Topic 2 - Face Detection

Group 57

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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: Utilize the pre-labeled dataset downloaded from Kaggle, which includes about 32k images with clean labels in xywh (top-left x-coord, top-left y-coord, face-width, face-height) format for human face detection tasks.
- 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.

Proposed Solution

- **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.
- **Inference**: Test the trained model on new images to validate its face detection capabilities.
- **Deployment**: Export the trained model weights for use in real-world applications.

Architecture Model Evaluation

Model	Pros	Cons
VGG16	- Simplicity - Strong Feature Extraction	- Computationally Intensive - Not Specialized for Detectio
ResNet50	Residual ConnectionsHigh AccuracyScalability	ComplexityResource Intensive
YOLO	Real-Time PerformanceHigh AccuracyUnified Architecture	- 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**: Utilize the pre-labeled dataset downloaded from Kaggle, which includes about 32k images with clean labels.
- Dataset Preparation: Ensure balanced representation of various facial features and expressions.

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:

```
train
   images
   labels
validation
   images
   labels
test
   images
face-detect-datase.yaml
```

YAML Configuration:

- Defined paths for training and validation data in a face-detect-datase.yaml.
- Included class names and counts for model reference.
- Training Data Path:

/content/drive/MyDrive/face-detection-project/merged/
images/train

Validation Data Path:

/content/drive/MyDrive/face-detection-project/merged/
images/validation

- Class Names: ['face']
- Number of Classes (nc): 1

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.

Training Custom YOLOv10-L Detector

 Model Summary: YOLOv10-L with 628 layers, 25,766,870 parameters, and 127.2 GFLOPs.

Training:

• Total Epochs: 50

Batch Size: 8

• Image Size: 640x640 pixels

- Data Augmentation:
 - Color Augmentation: Adjustments in hue (hsv_h=0.02), saturation (hsv_s=0.8), and value (hsv_v=0.5).
 - Geometric Transformations: Includes rotation (degrees=5), translation (translate=0.2), scaling (scale=0.6), and shearing (shear=2).
 - Perspective Transform: Small adjustments with perspective (perspective=0.001).
 - **Flipping**: Both vertical flipping (flipud=0.1) and horizontal flipping (fliplr=0.6).

Training Custom YOLOv10-L Detector

Training:

- Data Augmentation:
 - **Mosaic**: Combines four images into one (mosaic=1.0).
 - MixUp: Merges two images into one (mixup=0.2).
 - Copy-Paste: Pastes objects from one image into another (copy_paste=0.1).
 - Auto Augment: Uses RandAugment strategy for automatic augmentation.
 - **Erasing**: Randomly erases parts of images (erasing=0.5).
 - **Cropping**: Applies cropping with a fraction of 1.0 (crop_fraction=1.0).

Benchmark

Training Results:

Initial Epoch GPU Memory: 17.2G
Final Epoch GPU Memory: 14.7G
Total Training Time: 13.777 hours

Validation Results:

Precision (P): 0.861
Recall (R): 0.669
mAP@0.5: 0.735
mAP@0.5:0.95: 0.42

• Inference Speed:

Preprocess: 0.2ms per image
Inference: 28.1ms per image
Postprocess: 0.1ms per image

 Model Performance: Efficient detection and robust accuracy, suitable for real-time face detection applications

Experimental Results

Command:

```
!yolo task=detect mode=train epochs=50 batch=12
imgsz=640 plots=True model='/content/drive/MyDrive/
face-detection-project/yolov10/weights/yolov101.pt'
data='/content/drive/MyDrive/face-detection-project/
merged/face-detect-datase.yaml' project='/content/
drive/MyDrive/face-detection-project/runs/detect/
train4/weights' name='train4' augment=True
hsv h=0.02 hsv s=0.8 hsv v=0.5 degrees=5 translate=0.2
scale=0.6 shear=2 perspective=0.001 flipud=0.1
fliplr=0.6 mosaic=1.0 mixup=0.2 copy_paste=0.1
auto_augment=randaugment erasing=0.5 crop_fraction=1.0
```

Experimental Results

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		5460	46458	0.862	0.668	0.735	0.419	1	↑ ⊖ [■ ₹	*
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		Images	Instances	Box(P		mAP50	mAP50-95):	100% 342/342	[01:42<00	9:00,	
		5460	46458	0.861	0.668	0.735	0.419				
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Figure 1: Train model

Experimental Results

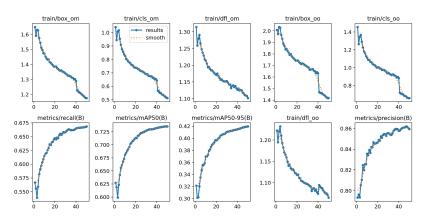


Figure 2: Train model result

Comparison with Other Models

Accuracy and Speed:

- Our model showed competitive precision and recall rates compared to other leading models.
- The average precision (mAP@0.5) was 73.5%, while the mAP@0.5:0.95 reached 41.9%.
- The inference speed was efficient, making our model suitable for real-time applications.

