EE P 596 - TinyML - Assignment 2

Total Points: 100
Spring Quarter, 2024
Department of Electrical and Computer Engineering
University of Washington, Seattle, WA 98195

Due: 11:59 pm (PST) on May 19 (Sun), 2024 via Canvas

Note:

- This homework contains both programming questions (marked as [Pro]) which are required to write
 Python codes and discussion questions (marked as [Dis]) which are required to write answers and
 provide detailed explanations for your answers.
- You can use and modify the Python functions and codes provided in the Labs or file section of the EEP 596 canvas page when coding the [Pro] questions.
- Your homework submissions must be uploaded to *Canvas* in the following formats: (i) a .ipynb file containing all Python code for the programming questions. If you have multiple .ipynb files, submit a .zip file with all the Jupyter notebooks. (ii) a .pdf file with either scanned handwritten or typed responses to discussion questions, and the final results for the programming questions. (iii) C++ files (.ino and .h files) which you compiled and uploaded to the Arduino Nano BLE.
- Name of your submission files should follow the following format: "#_\$_EEP596_HW2.ipynb" where "#" and "\$" should be replaced with your first name and last name, respectively. Use the same format for the .pdf writeup file.

In this assignment, you will make use of the following dataset and files. You will find these files when you unzip the **EE596_HW2_Files.zip**:

- Network Anomaly Dataset: This dataset consists of 125,973 data samples, each with 42 features corresponding to normal and attack network connections. The network_anomaly_data.txt file included with this homework assignment provides this dataset.
- FirstName_LastName_EEP596_HW2.ipynb: You will use this Python notebook file to answer Questions 2, 3, 4, and 5. When submitting your homework, please replace FirstName_LastName with your first and last name.
- c_writer.py: This script will be called in FirstName_LastName_EEP596_HW2.ipynb to automatically convert your tflite model to C and create the header file (network_model.h) for Arduino Nano BLE deployment.
- network_data.ino: You will use this Arduino Sketch file to answer Question 6. You will find this file inside the network_data folder.

- 1. [Pro] (Data Collection for TinyML Lab 6, "MagicWand," 10 points) The deadline was May 1st.
- 2. [Pro] (Data Preprocessing, 5 pts x 3 tasks = 15 pts)
 Complete the following tasks in the provided Python notebook (.ipynb) file:
 - (a) Drop the 'land', 'urgent', 'numfailedlogins', 'numoutboundcmds' columns from the dataframe data.
 - (b) Change any label that is not named **normal** to **attack** in the {'attack'} column of the dataframe data.
 - (c) Use **LabelEncoder()** function from the **sklearn.preprocessing** library to convert non-numerical attributes in the 'protocoltype', 'service', 'flag', 'attack' columns of the dataframe data to numerical values.
- 3. [Pro] (Dimensionality Reduction for Visualization, 5 pts x 3 tasks = 15 pts) Complete the following tasks in the provided Python notebook (.ipynb) file:
 - (a) Use **TSNE** from the **sklearn.manifold** library to visualize the data in the **test set** (**X_test**) in 2D. In your figure, use color **red** to mark **attack** data points and color **blue** to mark **normal** data points.
 - (b) Use **PCA** from the **sklearn.decomposition** library to visualize the data in the **test set** (**X_test**) in 2D. In your figure, use color **red** to mark **attack** data points and color **blue** to mark **normal** data points.
 - (c) Use **KernelPCA** from the **sklearn.decomposition** library to visualize the data in the **test set** (**X_test**) in 2D. Use **radial basis function** (**rbf**) as the kernel. In your figure, use color **red** to mark *attack* data points and color **blue** to mark *normal* data points.
- 4. [Pro] (Implementing a DNN on the dataset, 5 pts x 3 tasks = 15 pts)
 Complete the following tasks in the provided Python notebook (.ipynb) file:
 - (a) Implement a **deep neural network (DNN)** on the **Network Anomaly Dataset**. Ensure to include **two neurons** and **softmax activation** in the output layer of your DNN.
 - (b) Compile and train your DNN model on the **training set** (**X_train**). Denote the trained model as **base_model**.
 - (c) Evaluate the base_model on the test set (X_test) using classification_report and confusion_matrix from the sklearn.metrics library. Report these numbers in your .pdf writeup file using screenshots.
- 5. [Pro] (Implementing Quantized Model, 7.5 pts x 2 tasks = 15 pts)
 Complete the following tasks in the provided Python notebook (.ipynb) file:
 - (a) Implement **Dynamic Range Quantization** on the **base_model**. Designate the resulting quantized ML model as **tflite_quant_model**.
 - (b) Evaluate the **tflite_quant_model** on the **test set** (X_test) using **classification_report** and **confusion_matrix** from the **sklearn.metrics** library. **Report these numbers in your** .pdf writeup file using screenshots.

- 6. [Pro] (Deploying the Quantized Model, 5 pts x 4 tasks = 20 pts)
 Complete the following tasks in the provided Arduino sketch (.ino) file:
 - (a) Implement code to obtain the **prediction** from the **output tensor** and determine the **predicted** class label.
 - (b) Implement code to output **Sample #, Predicted Class**, and **Actual Class** for each sample to the serial monitor using **Serial.print** function.
 - (c) Add the **network_model.h** file, generated at the end of the Python notebook, to the **network_data folder** where **network_data.ino** is kept. Upload the **network_data.ino** to the Arduino Nano BLE. **Obtain screenshots of the printed Serial Monitor outputs for five samples and report these in your .pdf writeup.**
 - (d) Use the code at the end of the Python notebook to obtain 10 features and actual labels of the test set excluding the first five samples. Add these new data at the appropriate places in the network_data.ino. Upload the network_data.ino to the Arduino Nano BLE. Obtain screenshots of the printed Serial Monitor outputs for these ten samples and report these in your .pdf writeup.
- 7. [Dis] (Open-ended Discussion Questions, 5 pts x 2 tasks = 10 pts)
 - (a) In this homework, we hard code the test samples that need to be inferred using the TinyML model deployed on the Arduino Nano BLE. Is it possible to **stream network data** to the Arduino Nano BLE? If your answer is "YES," describe what **functionality within the Arduino Nano BLE** allows for this task.
 - (b) If we have high-dimensional nonlinear network data, which dimensionality reduction method among TSNE, PCA, and KernelPCA is most suitable to use and why? What parameters need to be passed to the Arduino Nano BLE to accomplish this?