

Bangladesh Road Object Detection for Autonomous Vehicles

DL Enigma 1.0 (SUST CSE Carnival)

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Competition Link: <https://www.kaggle.com/competitions/dl-enigma-10-sust-cse-carnival-2024>

Problem & Dataset

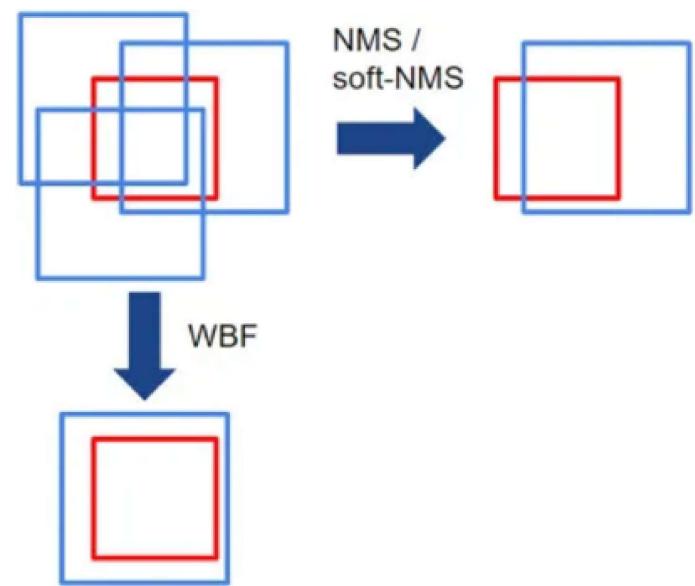
- Autonomous navigation in Bangladesh requires robust object detection across diverse roads and lighting.
- **Dataset:** BadODD – Bangladeshi Autonomous Driving Object Detection Dataset
 - 9 districts, ~9.8K images, ~79K objects, 13 classes (cars, rickshaws, buses, pedestrians, trains, wheelchairs, etc.)
 - Collected with smartphones → realistic driving conditions (day/night, motion, occlusion)
 - **Challenges:** class imbalance, varied environments, small/rare objects

Handling Class Imbalance & Lighting Variation

- Dataset had cars & buses dominant, while wheelchairs and other rare classes appeared very few times.
- Applied augmentations:
 - Flip and mosaic to increase variety.
- Used ensemble with **Weighted Box Fusion (WBF)**:
 - Boosted detection of rare classes.

Why Weighted Box Fusion (WBF)?

- Normal Non-Maximum Suppression (NMS) discards overlapping predictions.
- WBF fuses predictions → more accurate bounding boxes.
- Helps especially on rare/ambiguous objects.
- Boosted leaderboard score and reduced false positives.



