

Mini project 3

# PEDESTRIAN OBJECT DETECTION

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# AGENDA

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# INTRODUCTION

## Real-Time Pedestrian Detection System

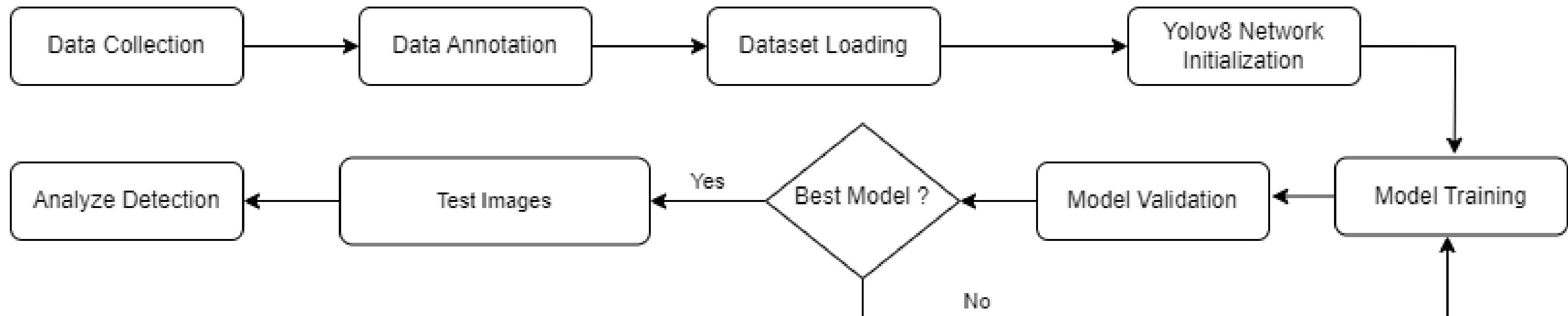
**we aims to develop a real-time pedestrian detection system using YOLO for object detection, enhancing traffic management and pedestrian safety through accurate and timely detection.it will be designed to detect and classify pedestrians in real-time, providing valuable insights for traffic management and enforcement**

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# PROPOSED FRAMEWORK

The framework consist of eight steps:



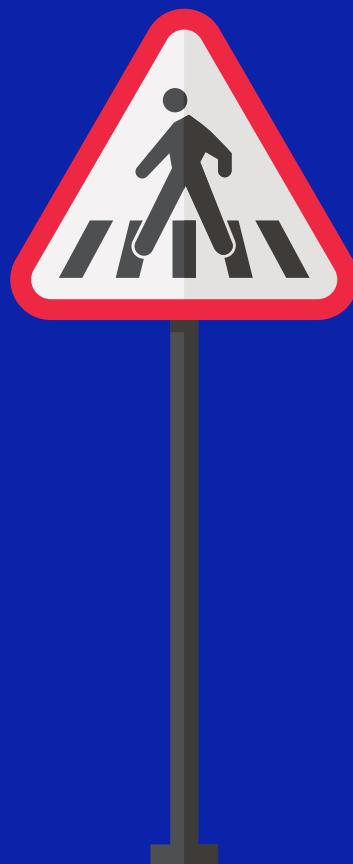
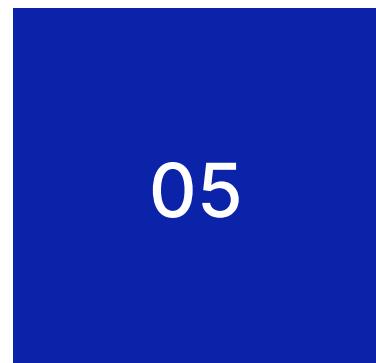
# DATA DESCRIPTION

