Vehicle Detection in Diverse Weather Conditions: YOLOv8 with Optuna Optimization

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Methodology:

• Model: YOLOv8x pre-trained, fine-tuned on AVD-Dataset

• Dataset: 3,200 images (2,600 training, 200 validation)

• Hyperparameter Optimization: Optuna framework

- Optimized: epochs, batch size, learning rate, image size

- Objective: Maximize mAP50-95

Optimal Hyperparameters:

Epochs: 8Batch size: 8

Learning rate: 2.65e-05Image size: 720 x 720 pixels

Performance Metrics:

• mAP50-95 (B): 29.24%

mAP50 (B): 56.08%Precision (B): 59.81%Recall (B): 50.20%

• Fitness score: 0.3192

Key Findings:

- 1. Strong performance on common vehicles (taxi, bike, car, bus)
- 2. Lower accuracy on smaller or less common vehicles (cycle, van)
- 3. Potential for real-time application with ~38 FPS processing speed
- 4. Room for improvement in precise localization (mAP50-95 vs mAP50)

Conclusion:

The YOLOv8-based approach with Optuna optimization shows promise for vehicle detection across diverse weather conditions. The model achieves a good balance between accuracy and speed, making it suitable for real-time applications. While performance is strong for common vehicle types, there's room for improvement, particularly for less frequent or smaller vehicles. This lays a solid foundation for further research and practical deployment in traffic monitoring and management scenarios.