National College of Ireland

Configuration Manual

MSc Research Project
Masters in Artificial Intelligence

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MSc Project Submission Sheet

School of Computing

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Programme:	Masters in Artificial Intelligence **Year:** 2023-2024
Module:	MSc Research Project
Lecturer:	Muslim Jameel Syed
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I hereby certify that the information contained in this (my submission) is information pertaining to research I conducted for this project. All information other than my own contribution will be fully referenced and listed in the relevant bibliography section at the rear of the project.
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Date: 14th December 2023

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Configuration Manual

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1 Introduction

Welcome to the Cybersecurity Fortification through Machine Learning project's configuration manual. This ground-breaking initiative aims to improve network environments' malware detection capabilities. This extensive manual is intended to assist users in setting up, configuring, and using the predictive models created to strengthen cybersecurity defences.

2 System Specification

- Supported, Metal GPUFamily Apple 7
- RAM; 16 GB
- SSD; 1TB + 500 Gb
- System Type: 64-bit Operating Systems
- Operating System: Apple MacBook Pro

3 Section 3

The Anaconda prompt and the Python programming language form the basis for putting this idea into practice. To guarantee accurate and neat results, the project has been integrated with the following libraries and packages:.

- NumPy
- Pandas
- Matplotlib
- Seaborn
- LightGBM
- Scikit-learn (KFold)
- Plotly
- Keras
- TensorFlow (Adam optimizer)

4 Steps for Configuration of Machine Learning:

1. To download and install Anaconda3 (Anconda, 2023)
2. extract the ‘ml_env.rar’ folder and paste it to the anaconda3’s envs folder
3. Extract the ‘HIDS_Fullcode 1’ folder
4. Run the Anaconda Navigator

5. Open in the anaconda3 prompt and in prompt used the command cd/d to change the directory to HIDS Fullcode 1.
6. (cd) Navigate to 'HIDS_Fullcode 1' folder.
7. Now run command: conda activate ml_env
8. Run command: python output_nsl_kdd.py
9. Run command : python output edge_iiot.py