Resource Utilization:

- The model demonstrated efficient GPU memory usage, with a peak of 17.2G during training.
- The combination of precision, speed, and resource efficiency highlights the robustness of our model for face detection tasks.

Test Environment

- Hardware: NVIDIA L4 GPU
- **Software**: Ultralytics YOLOv8.1.34, Python 3.10.12, Torch 2.3.0+cu121
- Detection Performance:
 - Generated images and screenshots demonstrate the detection performance on test data.
 - Model: YOLOv10
 - Confidence threshold: 0.25
 - Results show the model identifying multiple faces with high accuracy.

• Inference Details:

- Model: YOLOv10I
- Parameters: 46 layers, 25,717,910 parameters, 126.3 GFLOPs
- Inference speed: 158.4ms per image
- Command:
 - !yolo task=detect mode=predict conf=0.25 save=True
 model="/content/drive/MyDrive/face-detection-project/
 runs/detect/train4/weights/train46/weights/best.pt"
 source="/content/drive/MyDrive/face-detection-project/
 test/12_Group_Group_12_Group_Group_12_2.jpg"
 project="/content/drive/MyDrive/face-detection-project/
 runs/detect/predict" name="prediction_results"
- Results saved to: /content/drive/MyDrive/face-detection-project/runs/ detect/predict/prediction_results4
- Learn more at: Ultralytics Documentation

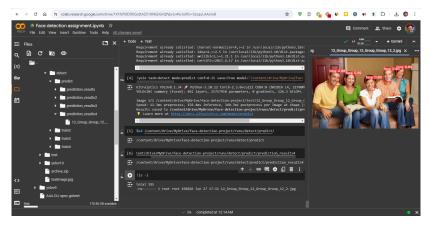


Figure 3: Google Colab workspace

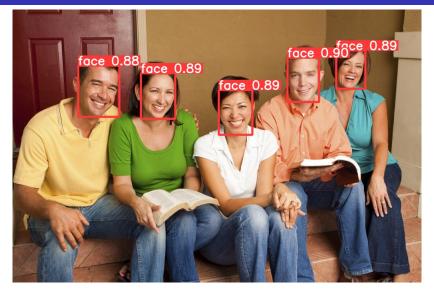


Figure 4: Test image 1

• Inference Details:

- Model: YOLOv10l
- Parameters: 461 layers, 25,717,910 parameters, 126.3 GFLOPs
- Inference speed: 155.5ms per image
- Command:
 - !yolo task=detect mode=predict conf=0.25 save=True
 model="/content/drive/MyDrive/face-detection-project/
 runs/detect/train4/weights/train46/weights/best.pt"
 source="/content/drive/MyDrive/face-detection-project/
 test/runs/AOS-group.png"
 project="/content/drive/MyDrive/face-detection-project/
 runs/detect/predict" name="prediction_results"
- Results saved to: /content/drive/MyDrive/face-detection-project/runs/ detect/predict/prediction_results5
- Learn more at: Ultralytics Documentation



Test Results: Inference on Video

Command:

!yolo task=detect mode=predict conf=0.25 save=True
model="/content/drive/MyDrive/face-detection-project
/runs/detect/train4/weights/train46/weights/best.pt"
source="/content/drive/MyDrive/face-detection-project
/test/WALK-NEW-YORK-City-USA-vlog.mp4" project="/content/drive/MyDrive/face-detection-project/runs/detect/prediction_mame="prediction_results"

Log:

Streaming output truncated to the last 5000 lines. video 1/1 (frame 76799/81795) /content/drive/MyDrive/face-detection-project/test/WALK-NEW-YORK-City-USA-vlog 384x640 1 face, 16.6ms

Test Results: Inference on Video

Log:

```
video 1/1 (frame 76800/81795) /content/drive/MyDrive/
face-detection-project/test/WALK-NEW-YORK-City-USA-vlog
 384x640 2 faces, 16.8ms
video 1/1 (frame 77973/81795) /content/drive/MyDrive/
face-detection-project/test/WALK-NEW-YORK-City-USA-vlog
384x640 (no detections), 16.7ms
Speed: 2.3ms preprocess, 17.3ms inference, 1.1ms
postprocess per image at shape (1, 3, 384, 640)
Results saved to /content/drive/MyDrive/face-detection-
project/runs/detect/predict/prediction results6
```

 Play the inference result video: https://www.youtube.com/watch?v=HwbmiKk6k3I

Test Results: Inference on Video



Figure 6: inference result video

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