

Mid Evaluation Report

AI/ML-012

- **Team Details:**

Team leader:

1.) Siddhi Borase

Roll no.: 240002072

GitHub: <https://github.com/ssb120107>

LinkedIn: <https://www.linkedin.com/in/siddhi-borase-0b0353351/>

Other team members:

2.) Aarushi Pandey

Roll no.: 240003002

GitHub: <https://github.com/aarushpd12>

LinkedIn: <https://www.linkedin.com/in/aarushi-pandey-2a8678320>

3.) Shriya Deo

Roll no.: 240041013

GitHub: <https://github.com/Shriya-1807>

LinkedIn: <https://www.linkedin.com/in/shriya-deo-396581334>

4.) Tanishka Sihara

Roll no.: 240003078

GitHub: <https://github.com/tanishka-sihara>

LinkedIn: <http://www.linkedin.com/in/tanishka-sihara>

- **Team:** AI/ML-012

- **PS Name:** Drone Footage Object Detection

- **Project Workflow:**

The project workflow that the team followed:

1. *Learning:*

- (a) Learning concepts and their applications, from various sources.
- (b) Usage of essential python libraries such as Pandas, Numpy, Matplotlib.
- (c) Exploring key steps and procedure to build the project.
- (d) Basics of PyTorch.

2. *Data Collection:*

The online websites containing sources for the datasets were provided in the problem statement. After exploring and searching among the datasets, VisDrone2019 dataset was found to be the most suitable for object detection. It had pre-split folders as train, test, and validation data. Each folder contained annotations corresponding to each frame(in the case of videos dataset), or image(in the case of images dataset).

3. *Data Analysis:*

To begin with coding, it was essential to understand the data storage hierarchy. As mentioned, a total of 6 folders were downloaded: train, val, and test folders for images and similarly for video footage.

- (a) Each of the folders in image format, consists of 2 sub folders- 'annotations', 'images'. The 'annotations' sub folder consisted of annotation text files corresponding to each image stored in the 'images' sub folder.

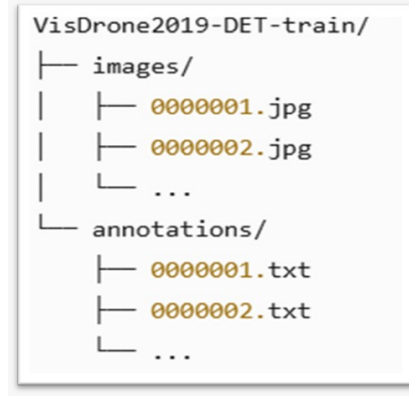


Figure 1: Dataset Hierarchy of image dataset folders

- (b) Each of the folders in video format, consists of 2 sub folders- 'annotations', 'sequences'. The 'sequences' sub folder consists of multiple sub folders for multiple video footage, each of which contains the frames as (frame-name).jpg files. The 'annotations' sub folder consisted of annotation text files corresponding to each frame stored in each of the video footage sub folders.

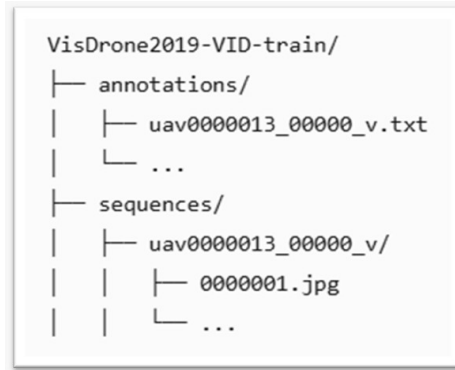


Figure 2: Dataset Hierarchy of video footage folders

4. *Data Pre-processing:*

- (a) Pre-processing the data includes transforming raw, messy, and inconsistent data into a clean and structured format, removing

noise, and making it suitable for the next step. YOLOv8n model was selected for training. This required the images/ frames to be resized to 928x928 pixels.

- (b) And, this model also requires the annotations to be converted to a specific format suitable for yolo model implementation. These requirements were satisfied during data pre-processing.
- (c) Converting RGB images to grayscale and canny edges is also a crucial step of pre-processing, but yolo models are designed to learn from the full color information present in RGB images, hence this was not a required step.
- (d) The annotations were grouped using the target ids. This would be beneficial in future for object tracking as well. This was done by grouping annotations according to a 2 dimensional tuple, grouping object category with target ids.