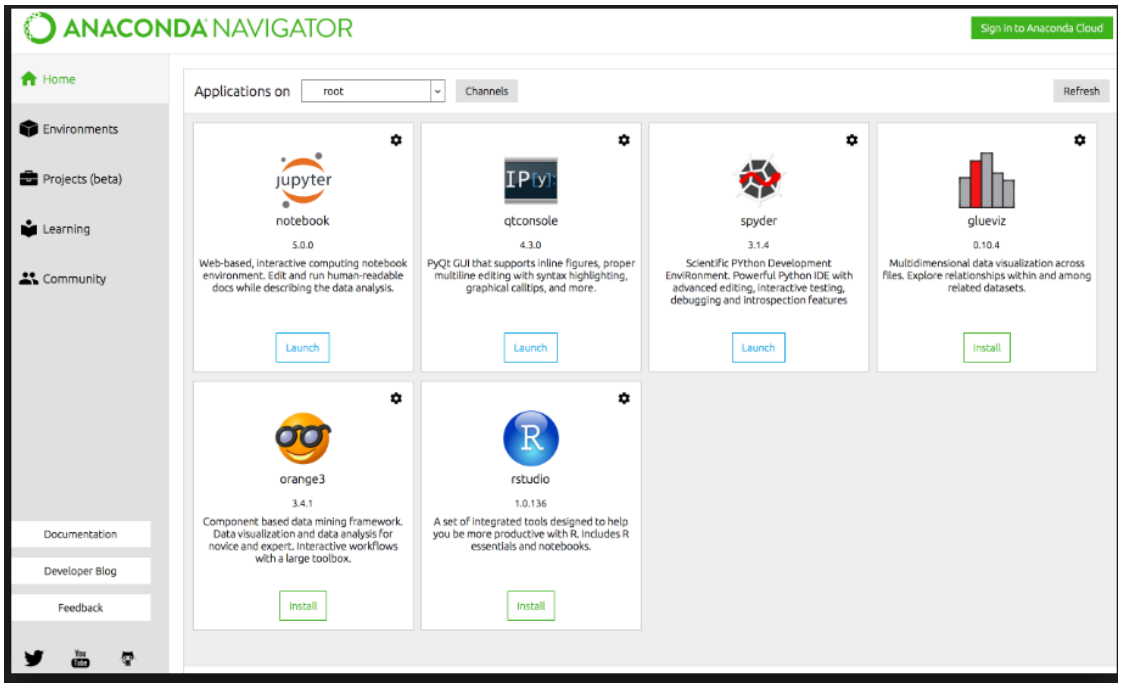


Figure 1: Anaconda Navigator

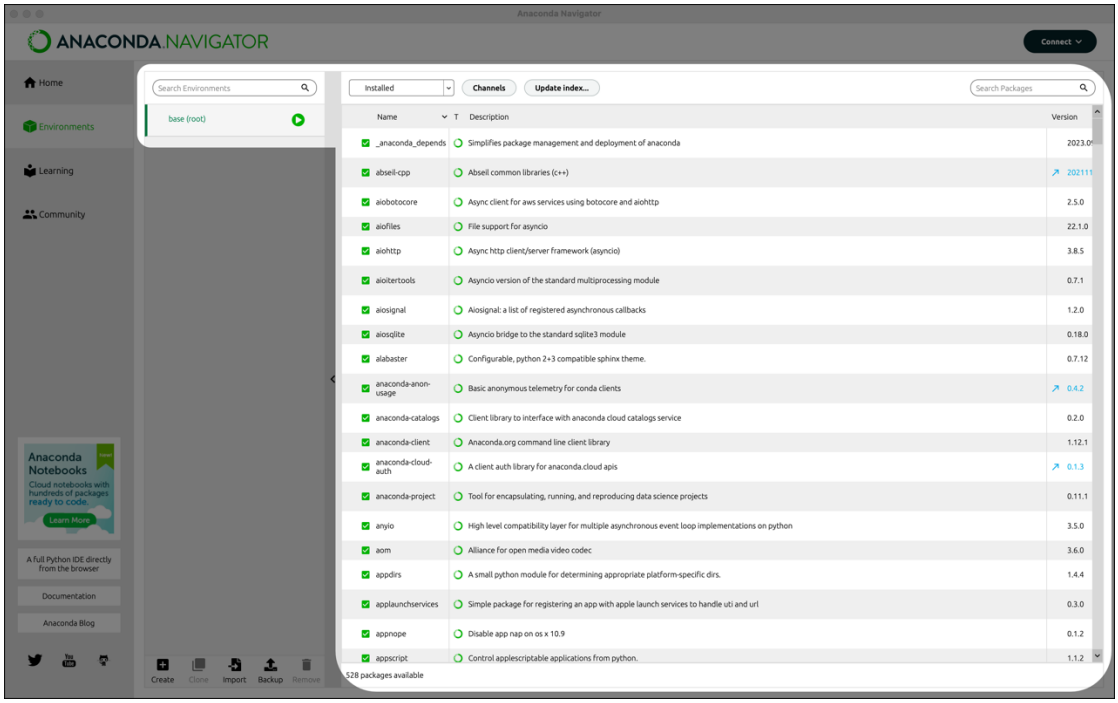


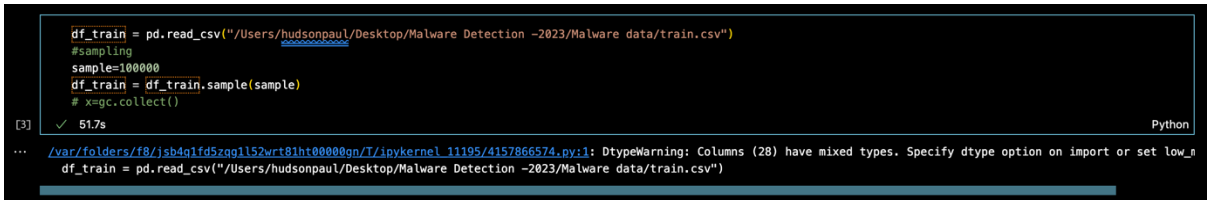
Figure 2: ml_env.rar'

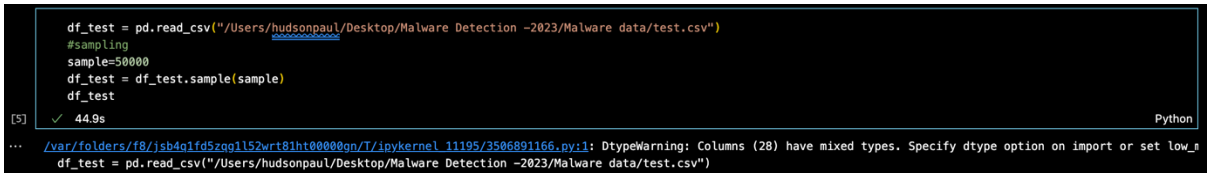
5 Steps for Configuration of Visual Studio Code

