

5243 Project 3: Image Predictive Analytics

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Task

We are tasked with building 2 additional image classification models that could improve upon a clients existing logistic regression model and tackle the issue of label noise

Priorities of model:

- Portability (holding storage and memory cost)
- Computational efficiency (running time cost)
- Predictive performance



Baseline Model

- Simple Logistic Regression model that treats the noisy labels as clean labels
- Accuracy: 20%
- Quick run time

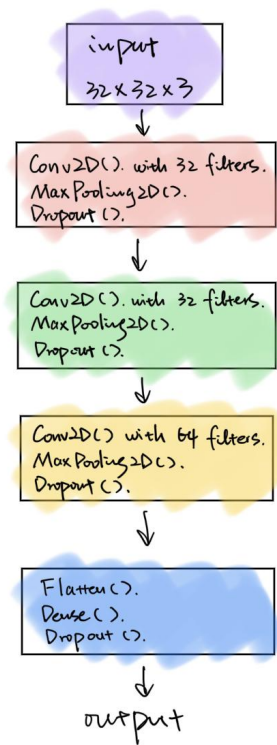
Logic Procedure for Model I

Dataset: 10k clean labels + 40k noisy labels = 50k images

- Assumes the input 50k dataset are all clean (ie, assuming no noisy labels).
- Mainly have 3 CNN layers + other layers such as flatten and dropout

The accuracy for model I is about 20%.

Training time: 5 mins



Logic Procedure for Model II

Dataset: 10k clean labels & 40k noisy labels

Model II uses effective strategies such as **transfer learning** (Mobile Net) and **label correction** to handle label noise

Step 1: Split 10k images with clean label into train (70%) & test (30%) then fit the Model by the training dataset.

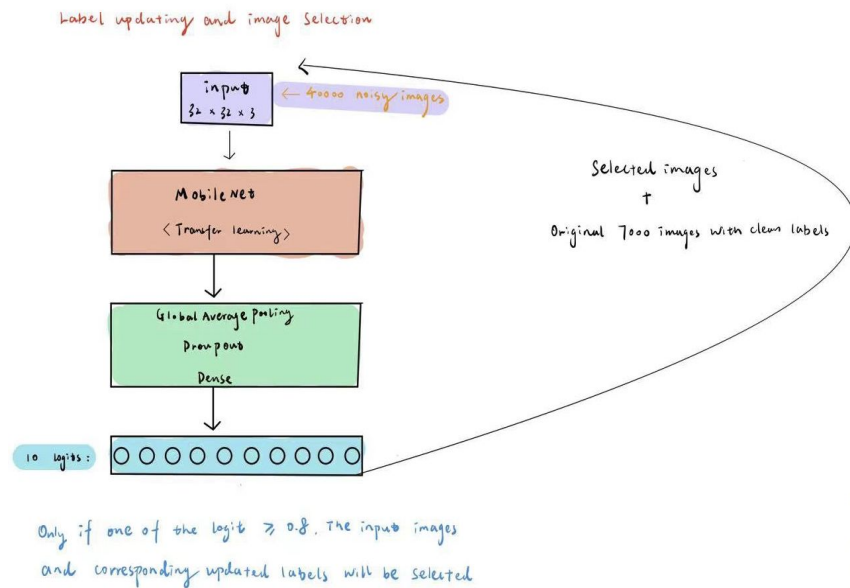
Step 2 (Label updating): Use trained model from step 1 to predict the labels of 40k images with noisy labels

Step 3 (Image selection): Each image will get 10 logits (classes) from step 2. We only select an image if one of its logit > 0.85 (setting parameter) as a way to be confident in our predicted labels

Step 4: Combine the selected images with the original training dataset (70%) to refit the model

Model accuracy : **70%**.

Runtime: about 20 minutes





Hyperparameters of Model II

Optimizer: Adam

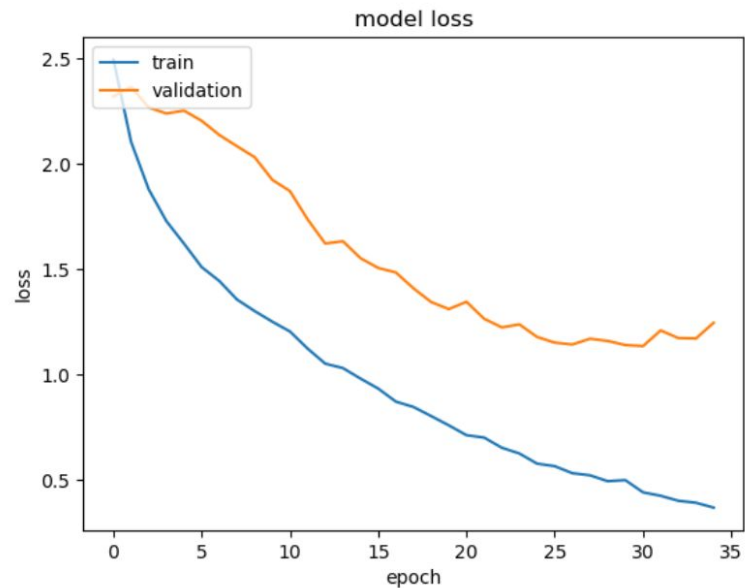
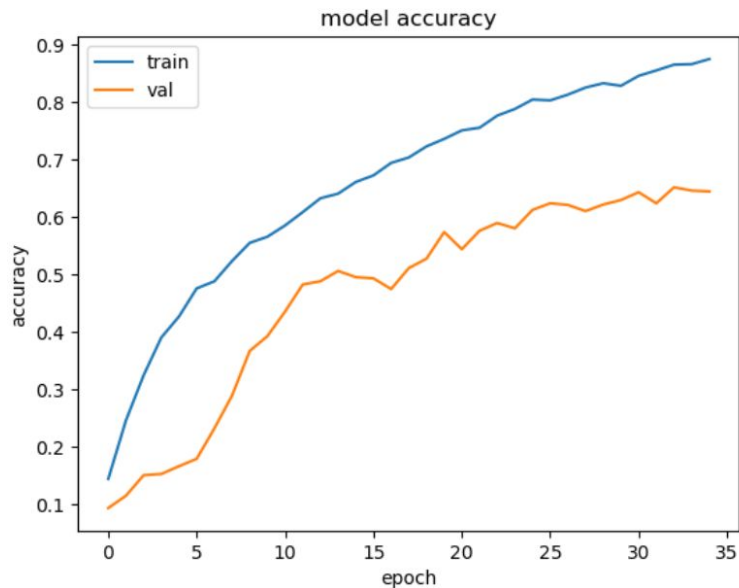
Learning rate: 0.0001

Batch size: 32

Epoch: 60

Early stop patience : 4

Model II Training and Validation for Accuracy and Loss





Model II Notes:

1. Cannot guarantee how many images will be selected for new training set in model II
 - a. The setting parameter 0.85 can change. After testing, we can get a reasonable number of images from the 40k images (eg, 22546/40000) by setting it to 0.85.
2. The sum of the 10 logits from model II equals to 1



Final Recommendation

- Model II provides a large increase in accuracy (70%) against the client's baseline model and would be a better option
- The runtime is slightly longer than the logistic regression model, but we will accept that for the increase in accuracy
- Interesting strategies used for label noise:
 - a. Transfer learning
 - b. Label correction