# Final Project for ECE 662 Deep Learning with Python Due Thursday December 14, 2023 by 4:00 p.m.

**Problem Statement** State the problem you are going to work on.

- 1. What is the dataset you are using? (CIFAR100, GoogleDraw, etc.)
- 2. Present any background information of your proposed problem (classification accuracies given in the literature, etc.).
- 3. Present and propose any new ideas/approaches to the problem that you might consider in your work.

## Progress Report Due Thursday November 30, 2023.

- 1. A progress report of 1-2 pages should (at least) include answering the problem statement questions above.
- 2. Note that you are to use PyTorch. You can, of course, get ideas from the internet, but be sure to properly **cite** them.
- 3. (Not required) Give any preliminary accuracy results you might have.

In addition to the Problem Statement, the **Final Report** should be *double-spaced single-column* and contain the following information:

**Dataset** Describe the dataset you choose.

- 1. What is the dataset and how big is the one you chose?
- 2. From your dataset, what is the size of the training\_data, validation\_data, and the test\_data.

Network Architectures Give the network architectures you considered for your problem.

- 1. Number of inputs.
- 2. For each Convolutional layer give the dimensions of the filter (mask/window) for that layer and the number and dimensions of the maps.
- 3. For each Pooling layer give how much you down sampled and the dimensions of the resulting pooled maps.
- 4. For each Fully Connected (FC) layer give the number of hidden neurons.
- 5. Describe your output layer and the cost function used (squared error, cross-entropy, maximum likelihood, etc.)
- 6. Give the values of the hyper-parameters used (mini-batch size,  $\eta$ ,  $p_{\text{dropout}}$ , etc.)
- 7. Give the accuracy on the validation\_data and the test\_data.

Setting up a Table to summarize your results is an effective way to present the various different architectures and sets of hyper-parameters values you considered.

Based on the classification accuracy on the validation\_data, explain your final choice of network architecture and hyper-parameter values.

**Results** Give the full results of your final design.

- 1. What was the performance on the test data?
- 2. How do your results compare to the results given in the literature?

#### References

- 1. Give the references of where you got your information. For example, Michael Nielsen's book. If you got something online, then give the source including the link.
- 2. Give proper citation of any ideas you got from the internet for your project.

#### **General Comments**

The proper way to cite work is to put the complete reference at the end of the report. It is okay to put in a hyperlink if you wish.

Do a formal report in the proper context. In the case the proper context is the terminology and concepts covered in the course. All other new ideas, terminology, etc. should have a brief explanation.

Don't use terms like "best possible accuracy" or "optimal results". The "best accuracy" is 100%. Further, "optimal results" implies that the accuracy you got is the best possible and no one else could get better.

#### Handin

- 1. Your final report.
- 2. The files (.ipynb, .py) needed to run your program (not the data). That is, if I don't understand something in your report I want to be able to perhaps understand it from your code.

### Academic Honesty

You are expected to do your own work. Any instance of plagiarism will result in a grade of zero on the project.

#### **Grading Scheme**

Progress Report (1-2 pages)	0%
Design Choices and Justification	33%
Performance	33%
Report Presentation, Clarity, English Usage, etc.	33%

## Existing Accuracy Results for the MNIST, CIFAR, SVHN datasets

 $https://rodrigob.github.io/are\_we\_there\_yet/build/classification\_datasets\_results.html \#43494641522d3130$ 

## Dataset for CIFAR:

https://www.cs.toronto.edu/~kriz/cifar.html~ If you use CIFAR then grab at least 20 categories from the CIFAR100.

## Dataset for QuickDraw

https://quickdraw.withgoogle.com/data

## Dataset for SVHN

http://ufldl.stanford.edu/housenumbers/