1. Download and Install Visual Studio Code:
2. Visit the official Visual Studio Code website (https://code.visualstudio.com/) and download the installer for your operating system.
3. Run the installer and follow the on-screen instructions to install Visual Studio Code on your machine.
4. Open Visual Studio Code:
5. Once the installation is complete, open Visual Studio Code.
6. Extensions and Plugins: Explore and install extensions and plugins based on your requirements. Common extensions include Python, Git, and various language support extensions.
7. Theme and Color Scheme: Choose a theme and color scheme that suits your preference. You can customize the appearance of Visual Studio Code through the "Preferences" menu.
8. Configure Settings: Adjust settings according to your preferences. You can access settings through the gear icon on the bottom left corner and selecting "Settings."
9. Integrated Terminal: Visual Studio Code comes with an integrated terminal. Familiarize yourself with its features, and customize the shell and appearance as needed.
10. Version Control Integration: If you're using version control (e.g., Git), integrate it with Visual Studio Code. Initialize a Git repository in your project folder and connect it to Visual Studio Code.
11. Workspace Setup: Set up your workspace by opening the desired project folder. You can customize your workspace settings to include specific folders and files.
12. Debugger Configuration: If you're working with a programming language that requires debugging, configure the debugger settings. Visual Studio Code supports various debugging configurations.
13. Explore Features: Take the time to explore additional features such as IntelliSense, code navigation, and integrated terminal functionalities.
14. Stay Updated: Regularly check for updates and new extensions to keep Visual Studio Code up to date with the latest features and improvements.

6 Procedure for Machine Learning

6.1 **Pre-Processing the data**





- Loading the Microsoft Malware dataset and removing the unwanted columns, duplicate rows, null values and removing other attack categories.
- Loading the Microsoft Malware dataset setting up the column plus data and creating a new csv file.

```python
#drop variables with missing values >=20% in the train dataframe
i=0
for col in df_train.columns:
    if (df_train[col].isnull().sum()/len(df_train[col])*100) >=10:
        print("Dropping column", col)
        df_train.drop(labels=col,axis=1,inplace=True)
        i=i+1

print("Total number of columns dropped in train dataframe", i)
```

