
VEHICLE -TYPE IDENTIFICATION PROJECT

Group 1: Matthew Caro, Kevin Huang,
Ankith Nagabandi, Yarid Tyran



OUR PROJECT

- Compare the performances of different models through training and testing accuracies
 - Identify and classify the body type of a car in an image
 - 2-Door
 - 4-Door
 - SUV
 - Convertible
 - Pickup
 - Van
-

OUR DATASET

- Preprocessing-
 - 60,000 Images
- Postprocessing-
 - Roughly 45,000 Images
 - ~34,000 Train
 - ~10,000 Test/Evaluation
- Removed some classes such as 3-Door

Acura_ILX_2013_28_16_110_15_4_70_55_179_39_FWD_5_4_4dr_Bbw.jpg

About this file

Image of an Acura

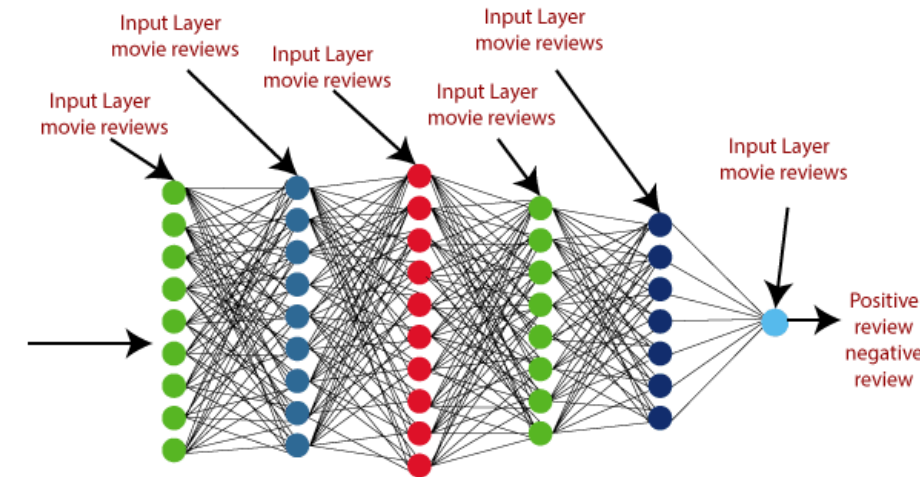
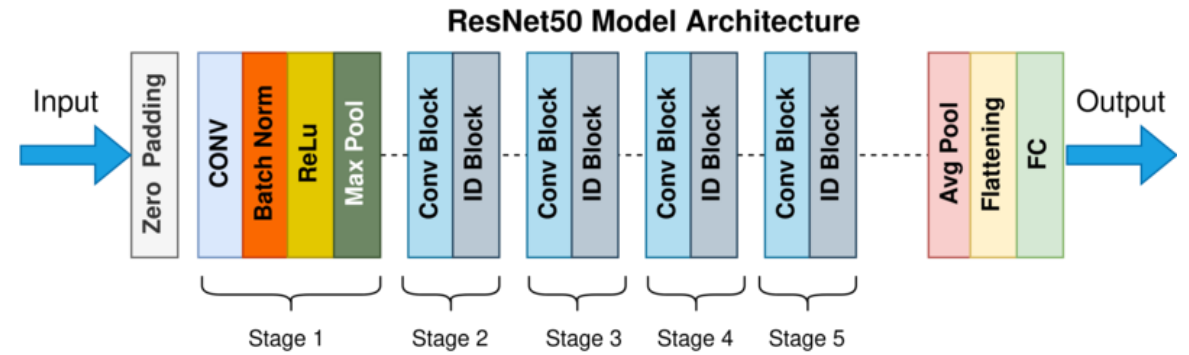
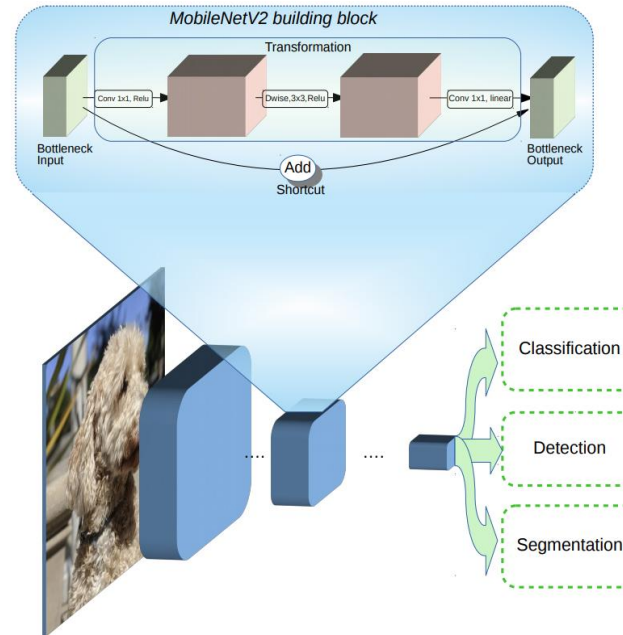


