Machine Learning for Engineering Applications Homework #5 Fall 2023

Due: 11/1/2023 by 11:59 PM – submit your zip file to CANVAS **Concrete Dataset** □ Dataset info and download: https://data.mendeley.com/datasets/5y9wdsg2zt/2 The CNN Network ☐ Develop your own *CNN* model to classify all classes. ☐ Provide the training and test confusion matrices. □ Provide the test accuracy, precision, recall, and F1-scores to a text file. ☐ Provide the Loss and Accuracy curves for training and validation (you can use a single plot for these four curves) ☐ Expected results: High 90's for training, validation, and testing without overfitting/underfitting. **Submit:** \square Python file(s) (.py) ☐ Confusion matrix image (.png or .jpg) \Box Curve image(s) (.png or .jpg) ☐ The text file with all the metrics ☐ The SLURM/.OUT files used for the LEAP cluster execution **Rubric per Part and Overall:** \Box Header in the code (5pts) □ Comments in the code (10pts) ☐ Running code without errors (20pts) ☐ Executes requirements & produces required output information (65pts)

Template:

```
#***********************************
# Damian Valles
# ML - HW#1
# Filename: hw1-perceptron.py
# Due: Sept. 6, 2023
# Objective:
# To demonstrate header and comment expectations for assignments for the semester.
#********************************
#Importing all required libraries
from sklearn import datasets
import numpy as np
from sklearn.model selection import train test split
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import Perceptron
#Downloading the dataset, creating dataframe
iris = datasets.load iris()
#Separating data & target information
X = iris.data[:, [2, 3]]
y = iris.target
#Data split for training and testing
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,
random state=1, stratify=y)
#Scaling training data
sc = StandardScaler()
sc.fit(X train)
X train std = sc.transform(X train)
#Creating perceptron with hyperparameters
ppn = Perceptron(max iter=40, eta0=0.01, shuffle=True)
#This is training the model
ppn.fit(X_train_std, y_train)
#Scaling test data
sc.fit(X test)
X test std = sc.transform(X test)
#Testing the model data
y_pred = ppn.predict(X_test_std)
```