 to 255. The dataset is divided into two parts:

- **Training Set:** Contains 7049 images, each annotated with  $(x, y)$  coordinates for 15 facial key points along with the corresponding pixel data.
- **Test Set:** Consists of 1783 images without labeled key points. Each row includes an `ImageId` and the pixel data in a row-ordered format.

## Expected Outcomes

- A trained model capable of detecting facial key points with high accuracy.
- Visualization of detected key points on test images.
- A submission file with predicted key points formatted as required.

## Performance Metrics

- **Accuracy**: can be defined as a performance metric by setting a threshold for acceptable error in key point predictions. One approach is **Key Point Accuracy** (The percentage of key points predicted within a certain error threshold (e.g., 5 pixels) from the actual location.)

$$Accuracy = \frac{\text{Number of key points with error} \leq \delta}{\text{Total number of key points}} \times 100$$

where:

- $\delta$  is the predefined threshold (e.g., 5 pixels),
  - A key point is considered correct if  $|y_i - \hat{y}_i| \leq \delta$  for both x and y coordinates.
- **Mean Absolute Error (MAE)**: Evaluates the average absolute differences between predictions and ground truth.

$$MAE = \frac{1}{N} \sum_{i=1}^N |y_i - \hat{y}_i|$$

- **Root Mean Squared Error (RMSE)**: Measures the square root of the average squared differences between predicted and actual key point locations. Compared to the Mean Absolute Error, RMSE punishes large errors.

$$RMSE = \sqrt{\frac{1}{N} \sum_{i=1}^N (y_i - \hat{y}_i)^2}$$

where:

- $N$  is the total number of key points,
- $y_i$  is the actual key point location,
- $\hat{y}_i$  is the predicted key point location.