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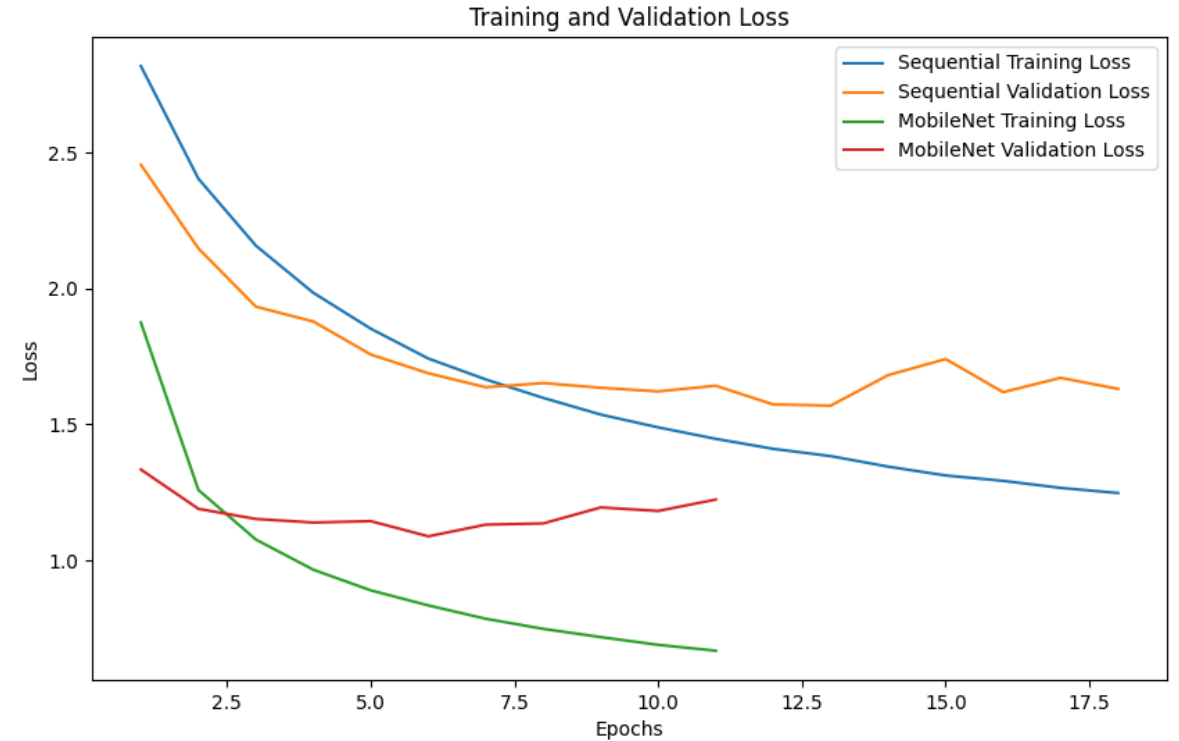
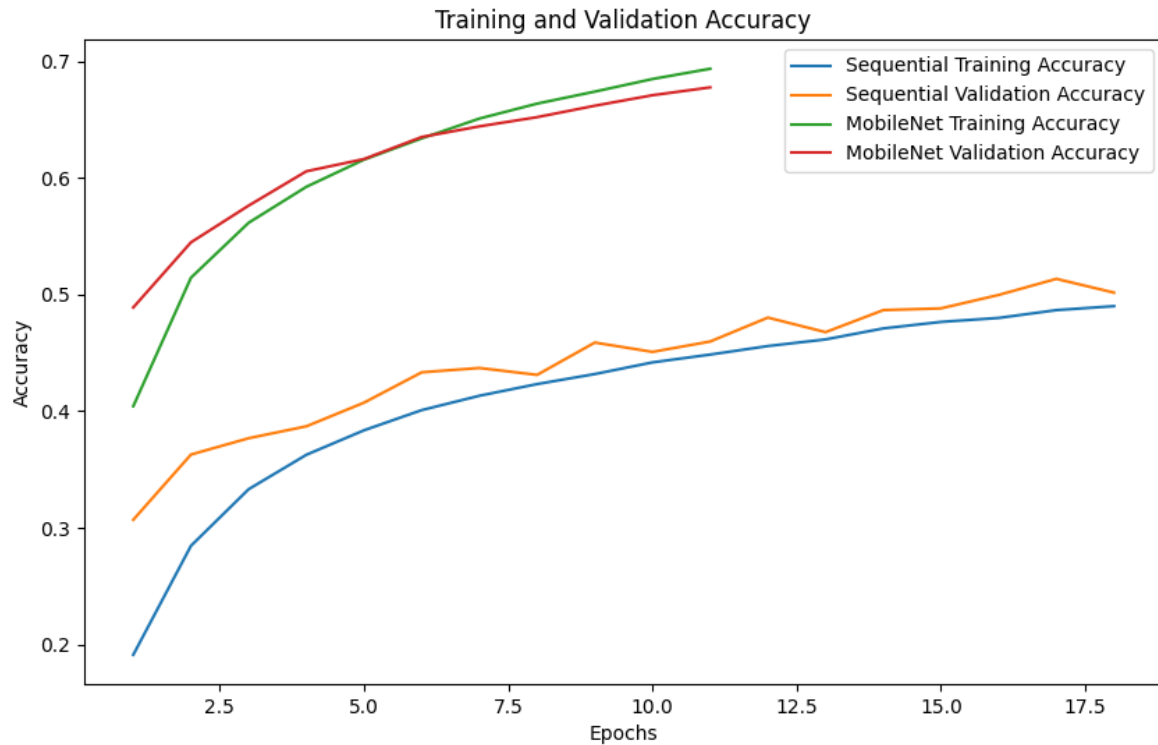
OUR MODELS