**Name of dataset :** pedestrian-data-500

**Dataset Source :** Roboflow

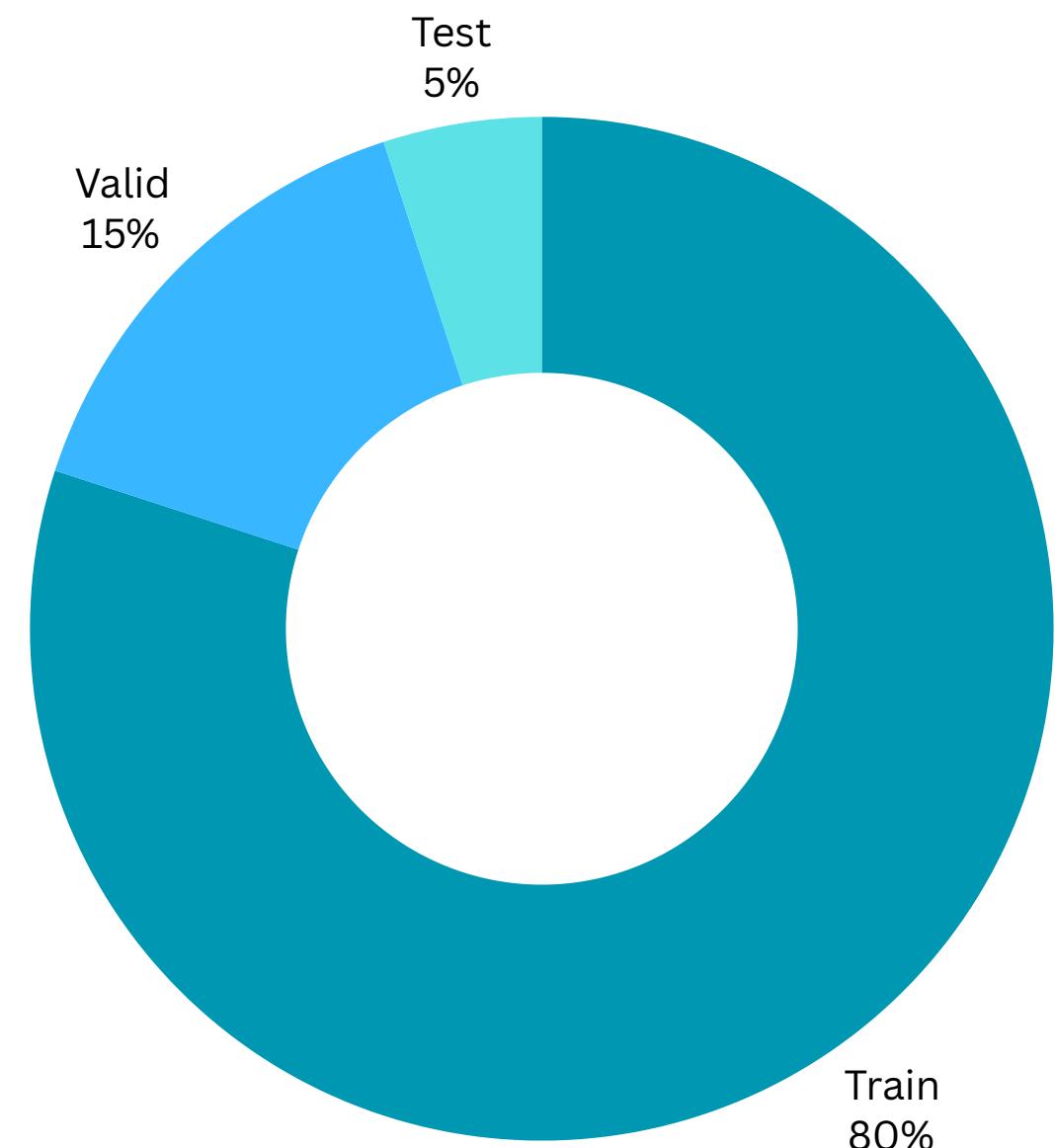
**Classes :** Pedestrian

**Framework :** Yolo object detection



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# SPLITTING THE DATASET

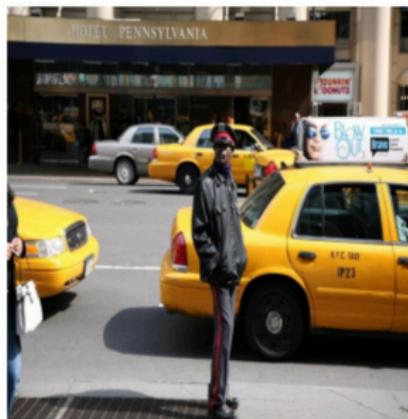


- Training set size: 400, Valid set size : 75 ,  
Test set size : 25

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# DATASET SAMPLE

Training Images



Validation Images



Testing Images





# DATA PRE-PROCESSING

The pre-processing consist of :

