Introduction to Computer Vision (ECSE 415) Assignment 3

Due: November 2^{nd} , 11:59PM

Please submit your assignment solutions electronically via the myCourses assignment dropbox. Attempt all parts of this assignment. The assignment will be graded out of total of **33 points**. Students are expected to write their own code. (Academic integrity guidelines can be found at https://www.mcgill.ca/students/srr/academicrights/integrity). Assignments received up to 24 hours late will be penalized by 30%. Assignments received more than 24 hours late will not be graded.

Submission Instructions

- 1. Prepare and submit two separate Google Colab notebooks for the two questions.
- 2. Comment your code appropriately.
- 3. Do not submit input/output images. Assume image folders are kept in a same directory as the codes.
- 4. Make sure that the submitted code is running without error. Add a README file if required.
- 5. Answers to reasoning questions should be comprehensive but concise.
- 6. Submissions that do not follow the format will be penalized 10%.

Note that you can use any of the OpenCV, sklearn, skimage and Pytorch functions shown during the tutorial sessions.

1 Image Classification using RF and SVM

For this task, you are given a dataset of flower images¹. The dataset contains images of 9 types of flowers. You can read the images and the corresponding labels as follows.

 $^{^1{}m The\ dataset}$ is derived from the 102-Category Flower dataset[1].

```
train_images = np.load('flower_subset.npz')['train_images']
train_labels = np.load('flower_subset.npz')['train_labels']
test_images = np.load('flower_subset.npz')['test_images']
test_labels = np.load('flower_subset.npz')['test_labels']
```

The arrays train_images and test_images are stacks of 1556 and 90 gray-scale images of size 128×128, respectively.

- Resize the train/test images to 64×64 and compute HoG features using cells of 8×8 pixels, blocks of 4×4 cells and 4 bins. This should yield a feature vector of size 1600 per image. (3 points)

 (Suggestion: Make a function which takes list of images as arguments and delivers list of HoG features as output. The same function can be used for train and test set.)
- Fit a non-linear SVM classifier (use RBF kernel with gamma='auto' and C=1) on the features and the class labels of the training images. (1 points)
- Predict labels of the test images by feeding the test features to the trained classifier and calculate classification accuracy. (2 points)
- Tune values of hyperparameters 'gamma' and 'C' to achieve test accuracy greater than 50%. (2 points)
- Fit a Random Forest(RF) classifier (set n_estimators=10, max_depth=5 and criterion='entropy') on the features and the class labels of the training images. (1 points)
- Predict labels of the test images by feeding the test features to the trained classifier and calculate classification accuracy. (2 points)
- Tune values of hyperparameters 'n_estimators' and 'max_depth' to achieve test accuracy greater than 50%. (2 points)
- Compare results of SVM and RF classifiers. Which one provides better results? Experiment training both classifiers with a range of random states (try different random values for the argument 'random_state') and measure classification accuracy of the test set. Which classifier is more stable or robust to the change in random state? (3 points)

2 Image Classification with Convolution Neural Network (CNN).

In this part, you will classify MNIST digits [2] into 10 categories using a CNN. You may chose to run the code on GPU.

- 1. Use Pytorch class torchvision.datasets.MNIST to (down)load the dataset. Use batch size of 32. (3 points)
- 2. Implement a CNN with the layers mentioned below. (5 points)
 - A convolution layer with 32 kernels of size 3×3 .
 - A ReLU activation.
 - A convolution layer with 64 kernels of size 3×3 .
 - A ReLU activation.
 - A maxpool layer with kernels of size 2×2 .
 - A convolution layer with 64 kernels of size 3×3 .
 - A ReLU activation.
 - A convolution layer with 64 kernels of size 3×3 .
 - A ReLU activation.
 - A flattening layer. (This layer resizes 2D feature map to a feature vector. The length of this feature vector should be 4096.)
 - A Linear layer with output size of 10.

(Suggestion: you can start with the code from Tutorial 6 and adapt it for the current problem.)

- 3. Create an instance of SGD optimizer with learning rate of 0.001. Use the default setting for rest of the hyperparameters. Create an instance of categorical cross entropy criterion. (1 point)
- 4. Train the CNN for 10 epochs. (5 points)
- 5. Predicts labels of the test images using the above trained CNN. Measure and display classification accuracy. (3 points)

References

- [1] Nilsback, Maria-Elena, and Andrew Zisserman. "Automated flower classification over a large number of classes." 2008 Sixth Indian Conference on Computer Vision, Graphics & Image Processing. IEEE, 2008.
- [2] Y. LeCun, L. Bottou, Y. Bengio, and P. Haffner. "Gradient-based learning applied to document recognition." Proceedings of the IEEE, 86(11):2278-2324, November 1998.