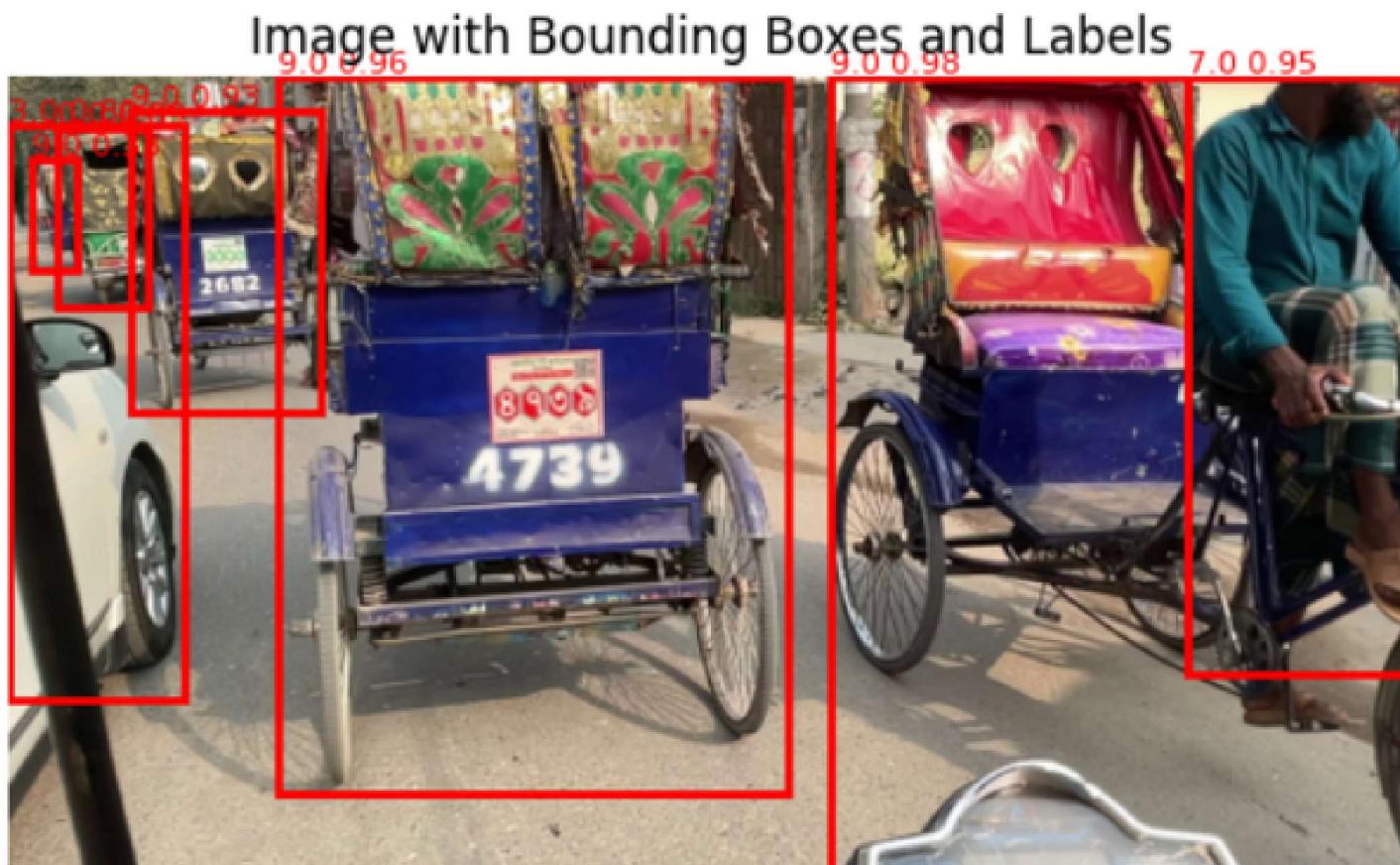
Model Selection

- Traditional detectors like YOLO use convolutional networks.
- DETR introduced transformers for object detection, great at capturing relationships between objects.
- Initially used YOLO, but moved to **RT-DETR** (Real-Time Detection Transformer):
 - Balances accuracy & real-time speed.
 - Handles multi-scale object sizes better.
 - More robust for diverse road conditions in Bangladesh.

Training and Inference Pipeline

- **Framework:** Ultralytics RT-DETR
- **Two models trained:**
 - Model 1 → 100 epochs
 - Model 2 → 60 epochs
- **Settings:** batch 16, image size 640, LR 0.018
- **Inference Pipeline:**
 - ① Two trained models loaded: Model 1 (last.pt, 100 epochs), Model 2 (best.pt, 60 epochs)
 - ② Prediction steps:
 - ① Run inference on test set.
 - ② Collect boxes, scores, classes.
 - ③ Apply Weighted Boxes Fusion (WBF) → merge predictions.
 - ③ Parameters: weights = [1,1], iou_thr = 0.4, skip_box_thr = 0.001

Output Images with Bounding Boxes and Labels



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Output Images with Bounding Boxes and Labels



Achievement

- Our team – **SUST Rising Team** – became 1st in SUST due to:
 - Careful model selection (RT-DETR)
 - Weighted Box Fusion (WBF) ensembling
 - Handling rare classes and lighting variation effectively