# Topic 2 - Face Detection

Group 57

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# Group Members

- NGUYEN DUY THAI 175906
- PHAM HOANG DUY 220607
- NGUYEN HOANG VINH QUANG 219130
- PHAN VAN TAN 219200
- TRUONG MINH NGHIA 164626

## Overview

- Introduction to Face Detection
- Importance of Face Detection
- Applications of Face Detection
- Face Detection Techniques
- Project Goals

## Introduction to Face Detection

 Definition: Technology to identify and locate human faces in images and videos.

#### • Importance:

- Foundation for facial recognition, emotion detection, and security systems.
- Crucial in various applications like surveillance, user authentication, and personalized marketing.

#### How it Works:

- Uses algorithms and machine learning techniques.
- Detects facial features and distinguishes them from other objects.

#### Advancements:

- Deep learning has enhanced accuracy and efficiency.
- Modern systems are more robust and reliable.

## **Proposed Solution**

- Model Selection: Use YOLOv10-L, a state-of-the-art object detection model known for its speed and accuracy.
- Data Collection: Gather a dataset of diverse images containing faces, ensuring a balanced representation of different facial features.
- **Data Annotation**: Label the images with bounding boxes around faces using tools like LabelImg or CVAT.
- Environment Setup: Clone the YOLOv10-L repository and install dependencies in Google Colab for free GPU access.
- Model Configuration: Define the model architecture and configuration using a custom YAML file tailored for face detection.
- **Training**: Train the YOLOv10-L model on the annotated dataset, optimizing for accuracy and performance.
- Evaluation: Assess the model's performance using metrics such as mAP (mean Average Precision) and adjust parameters as needed.

## Architecture Model Evaluation

Model	Pros	Cons
VGG16	- Simplicity - Strong Feature Extraction	- Computationally Intensive - Not Specialized for Detectio
ResNet50	<ul><li>Residual Connections</li><li>High Accuracy</li><li>Scalability</li></ul>	<ul><li>Complexity</li><li>Resource Intensive</li></ul>
YOLO	<ul><li>Real-Time Performance</li><li>High Accuracy</li><li>Unified Architecture</li></ul>	- Complexity - Resource Intensive

### Architecture

#### YOLOv10 Detailed Structure:

- Backbone:
  - Uses CSPDarknet architecture for feature extraction.
  - Includes multiple convolutional layers and residual blocks.
- Neck:
  - PANet structure for path aggregation.
  - Enhances feature pyramid for better detection at various scales.
- Head:
  - Outputs bounding box coordinates, objectness scores, and class probabilities.
  - Utilizes anchor boxes for improved localization accuracy.
- Advantages:
  - Superior performance on small and large objects.
  - Optimized for both accuracy and speed.

## **Data Processing**

### • Preprocessing:

- Normalization: Adjust image pixel values to a common scale to improve model performance.
- Augmentation: Apply techniques such as rotation, flipping, and scaling to increase dataset diversity.
- Annotation: Label images with bounding boxes around faces using tools like LabelImg or CVAT.
- **Dataset Preparation**: Ensure balanced representation of various facial features and expressions.

## **Data Processing**

- Post-Processing:
  - Non-Max Suppression (NMS): Filter out overlapping bounding boxes to retain the best predictions.
  - Bounding Box Refinement: Adjust predicted boxes to better align with detected faces.
  - Confidence Thresholding: Discard predictions below a certain confidence level to reduce false positives.
  - Evaluation Metrics: Use metrics like mAP (mean Average Precision) to assess model accuracy.

#### **Dataset**

#### Source:

- We used the Face Detection Dataset from Kaggle.
- This dataset is specifically curated for training and testing face detection models.

#### • Dataset Composition:

- Training Set: 26,300 images with annotated face locations.
- Validation Set: 6,500 images with similar annotations.

#### Annotations:

 Each image comes with corresponding labels indicating face positions using bounding boxes.

### **Dataset**

#### • Preparation:

- Downloaded and extracted the dataset using a simple helper script.
- Ensured the removal of duplicate images and corresponding labels.

#### Directory Structure:

Organized as follows:

### YAML Configuration:

- Defined paths for training and validation data in a data.yaml file.
- Included class names and counts for model reference.

## Evaluation Metrics

## Average Precision (AP):

- Measures precision and recall at various thresholds.
- Calculates the weighted mean of precisions achieved at each threshold.
- Provides a comprehensive view of model performance across different confidence levels.

#### AP@0.5:

- Measures precision and recall with a fixed Intersection over Union (IoU) threshold of 0.5.
- Indicates how well the model distinguishes true positives from false positives.
- Important for evaluating object detection models in real-world applications.

## **Evaluation Metrics**

## Mean Average Precision (mAP):

- Combines AP scores over multiple IoU thresholds (e.g., 0.5 to 0.95).
- Averages AP across all classes in the dataset.
- Offers a comprehensive metric for overall model performance comparison.

#### • Importance:

- These metrics provide insights into the trade-offs between precision and recall.
- Essential for fine-tuning the model to achieve optimal detection accuracy.
- Used to benchmark performance against other models and datasets.

# Experimental Results

## **Evaluation Metrics**

- VGG16: AP, AP@0.5 results
- ResNet50: AP, AP@0.5 results
- YOLO: AP, AP@0.5 results

## Benchmark

• Comparison with state-of-the-art models in face detection tasks

# Ablation Study (Optional)

 Analyzing the impact of different architectural changes on model performance

## Examples of Test Results

 Generated images and screenshots demonstrating detection performance on test data

## Conclusion

- Face detection is a vital technology
- Wide range of applications
- Project aims to contribute to this field

# Questions?

• Open for any questions or discussions

## References

- Dataset: Face Detection Dataset
- Information: Train set 26,300 images, Test set 6,500 images
- Evaluation Metric: AP, AP@0.5