- Used CNNs
 - Sequential
 - MobileNet
 - ResNet50

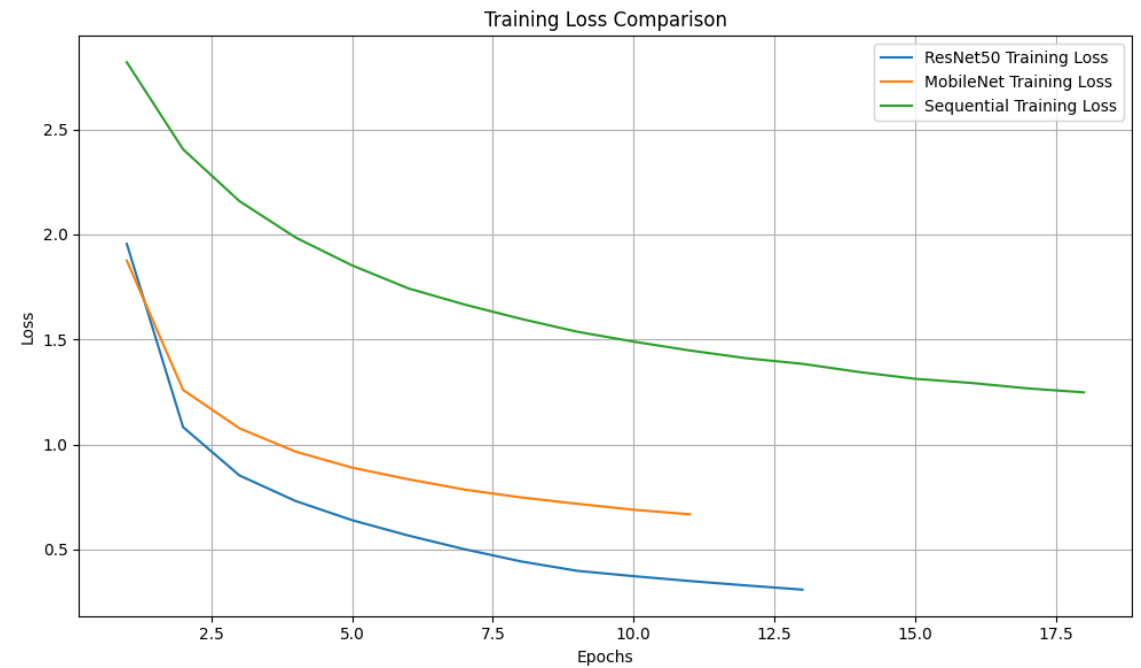
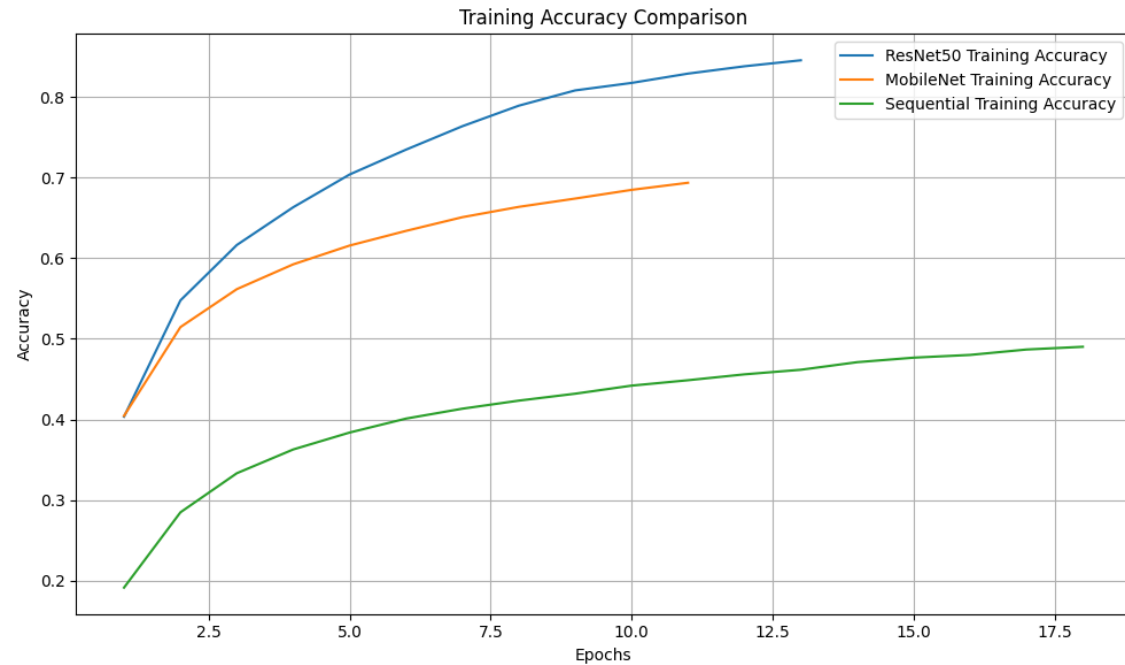