- 1- Resize images: 640 \* 640 pixels**
- 2- Convert from BGR to RGB**
- 3- Data Splitting: Train, Validation , Test Sets**
- 4- Data Augmentation: flipud, mixup, shear**

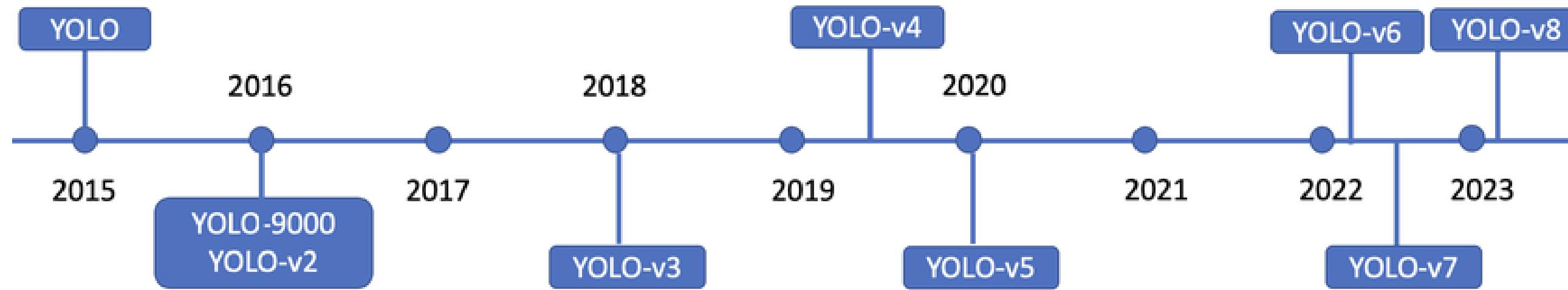
