



Bottle Image Classification Final Presentation

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Problem

The problem we are solving through our project is to be able to **classify different types of bottles through image classification**. We think this could be useful in practical applications, like sorting for recycling. The below shows different types of bottles that we want to be able to label.

Water Bottle



Water Bottle



Beer Bottles



Wine Bottle



Water Bottle



Soda Bottle



Wine Bottle



Plastic Bottles



Soda Bottle



Beer Bottles



Initial Modeling

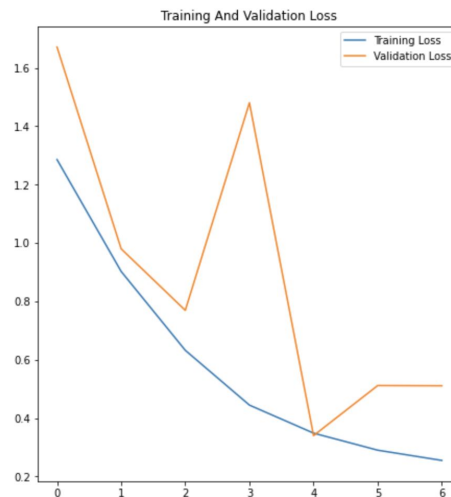
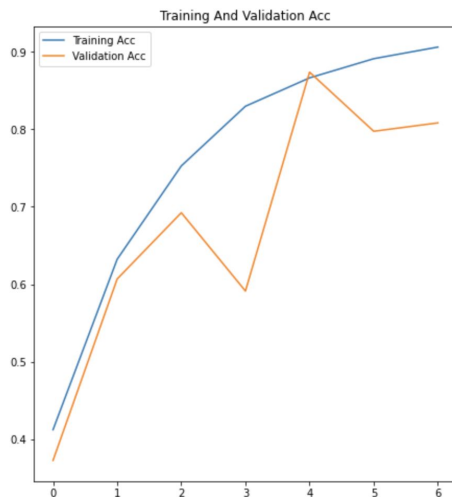
- In our last progress presentation, we presented the below results for our baseline model and a more advanced model architecture using ResNet
- Our takeaways from that presentation were to explore other model architectures and explore a realistic approach to reducing the long runtime of the ResNet model

Model	Accuracy	Model Runtime
Baseline CNN Model	0.91 (10 epochs)	~ 1 hour
ResNet	0.96 (9 epochs)	~ 3 hours (timed out)

*Increased accuracy using
ResNet model architecture*

ResMLP Model

- In exploring other architectures, we implemented another model architecture using ResMLP with a validation accuracy of ~80% but much faster training time
- In this code, we implemented early stopping based on the validation loss increasing for two epochs



Early stopping due to validation loss

```
early_stopping = tf.keras.callbacks.EarlyStopping(monitor='val_loss', patience=2, min_delta=1e-4)
```

Overall Results

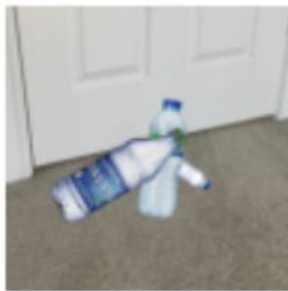
- Due to the lower accuracy of the ResMLP architecture, we chose the ResNet architecture as our final model
- To realistically deal with the runtime issue for ResNet, we reduced the epoch number for the ResNet model to 5 without sacrificing accuracy and drastically improving runtime

Model	Accuracy	Model Runtime
Baseline CNN Model	0.91 (10 epochs)	~ 1 hour
ResNet	0.95 (5 epochs)	~ 1 hour
ResMLP	0.80 (6 epochs)	~ 20 minutes

Model Testing (1/2)

- To confirm the model was predicting correctly, we tested the model on random images that we choose from the data and confirmed that the images matched the labels

Beer Bottles:True Plastic Bottles:True Soda Bottle:True Soda Bottle:True Water Bottle:True



Model Testing (2/2)

- We also tested the model on images that we took ourselves to confirm that the model could predict correctly outside of the dataset

