MSBD 6000B Deep Learning Project 2

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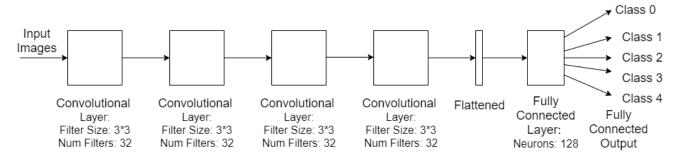
1. Data Load

I read the images' paths and labels from train.txt and val.txt. For paths, I read the target images and store them into a list of tensors. For labels, I predefine an all zeros list for each image with the length of classes and set the #label content as 1. Each label list for image contains the probabilities that this image should be classified into this class. For example, [0, 0, 1, 0, 0] means this image has 100% probability to be classified into class No.2.

2. CNN model

I choose CNN with tensorflow to build the model for this project.

2.1



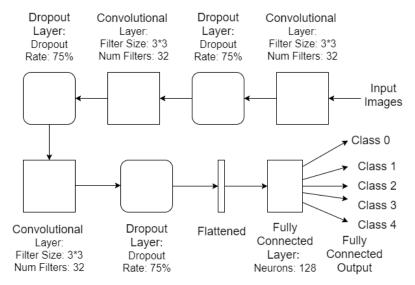
In this CNN model, there exists overfitting problem. In the later epochs, the training accuracies are around 100%, while the validation accuracy is around 60%.



```
Training Epoch 26—Training Accuracy: 99.0%, Validation Accuracy: 56.8%, Validation Loss: 0.662
Training Epoch 27—Training Accuracy: 99.7%, Validation Accuracy: 57.9%, Validation Loss: 0.701
Training Epoch 28—Training Accuracy: 99.7%, Validation Accuracy: 59.0%, Validation Loss: 0.781
Training Epoch 30—Training Accuracy: 99.7%, Validation Accuracy: 59.0%, Validation Loss: 0.817
Training Epoch 31—Training Accuracy: 99.9%, Validation Accuracy: 59.0%, Validation Loss: 0.832
Training Epoch 32—Training Accuracy: 100.0%, Validation Accuracy: 59.7%, Validation Loss: 0.881
Training Epoch 33—Training Accuracy: 100.0%, Validation Accuracy: 59.2%, Validation Loss: 0.954
Training Epoch 33—Training Accuracy: 100.0%, Validation Accuracy: 60.1%, Validation Loss: 0.971
Training Epoch 35—Training Accuracy: 100.0%, Validation Accuracy: 60.1%, Validation Loss: 0.981
Training Epoch 36—Training Accuracy: 100.0%, Validation Accuracy: 60.3%, Validation Loss: 0.991
Training Epoch 37—Training Accuracy: 100.0%, Validation Accuracy: 60.9%, Validation Loss: 0.994
Training Epoch 38—Training Accuracy: 100.0%, Validation Accuracy: 59.7%, Validation Loss: 0.994
Training Epoch 39—Training Accuracy: 100.0%, Validation Accuracy: 59.7%, Validation Loss: 0.994
Training Epoch 49—Training Accuracy: 100.0%, Validation Accuracy: 59.7%, Validation Loss: 0.994
Training Epoch 41—Training Accuracy: 99.8%, Validation Accuracy: 59.6%, Validation Loss: 0.812
Training Epoch 43—Training Accuracy: 99.8%, Validation Accuracy: 59.6%, Validation Loss: 0.813
Training Epoch 44—Training Accuracy: 99.9%, Validation Accuracy: 59.6%, Validation Loss: 0.813
Training Epoch 44—Training Accuracy: 99.9%, Validation Accuracy: 59.6%, Validation Loss: 0.813
Training Epoch 44—Training Accuracy: 99.9%, Validation Accuracy: 59.6%, Validation Loss: 0.813
Training Epoch 45—Training Accuracy: 99.9%, Validation Accuracy: 59.6%, Validation Loss: 0.813
Training Epoch 47—Training Accuracy: 99.9%, Validation Accuracy: 59.6%, Validation Loss: 0.813
Training Epoch 47—Training Accuracy: 99.9%
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2.2

To solve overfitting problem, I reduce the number of convolutional layers and add some dropout layer.



This kind of change cannot obviously improve the validation accuracy although it seems that the overfitting problem is partially solved.

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Training Epoch 1—Training Accuracy: 26.5%, Validation Accuracy: 43.6%, Validation Loss: 0.274

Training Epoch 3—Training Accuracy: 52.5%, Validation Accuracy: 49.8%, Validation Loss: 0.257

Training Epoch 4—Training Accuracy: 57.2%, Validation Accuracy: 50.5%, Validation Loss: 0.254

Training Epoch 5—Training Accuracy: 61.0%, Validation Accuracy: 54.9%, Validation Loss: 0.242

Training Epoch 6—Training Accuracy: 61.8%, Validation Accuracy: 56.1%, Validation Loss: 0.238

Training Epoch 7—Training Accuracy: 76.3%, Validation Accuracy: 59.2%, Validation Loss: 0.234

Training Epoch 10—Training Accuracy: 70.3%, Validation Accuracy: 59.2%, Validation Loss: 0.240

Training Epoch 11—Training Accuracy: 78.9%, Validation Accuracy: 60.6%, Validation Loss: 0.250

Training Epoch 12—Training Accuracy: 79.9%, Validation Accuracy: 60.0%, Validation Loss: 0.256

