# 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 images<sup>1</sup>. The dataset contains images of 9 types of flowers. You can read the images and the corresponding labels as follows.

 $<sup>^1{</sup>m The\ dataset}$  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 25%. (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 25%. (2 points)
- Compare results of SVM and RF classifiers. Which one provides better results? Experiment training both classifiers with a range of random stats 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  $3\times3$ .
  - A ReLU activation.
  - A convolution layer with 64 kernels of size 3×3.
  - A ReLU activation.
  - A maxpool layer with kernels of size  $2\times 2$ .
  - A convolution layer with 64 kernels of size  $3\times3$ .
  - A ReLU activation.
  - A convolution layer with 64 kernels of size  $3\times3$ .
  - 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.