Keras Neural Network Sequential Model

SEQUENTIAL AND MOBILE NET PERFORMANCE



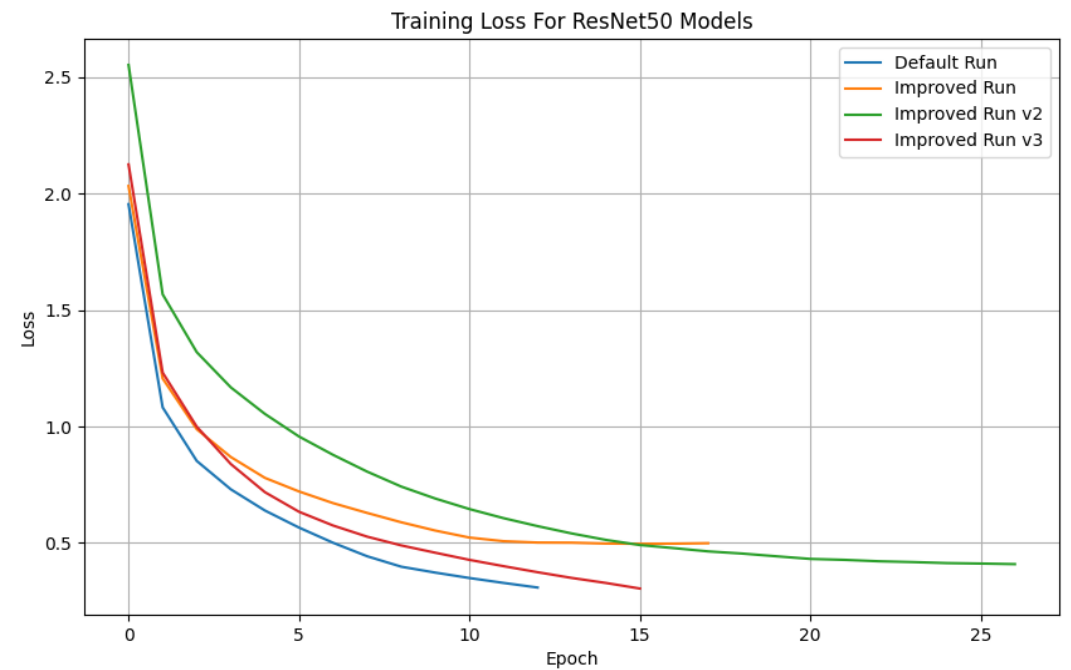
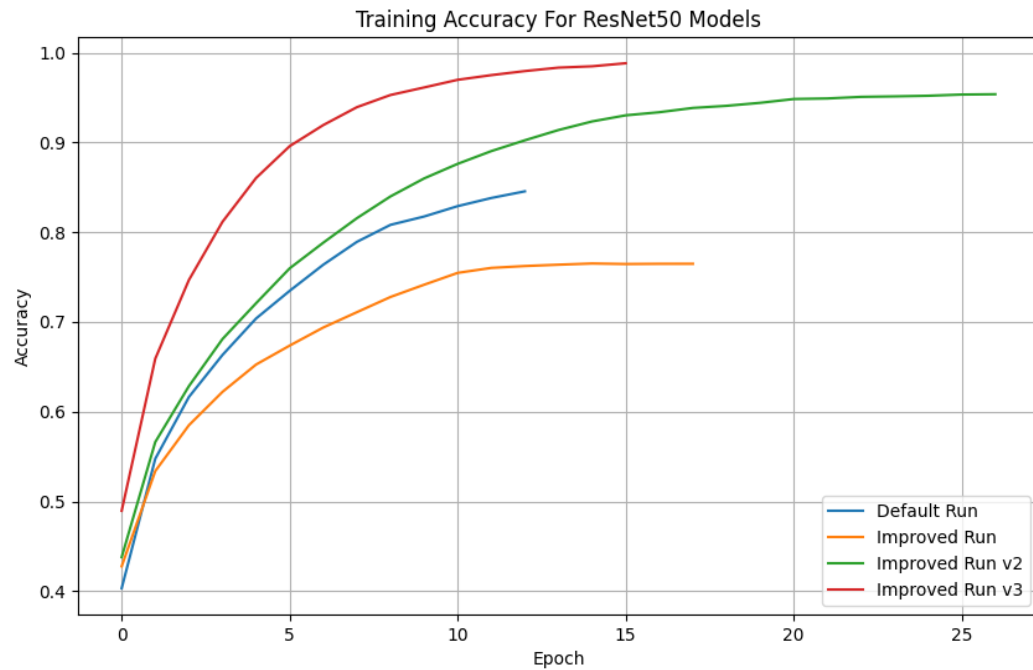
RESNET50 VERSUS OTHER MODELS



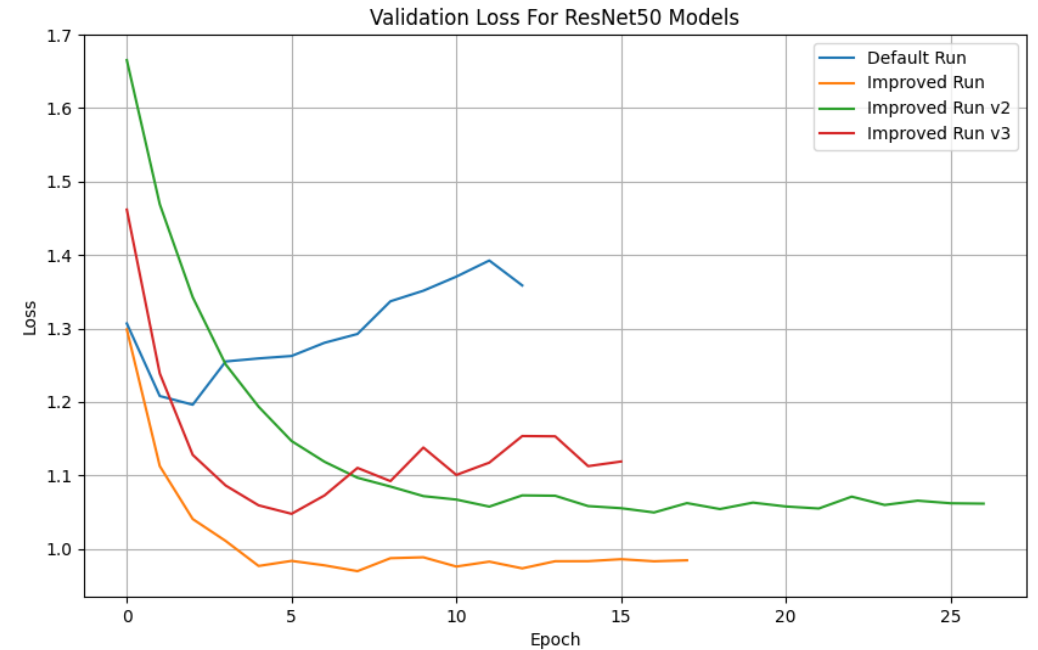
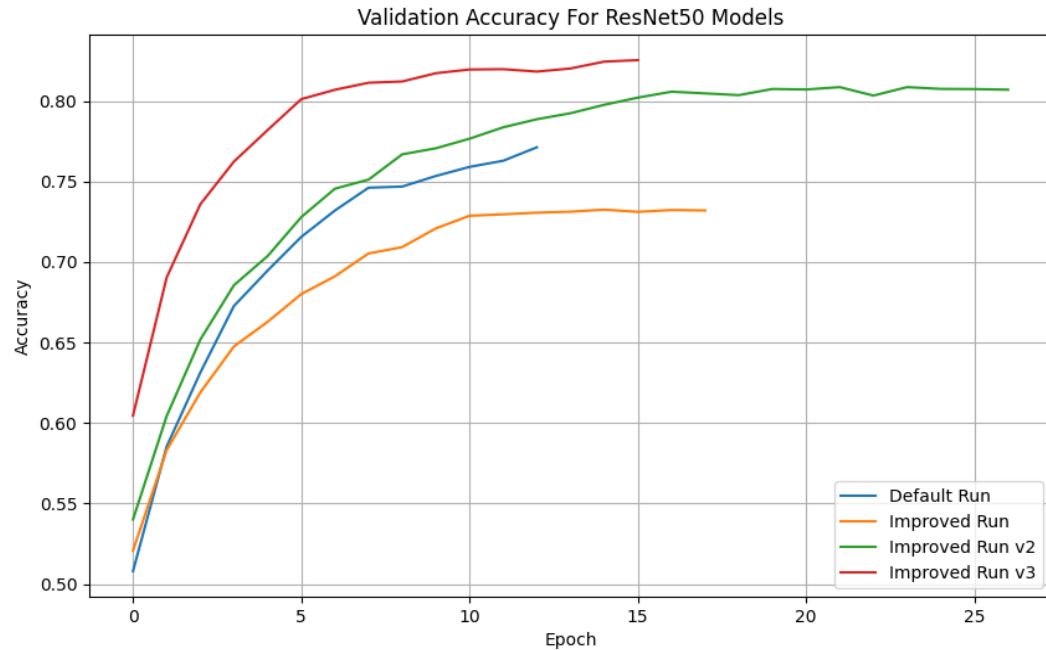
RESNET50 TRAINING TECHNIQUES

