

APL 745: Deep Learning for Mechanics

Lab – 2

To be submitted by 3rd March midnight

The objective is to develop (step by step) a convolutional neural network to perform classification problem.

Load the following packages for this problem:

- PyTorch is a deep learning package. (other deep learning packages can also be used).
- Numpy is the fundamental package for scientific computing with Python.
- H5py is a common package to interact with a dataset that is stored on an H5 file
- Matplotlib is a famous library to plot graphs in Python.
- Sklearn is a commonly used Python package for basic data pre-processing.

Step 1: Load the data (problem1.hdf5) using H5py and Numpy. The dataset (MNIST dataset) consists of 28×28 pixel grayscale images of handwritten single digits from 0 to 9 with the corresponding true labels.

Step 2: Define the following model structure with different Python files as follows:

- Model-configuration – 1: Build 2 convolutional and 2 linear layers as follows:

Input \rightarrow Conv1 (with output channels = 16) \rightarrow Conv2 (with output channels = 32) \rightarrow Maxpool \rightarrow Flatten \rightarrow linear1 (512 hidden units) \rightarrow linear 2 (128 hidden units) \rightarrow output.

- Model-configuration – 2: Build 2 convolutional and 2 linear layers as follows with ReLU activation and dropout as follows:

Input \rightarrow Conv1 (with output channels = 16) \rightarrow ReLU \rightarrow Conv2 (with output channels = 32) \rightarrow ReLU \rightarrow Maxpool \rightarrow dropout (0.5) \rightarrow Flatten \rightarrow linear1 (512 hidden units) \rightarrow ReLU \rightarrow dropout (0.5) \rightarrow linear 2 (128 hidden units) \rightarrow output.

- For all conv layers, use kernel of size 5x5 with no padding and stride = 1
- For all max pool layers, use window of size 2x2 with no padding and stride = 2
- For all the above configurations, use a suitable output activation function for this classification problem.

Step 3: Train the NN model using ADAM optimizer.

Step 4: Test the NN model using the test data that was unseen during training. Make sure that the gradient calculation is disabled during this inference stage

Write a computer code to answer the following questions related to Step-1:

- (A) Split the dataset into two parts as training data and testing data with the ratio 70 : 30.
- Print the input and the output shape of the training and the testing dataset.
 - Plot first 5 test images and provide the true label as the caption for each test image.
- (B) Create an iterable DataLoaders that we can feed into a neural network for training and testing. Use batch size as 64.

Write a computer code to answer the following questions related to Step-2 and -3:

- (C) Use a suitable loss function for training the NN model.
- (D) Train all the above model-configurations with the following learning rate and weight decay training-configurations:
- Training-configuration-1: learning rate: 5×10^{-2} , weight decay: 1×10^{-3}
 - Training-configuration-2: learning rate: 1×10^{-4} , weight decay: 1×10^{-5}
- (E) Plot the following separately for the training data and test data for all the above configurations:
- Epoch vs Training loss
 - Epoch vs Testing loss
 - Epoch vs Accuracy
- (F) Plot the first 5 ground truth test data, the ground truth label, and the predicted labels as the caption.
- (G) Print the number of parameters for all the above model-configurations.
- (H) Compare the results obtained with those obtained in Lab-1. Choose (training config 1 + model config 2 + batch size 64) for CNN and (training config 1 + model config 4 + batch size 64) for FNN (from Lab 1).
- (I) Comment on your observations along with the computational time (wall clock time in seconds) for the above comparison.