Wine Bottle:True



Soda Bottle:True



Plastic Bottles:True



Plastic Bottles:True



Plastic Bottles:True



Water Bottle:True



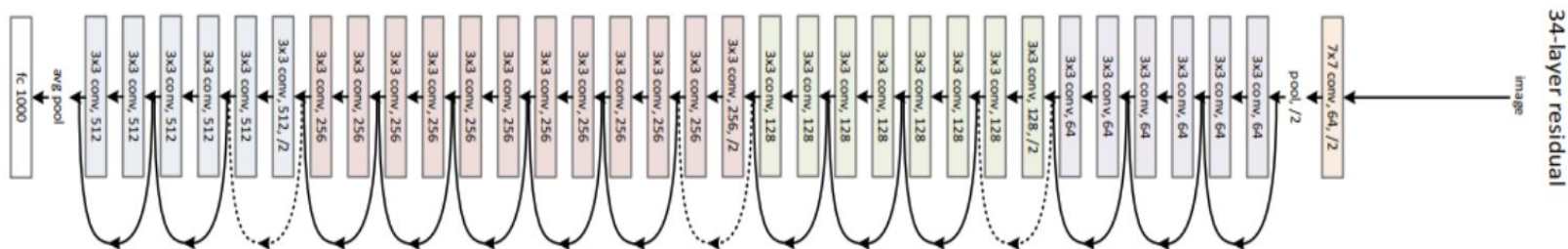
Beer Bottles:True



Conclusions

- Image classification accuracy improved with advanced model architectures
- Having a large dataset is helpful to training, but can create an issue for training time
 - Did not do data augmentation as many images in the dataset were already shrunk, rotated or on different backgrounds
 - Would consider data augmentation for a different data set with less images
- Lengthy training time might be an issue, even with GPU resources
- Github Link: <https://github.com/sydneysimmons/bottle-image-classification-project>

Appendix: ResNet



ResNet -34 architecture

Source: <https://www.geeksforgeeks.org/residual-networks-resnet-deep-learning/>

Appendix: ResMLP

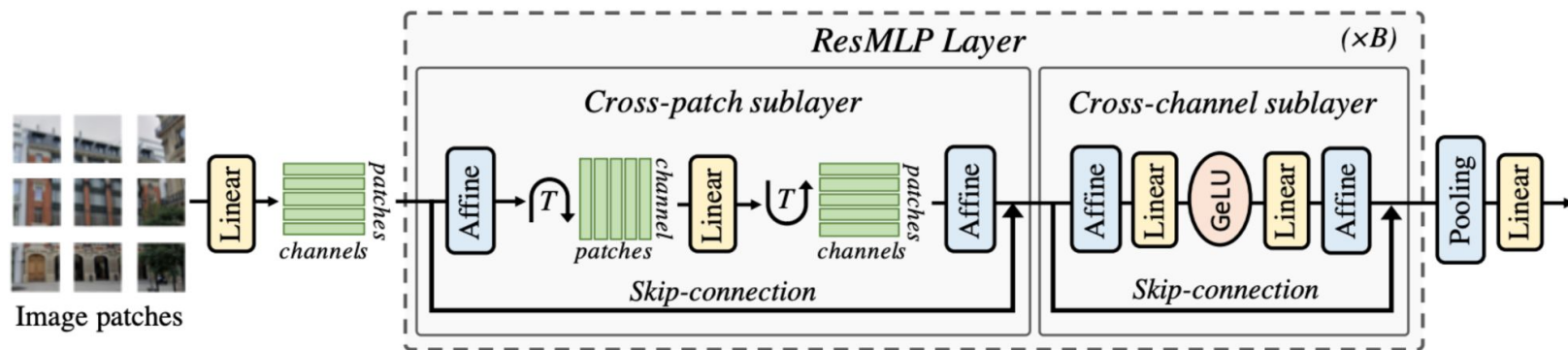


Figure 1: **The ResMLP architecture.** After linearly projecting the image patches into high dimensional embeddings, ResMLP sequentially processes them with (1) a cross-patch linear sublayer; (2) a cross-channel two-layer MLP. The MLP is the same as the FCN sublayer of a Transformer. Each sublayer has a residual connection and two Affine element-wise transformations.

Source: <https://paperswithcode.com/method/resmlp>