# IMPLEMENTED MODEL

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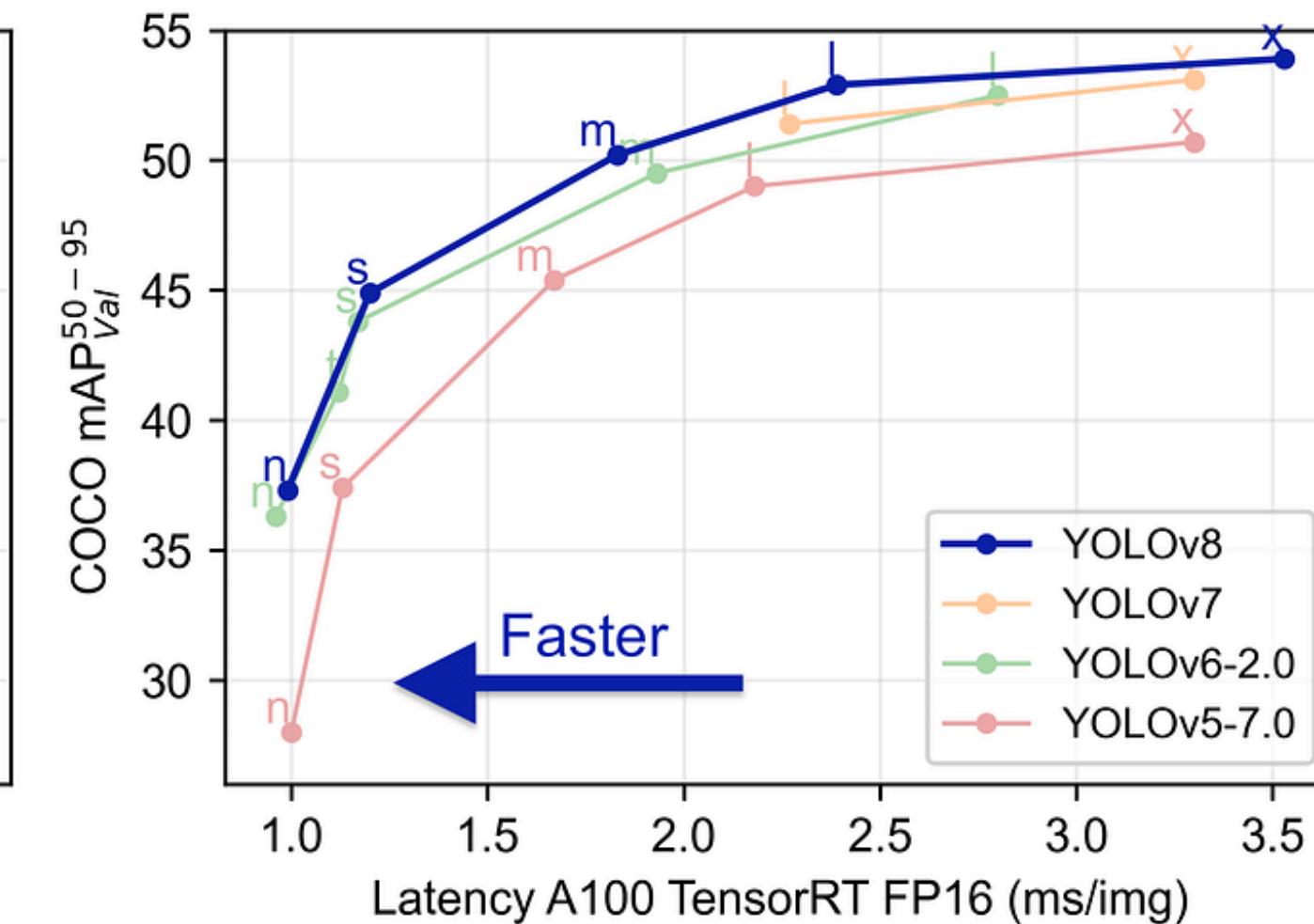
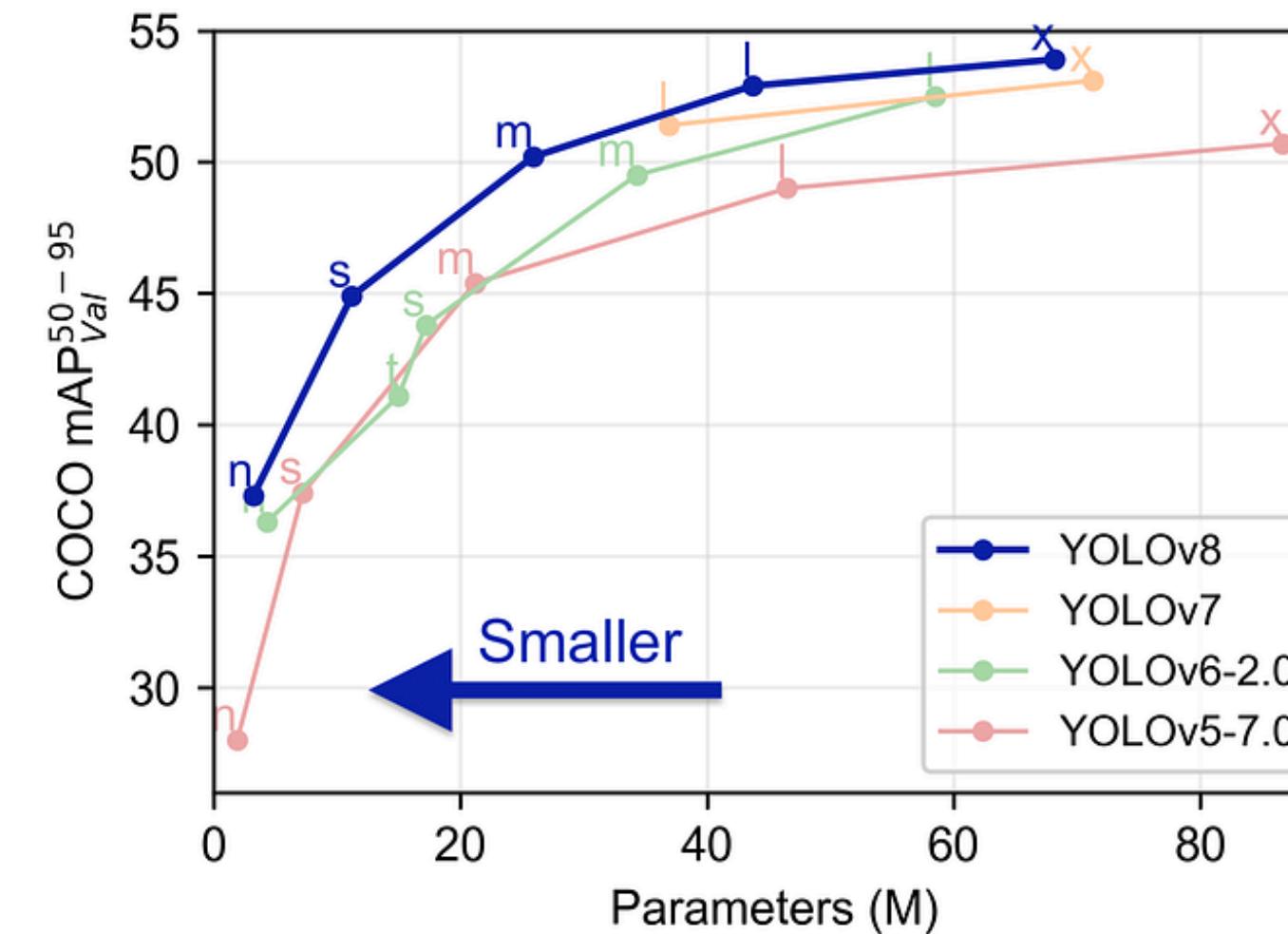
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# WHAT IS YOLOV8 ?