Training Epoch 13—Training Accuracy: 82.1%, Validation Accuracy: 60.0%, Validation Loss: 0.265

Training Epoch 14—Training Accuracy: 83.5%, Validation Accuracy: 60.0%, Validation Loss: 0.265

Training Epoch 15—Training Accuracy: 85.3%, Validation Accuracy: 60.0%, Validation Loss: 0.265

Training Epoch 16—Training Accuracy: 85.3%, Validation Accuracy: 60.0%, Validation Loss: 0.265

Training Epoch 16—Training Accuracy: 89.3%, Validation Accuracy: 60.3%, Validation Loss: 0.265

Training Epoch 16—Training Accuracy: 90.7%, Validation Accuracy: 60.6%, Validation Loss: 0.365

Training Epoch 19—Training Accuracy: 90.7%, Validation Accuracy: 60.8%, Validation Loss: 0.374

Training Epoch 19—Training Accuracy: 91.7%, Validation Accuracy: 60.8%, Validation Loss: 0.374

Training Epoch 29—Training Accuracy: 91.7%, Validation Accuracy: 60.8%, Validation Loss: 0.374

Training Epoch 29—Training Accuracy: 91.7%, Validation Accuracy: 60.6%, Validation Loss: 0.382

Training Epoch 29—Training Accuracy: 91.7%, Validation Accuracy: 60.8%, Validation Loss: 0.382

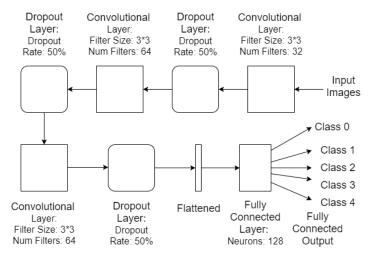
Training Epoch 29—Training Accuracy: 91.7%, Validation Accuracy: 60.6%, Validation Loss: 0.382

Training Epoch 29—Training Accuracy:
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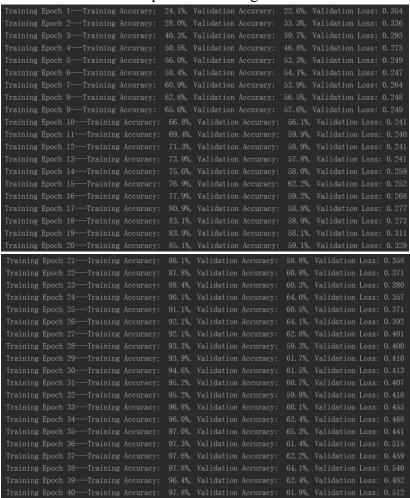
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Training Epoch 25—Training Accuracy: 97. 2%, Validation Accuracy: 58. 0%, Validation Loss: 0. 444
Training Epoch 26—Training Accuracy: 97. 6%, Validation Accuracy: 60. 5%, Validation Loss: 0. 415
Training Epoch 28—Training Accuracy: 98. 1%, Validation Accuracy: 59. 5%, Validation Loss: 0. 509
Training Epoch 29—Training Accuracy: 97. 4%, Validation Accuracy: 58. 9%, Validation Loss: 0. 509
Training Epoch 30—Training Accuracy: 97. 6%, Validation Accuracy: 58. 9%, Validation Loss: 0. 509
Training Epoch 31—Training Accuracy: 98. 6%, Validation Accuracy: 58. 9%, Validation Loss: 0. 559
Training Epoch 32—Training Accuracy: 98. 6%, Validation Accuracy: 68. 9%, Validation Loss: 0. 525
Training Epoch 33—Training Accuracy: 98. 6%, Validation Accuracy: 61. 7%, Validation Loss: 0. 527
Training Epoch 34—Training Accuracy: 98. 4%, Validation Accuracy: 61. 7%, Validation Loss: 0. 551
Training Epoch 35—Training Accuracy: 98. 2%, Validation Accuracy: 61. 7%, Validation Loss: 0. 551
Training Epoch 36—Training Accuracy: 98. 2%, Validation Accuracy: 61. 0%, Validation Loss: 0. 528
Training Epoch 37—Training Accuracy: 98. 5%, Validation Accuracy: 61. 0%, Validation Loss: 0. 528
Training Epoch 39—Training Accuracy: 98. 5%, Validation Accuracy: 61. 0%, Validation Loss: 0. 528
Training Epoch 39—Training Accuracy: 98. 5%, Validation Accuracy: 61. 0%, Validation Loss: 0. 528
Training Epoch 40—Training Accuracy: 98. 5%, Validation Accuracy: 62. 9%, Validation Loss: 0. 659
Training Epoch 41—Training Accuracy: 99. 8%, Validation Accuracy: 61. 2%, Validation Loss: 0. 635
Training Epoch 42—Training Accuracy: 99. 2%, Validation Accuracy: 61. 2%, Validation Loss: 0. 634
Training Epoch 44—Training Accuracy: 99. 2%, Validation Accuracy: 62. 5%, Validation Loss: 0. 634
Training Epoch 44—Training Accuracy: 99. 2%, Validation Accuracy: 62. 5%, Validation Loss: 0. 604
Training Epoch 45—Training Accuracy: 99. 2%, Validation Accuracy: 62. 6%, Validation Loss: 0. 604
Training Epoch 46—Training Accuracy: 99. 3%, Validation Accuracy: 62. 6%, Validation L
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2.3

In the end, I tried to reduce the dropout rate to 50% which is the so-called the best choice for CNN. From previous CNN model, 40 epochs are enough for training. Meanwhile, I set more filters in some convolutional layer.



The overfitting problem is further solved in this CNN. However, the validation accuracy is still around 62%. It seems that 62% is the maximum accuracy for CNN models in this project. Thus, I choose to use this CNN model to predict test images.



3. Predict test data

In "project2_predict_20459996.py", I load test images and predict them one-by-one. After predicting, I write the predicted labels into a file: "project2 20459996.txt".