5. *Train-test splitting:*

- (a) According to a general flow of any ML project, the next step is to split the data into training and testing data.
- (b) But the VisDrone2019 dataset already had separate folders containing distinct training, testing, and validation data. So this step was not required.

6. *Model Training:* Model training for image dataset was done on google colab as it offered a better GPU for running 100 epochs. Due to issues faced in uploading folders on drive, training for videos dataset was done on VScode using local GPU.

- (a) YAML file was created, which serves as a configuration file that defines and manages key parameters, settings, and structures needed for the training process. The simple, indentation-based syntax of YAML makes it easy for users to read, write, and understand configuration files, even for complex projects.
- (b) The YOLOv8n model was trained for image dataset for 100 epochs on Google Colab, and for video dataset for —epochs on VScode.

7. *Evaluation*: The model was evaluated, with unsatisfactory results. Thus, it requires changes as it has huge scope for improvement.

- **Progress and Current Status:**

1. YOLOv8n model has been trained for images dataset, and run on 100 epochs. The model has been evaluated, but mAP, precision, F1 score, and other metric values are not satisfactory. The immediate next step will, definitely, be to improve on evaluation metric values.
2. The model is under training for videos dataset, and is not complete due to some errors, and efforts towards debugging are in progress.

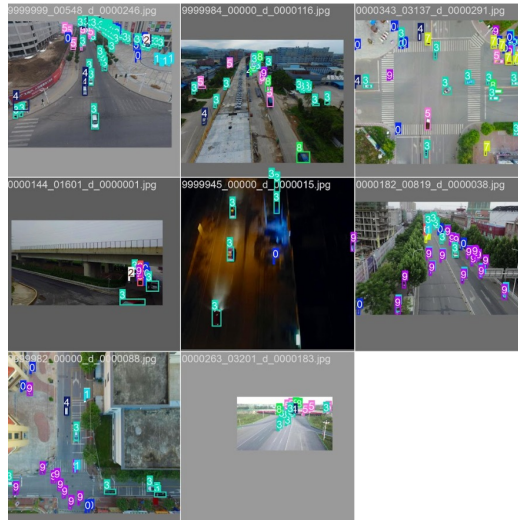


Figure 3: Images after Training

This was done for 100 epochs, which was successfully completed on Google Colaboratory, after facing multiple issues and actively resolving all of them. The following image shows the output.

```

Epoch 98/100   all      548      38759      0.496      0.376      0.389      0.232
GPU_mem      box_loss  cls_loss  dfl_loss  Instances  Size
8.11G        1.27      0.8784    0.9276    265        640: 100%
Class        Images    Instances Box(P      R
all          548      38759    0.489      0.379      0.388      0.232
mAP50      mAP50-95): 100%
405/405 [01:57<00:00, 3.44it/s]
18/18 [00:07<00:00, 2.49it/s]

Epoch 99/100   all      548      38759      0.488      0.378      0.388      0.231
GPU_mem      box_loss  cls_loss  dfl_loss  Instances  Size
8.13G        1.268     0.8746    0.9276    529        640: 100%
Class        Images    Instances Box(P      R
all          548      38759    0.488      0.378      0.388      0.231
mAP50      mAP50-95): 100%
405/405 [01:56<00:00, 3.47it/s]
18/18 [00:07<00:00, 2.32it/s]

Epoch 100/100  all      548      38759      0.485      0.38      0.389      0.232
GPU_mem      box_loss  cls_loss  dfl_loss  Instances  Size
6.52G        1.264     0.8699    0.9251    321        640: 100%
Class        Images    Instances Box(P      R
all          548      38759    0.485      0.38      0.389      0.232
mAP50      mAP50-95): 100%
405/405 [01:57<00:00, 3.45it/s]
18/18 [00:08<00:00, 2.25it/s]

45 epochs completed in 1.656 hours.
Optimizer stripped from /content/drive/MyDrive/visdrone_train/yolov8/weights/last.pt, 6.3MB
Optimizer stripped from /content/drive/MyDrive/visdrone_train/yolov8/weights/best.pt, 6.3MB

Validating /content/drive/MyDrive/visdrone_train/yolov8/weights/best.pt...
Ultralytics 8.3.160 Python-3.11.13 torch-2.6.0+cu124 CUDA:0 (Tesla T4, 15095MiB)
Model summary (fused): 72 layers, 3,007,598 parameters, 0 gradients, 8.1 GFLOPs
Class      Images    Instances Box(P      R      mAP50      mAP50-95): 100%
all         548      38759    0.499      0.377      0.389      0.232
pedestrian  520      8844    0.508      0.382      0.407      0.18
people      482      5125    0.564      0.286      0.344      0.133
bicycle     364      1287    0.314      0.131      0.128      0.0533
car         515      14804   0.734      0.784      0.816      0.579
van         421      1975    0.546      0.443      0.46      0.325
truck       266      750     0.457      0.368      0.344      0.23
tricycle    337      1045    0.435      0.291      0.271      0.146
awning-tricycle 220     532     0.267      0.19      0.147      0.0919
bus         131      251     0.63      0.474      0.539      0.392
motor       485      4886    0.534      0.419      0.437      0.192

Speed: 0.3ms preprocess, 2.4ms inference, 0.0ms loss, 5.0ms postprocess per image
Results saved to /content/drive/MyDrive/visdrone_train/yolov8
ultralytics.utils.metrics.DetMetrics object with attributes:

```

Figure 4: Output after running for 100 epochs

• Conclusion:

The trained model, with images resized to 640x640, was evaluated on evaluation metrics. The results were as follows:

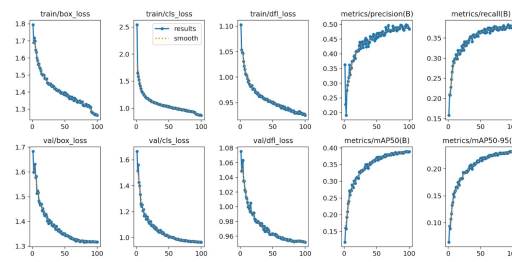


Figure 5: Results after training images resized to 640x640

Low accuracy score might have occurred due to class imbalance (more 'cars' objects, and comparatively very few 'tricycles' objects).

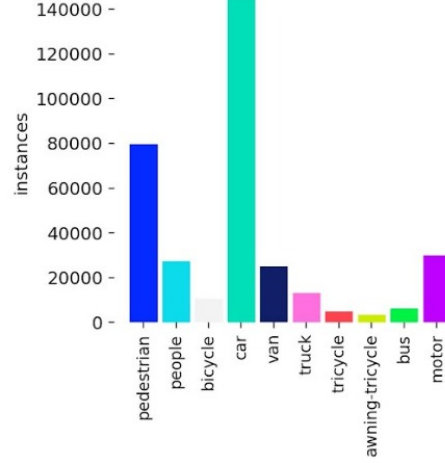


Figure 6: Class Imbalance

Other reasons could be that the images being resized to 640x640, while larger dimensions are actually required; (and other such issues). This was one-step rectified by changing image resize to 928x928. Following were the results after rectification (results were improved but still not satisfactory).

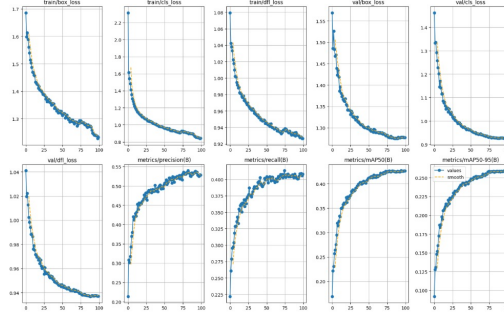


Figure 7: Results after training images resized to 928x928

The immediate next focus of the team will be to further improve on the evaluation metrics.

References

- Machine Learning Google Crash Course: <https://developers.google.com/machine-learning/crash-course>
- Machine Learning playlist: <https://youtube.com/playlist?list=PLfFghEzKVmjuYSU8vohCLxkKuod8ulUF1>
- OpenCV from freeCodeCamp.org: <https://youtu.be/oXlwWbU8l2o?si=ocHU8b7W6694Zhv6>