YOLOv8 is a real-time object detection model that provides fast and accurate detection of objects in images and videos. It is commonly used for tasks like pedestrian detection, vehicle tracking, and other vision-based applications.



# WHY VERSION 8 ?



# YOLOV8

## Model Structure:

- Model Type: YOLOv8 (You Only Look Once, version 8)

- Training Details:

Training Task: Object Detection

Dataset: Custom dataset (e.g., pedestrian data)

- Epochs: 50

- Batch Size: 64

- Augmentation Techniques:

Flip Up/Down: 50% probability

Mixup: 50% probability

Shear: 10% probability

- Early Stopping: Enabled with a patience of 10 epochs

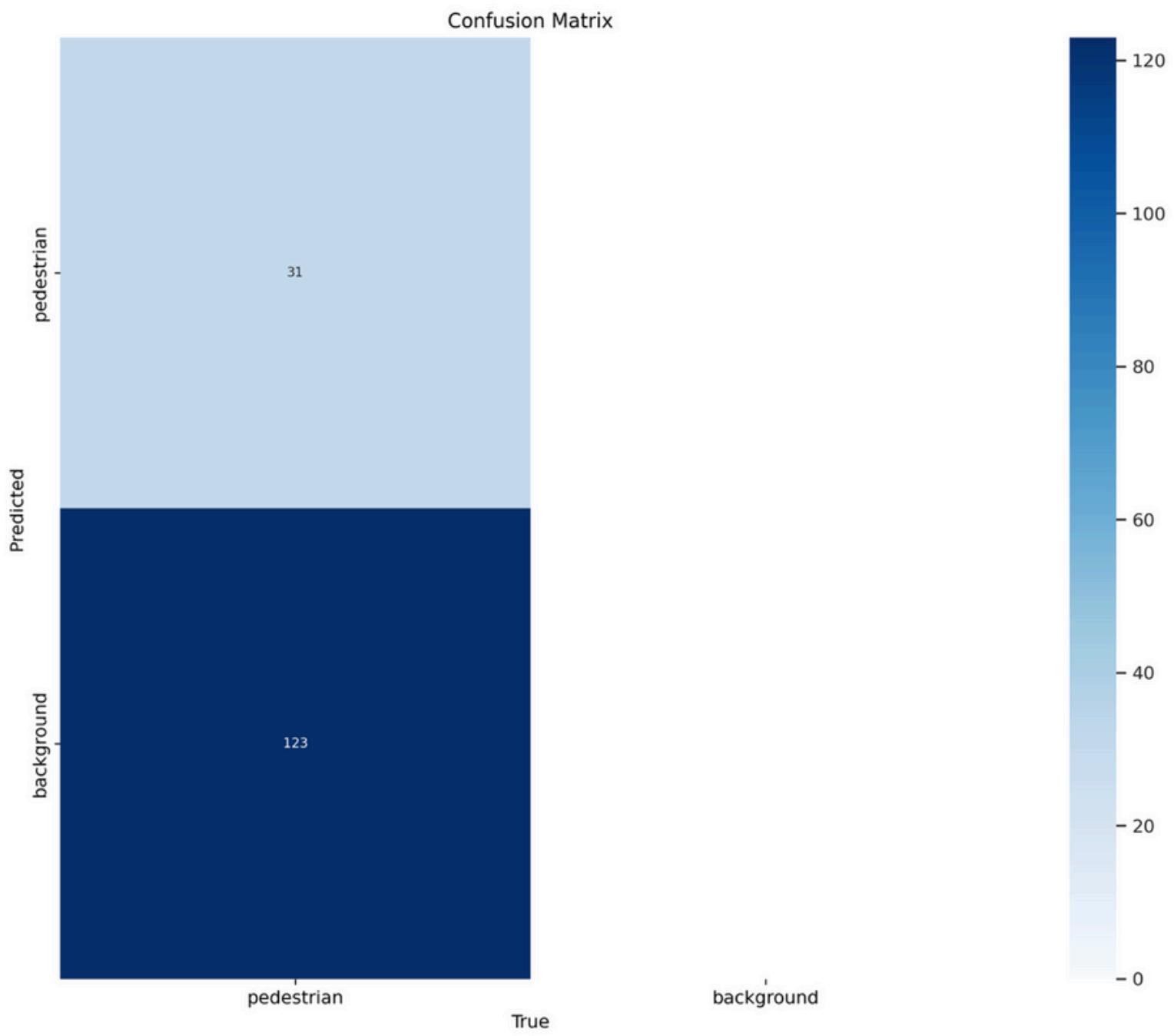
# PERFORMANCE METRICS

## Yolov8 Performance Metrics

1. **mAP Curve:** The mean Average Precision curve shows mAP at different IoU thresholds.
2. **Precision-Recall Curve:** Provides a visual representation of the trade-off between precision and recall.
3. **F1 Score:** Provides a balance between precision and recall, especially useful when the class distribution is imbalanced.
4. **Loss Curve:** This curve shows the loss value during training epochs.
5. **Intersection over Union (IoU):** IoU measures the overlap between the predicted bounding box and the ground truth bounding box.
6. **mAP@0.50 :** Measures average precision at an IoU threshold of 0.50
7. **mAP@0.50:0.95 :** Averages precision across a range of IoU thresholds , for comprehensive assessment of detection performance.
8. **Confusion Matrix:** Detailed breakdown of the model's predictions compared to the actual labels