- Fine-Tuning Layers
 - Adjust trainable layers
 - Unfreezing and freezing layers to control deepness of fine-tuning
 - Optimizers and Learning Rate Adjustment
 - Adam optimizer
 - Stochastic Gradient Descent
 - Adjust learning rate
 - ReduceLROnPlateau – Dynamically adjust learning rate based on validation loss plateau (Later iterations)
 - Regularization
 - Dropout layers set to 50%
 - L2 regularization on dense layers
 - Class Imbalance
 - Class weights (scikit-learn compute_class_weight function)
 - Weights higher importance to underrepresented classes in loss function
 - Dynamic Training
 - EarlyStopping - Prevent overfitting based on validation loss
 - ModelCheckpointing – Save model at different instances of training
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RESNET50 TRAINING PERFORMANCE (TRAINING)



RESNET50 VALIDATION PERFORMANCE (TRAINING)



RESNET50 TRAINED MODELS EVALUATION (TEST SET)

```
Classification Report:
      precision    recall  f1-score   support

   Van           0.18      0.89      0.30        197
   SUV           0.85      0.55      0.67       3460
  Pickup         0.65      0.97      0.77        971
 Convertible     0.40      0.57      0.47       1185
       4dr        0.79      0.15      0.25       3367
       2dr        0.24      0.70      0.35       1119

 accuracy              0.48       10299
 macro avg           0.52      0.64      0.47       10299
 weighted avg        0.68      0.48      0.48       10299
```

Default Run

```
Classification Report:
      precision    recall  f1-score   support

   Van           0.42      0.91      0.58        197
   SUV           0.86      0.73      0.79       3460
  Pickup         0.74      0.98      0.84        971
 Convertible     0.48      0.67      0.56       1185
       4dr        0.83      0.38      0.52       3367
       2dr        0.35      0.74      0.47       1119

 accuracy              0.64       10299
 macro avg           0.61      0.74      0.63       10299
 weighted avg        0.73      0.64      0.64       10299
```

Improved Run V1

```
Classification Report:
      precision    recall  f1-score   support

   Van           0.73      0.84      0.78        197
   SUV           0.86      0.86      0.86       3460
  Pickup         0.91      0.96      0.94        971
 Convertible     0.68      0.73      0.71       1185
       4dr        0.82      0.74      0.77       3367
       2dr        0.61      0.70      0.65       1119

 accuracy              0.80       10299
 macro avg           0.77      0.81      0.79       10299
 weighted avg        0.80      0.80      0.80       10299
```