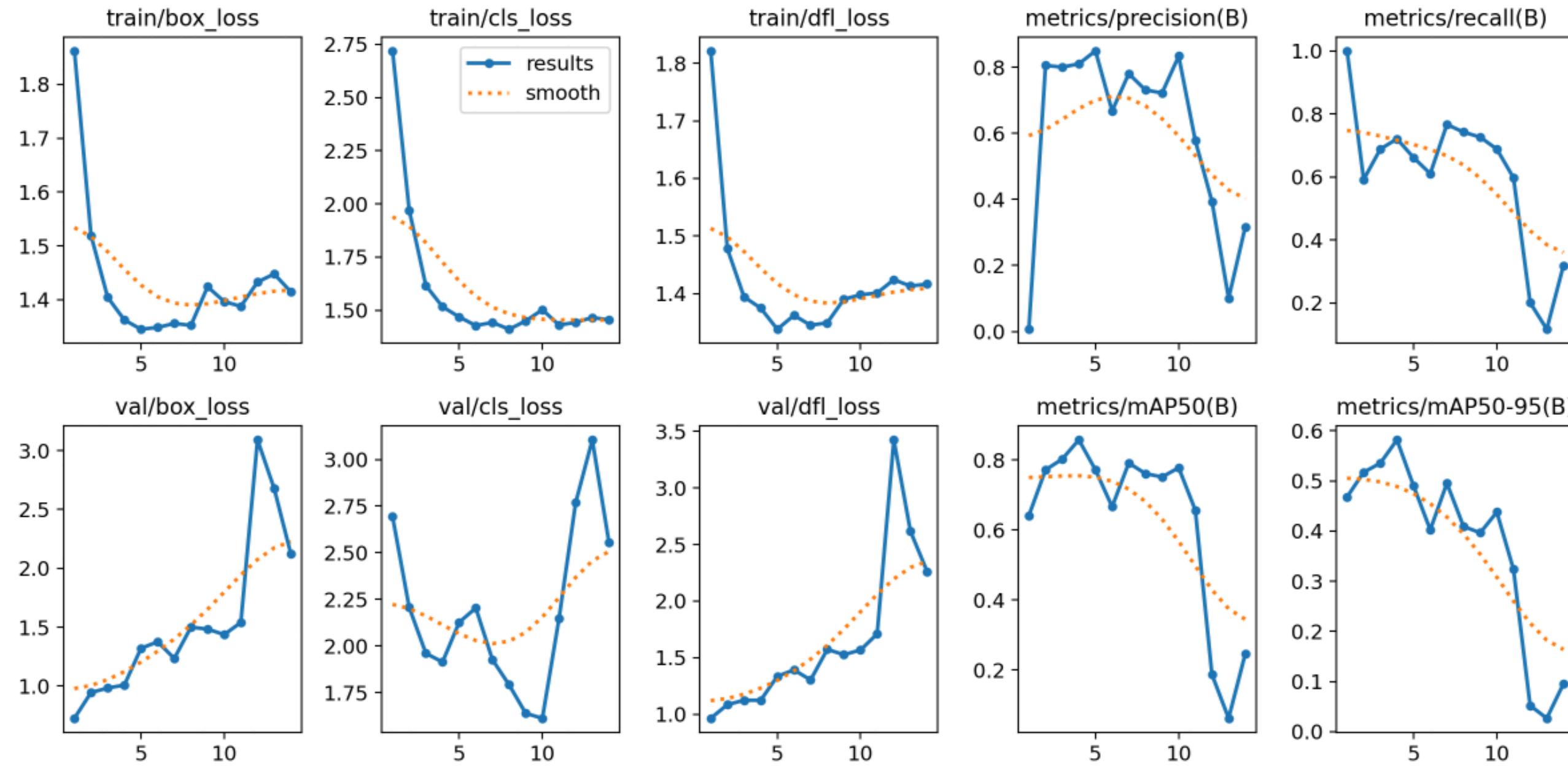
# RESULTS AND DISCUSSION

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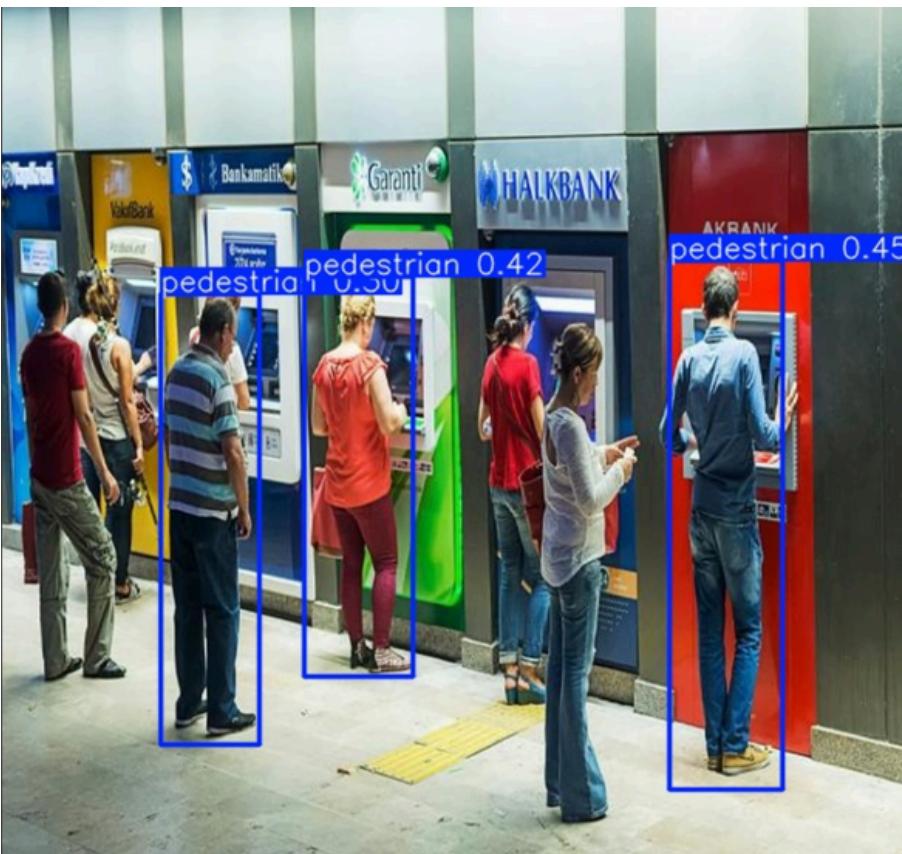
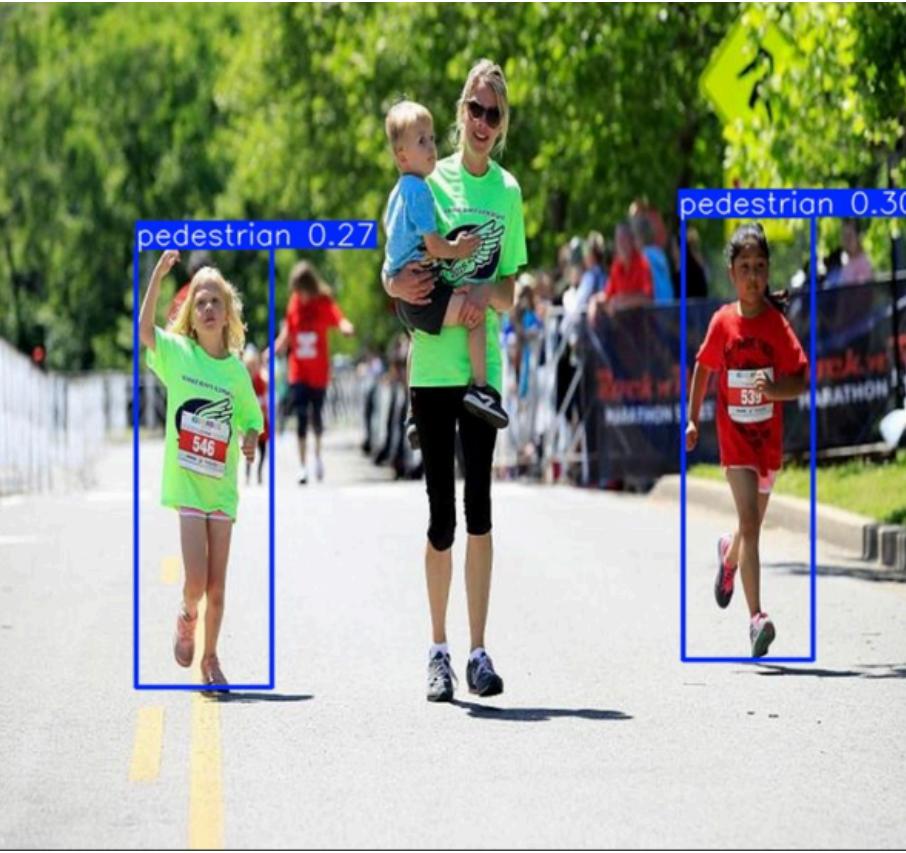
# RESULTS AND DISCUSSION

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# RESULTS AND DISCUSSION

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# DETECTION ON VIDEO RESULT

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# CONCLUSION AND FUTURE WORK

## Conclusion

- Successfully detected pedestrians in different environments (streets, ATMs, parks) using YOLOv8.
- Confidence scores vary, showing potential for model improvement.

## Future Work

- **Expand Dataset:** Increase the training data beyond the current 500 images to enhance model robustness and generalization.
- **Increase Accuracy:** Fine-tune the model to improve pedestrian detection, especially in complex environments.
- **Real-time Deployment:** Apply in smart city traffic monitoring or surveillance systems.
- **Speed Optimization:** Improve model speed for real-time video processing.

**THANK YOU  
FOR  
LISTENING!**