Improved Run V2

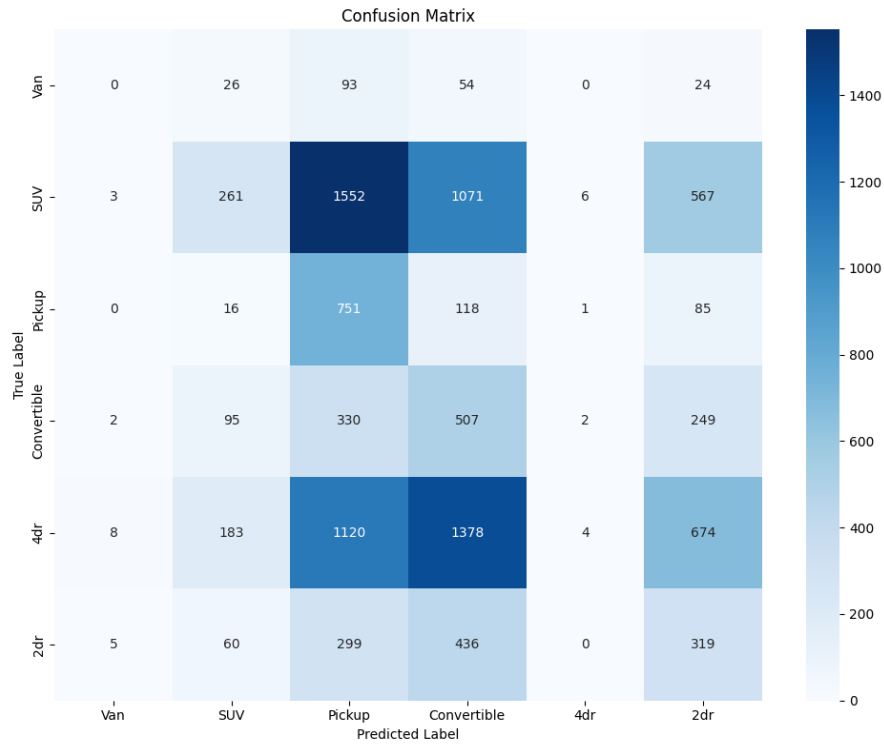
Improved Run V3

```
Classification Report:
      precision    recall  f1-score   support

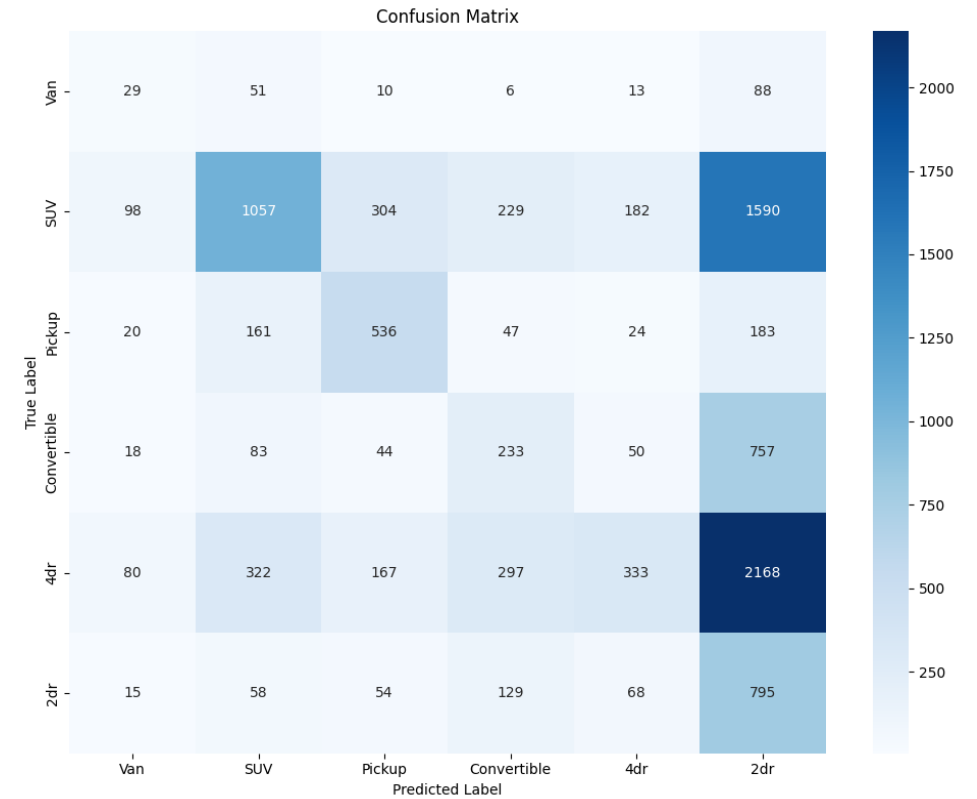
   Van           0.69      0.87      0.77        197
   SUV           0.88      0.85      0.86       3460
  Pickup         0.88      0.97      0.92        971
 Convertible     0.66      0.70      0.68       1185
       4dr        0.81      0.70      0.75       3367
       2dr        0.55      0.72      0.62       1119

 accuracy              0.78       10299
 macro avg           0.74      0.80      0.77       10299
 weighted avg        0.79      0.78      0.78       10299
```

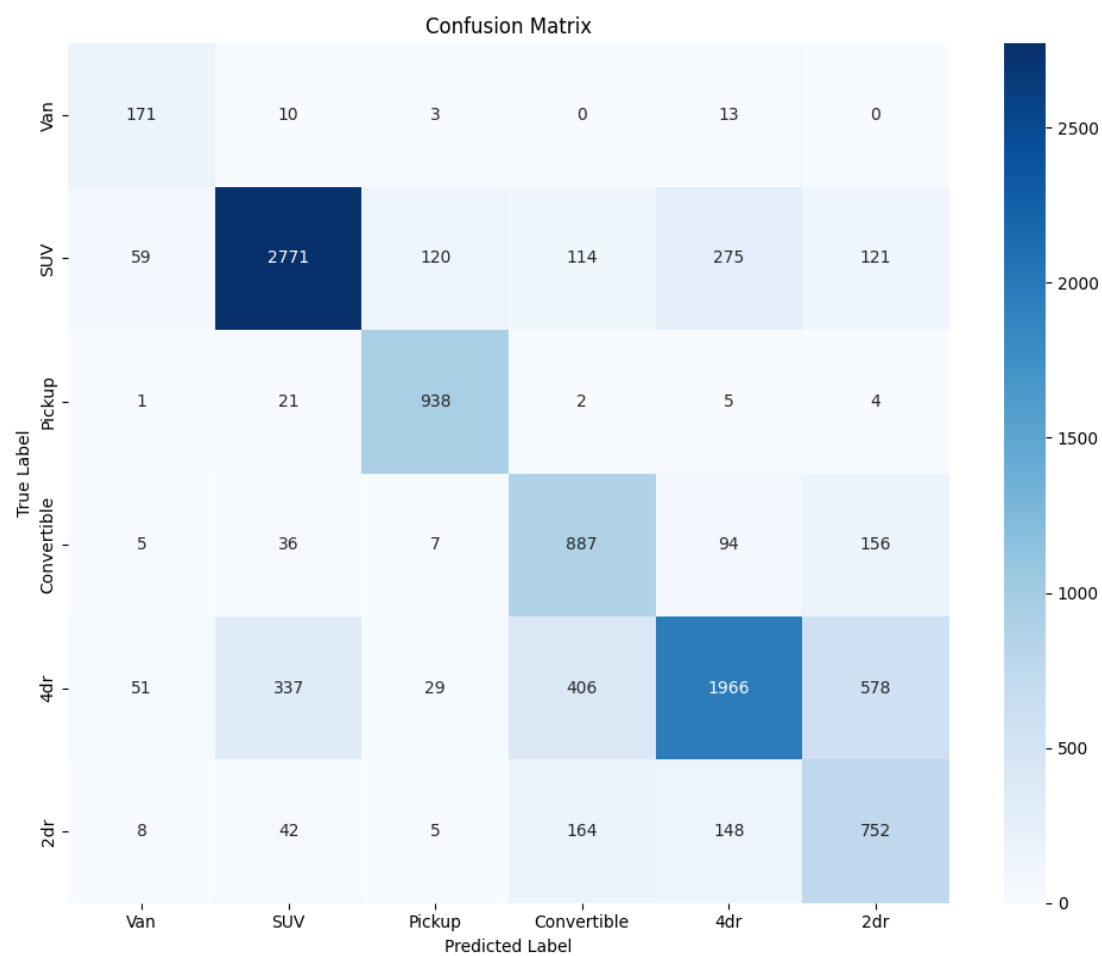
SEQUENTIAL & MOBILENET



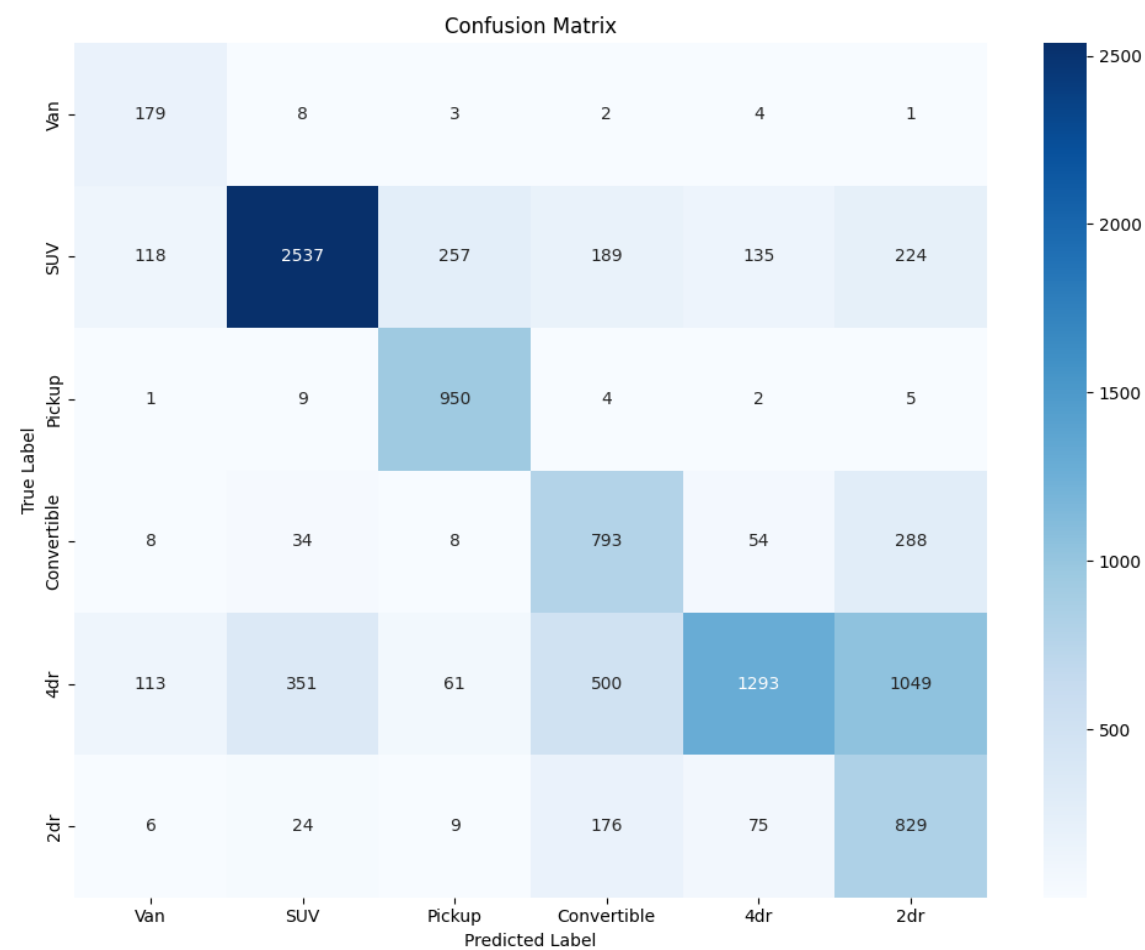
Sequential



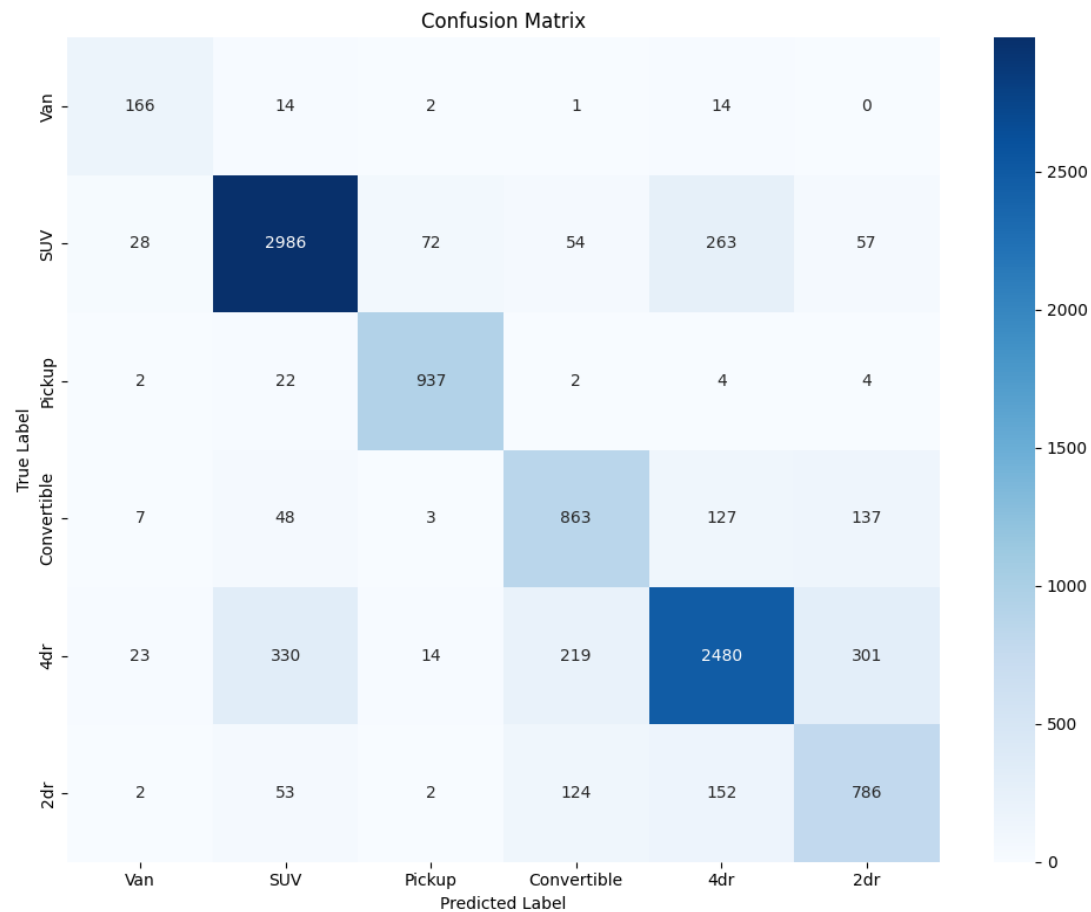
MobileNet



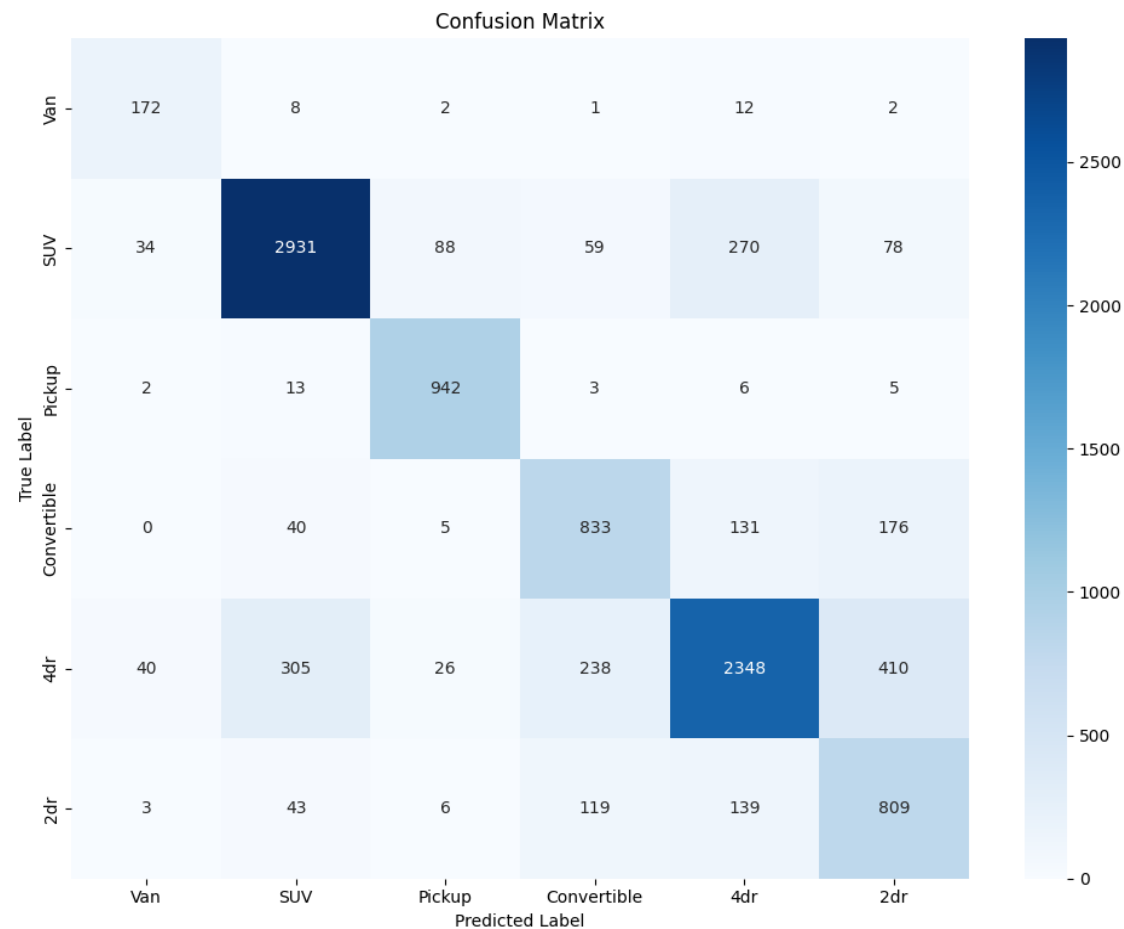
Resnet Default



Resnet V1 Iteration



Resnet V2 Iterations



Resnet V3 Iterations

ERROR ANALYSIS

- Accuracy on evaluation set was around 80%
- Post-processed images still included some interior images or part of the vehicle
- Unbalanced data- Different classes were overrepresented and underrepresented
 - Duplicate images
 - Vans had around ~600 images in total vs SUV had over 11,000
- Most of the images were very clean, no clutter
 - Only the car in the center, nothing else
 - For more real-life applications, we should use images with varied clutter
- Low sample size in evaluation set

AREAS OF IMPROVEMENT & FUTURE WORK

- More Data!
 - More even distributing
- Cleaner data
 - Less error images
- More Models
- Expanding past Vehicle Type

Data-driven Process Improvement



USER INTERFACE DEMO

[HTTP://10.194.212.45:8501/](http://10.194.212.45:8501/)