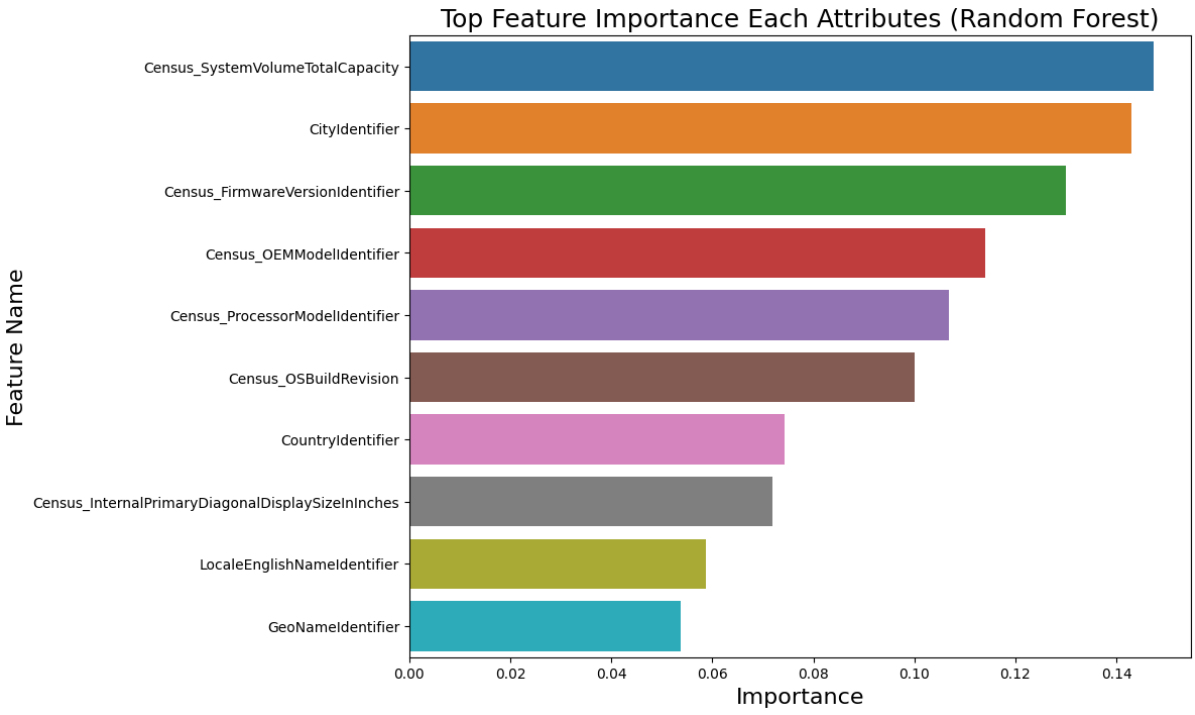
[10] ✓ 0.3s Python

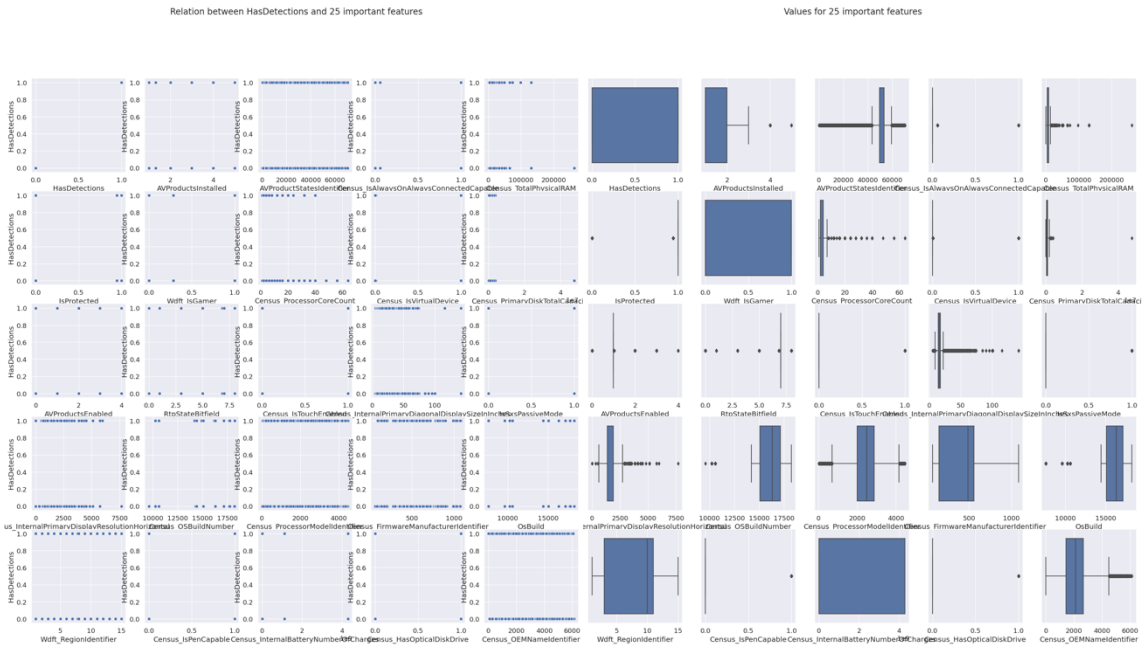
```
Dropping column DefaultBrowsersIdentifier
Dropping column OrganizationIdentifier
Dropping column PuaMode
Dropping column SmartScreen
Dropping column Census_ProcessorClass
Dropping column Census_InternalBatteryType
Dropping column Census_IsFlightingInternal
Dropping column Census_ThresholdOptIn
Dropping column Census_IsWIMBootEnabled
Total number of columns dropped in train dataframe 9
```

6.2 Feature selection

```python
imp_df = pd.DataFrame({
    "Feature Name": X_train.columns,
    "Importance": rfr.feature_importances_
})
fi = imp_df.sort_values(by="Importance", ascending=False)

fi2 = fi.head(15)
plt.figure(figsize=(10,8))
sns.barplot(data=fi2, x='Importance', y='Feature Name')
plt.title('Top Feature Importance Each Attributes (Random Forest)', fontsize=18)
plt.xlabel ('Importance', fontsize=16)
plt.ylabel ('Feature Name', fontsize=16)
plt.show()
```

[128] Python





6.3 Training and Testing of models

Model Traning

+ Code + Markdown

```
# from keras import callbacks
# from sklearn.metrics import roc_auc_score


class printAUC(callbacks.Callback):
    def __init__(self, X_train, y_train, validation_data):
        super(printAUC, self).__init__()
        self.bestAUC = 0
        self.X_train = X_train
        self.y_train = y_train
        self.validation_data = validation_data

    def on_epoch_end(self, epoch, logs={}):
        pred = self.model.predict(np.array(self.X_train))
        auc = roc_auc_score(self.y_train, pred)
        print("Train AUC: " + str(auc))
        pred = self.model.predict(self.validation_data[0])
        auc = roc_auc_score(self.validation_data[1], pred)
        print ("Validation AUC: " + str(auc))
        if (self.bestAUC < auc) :
            self.bestAUC = auc
            self.model.save("bestNet.h5", overwrite=True)
        return
```

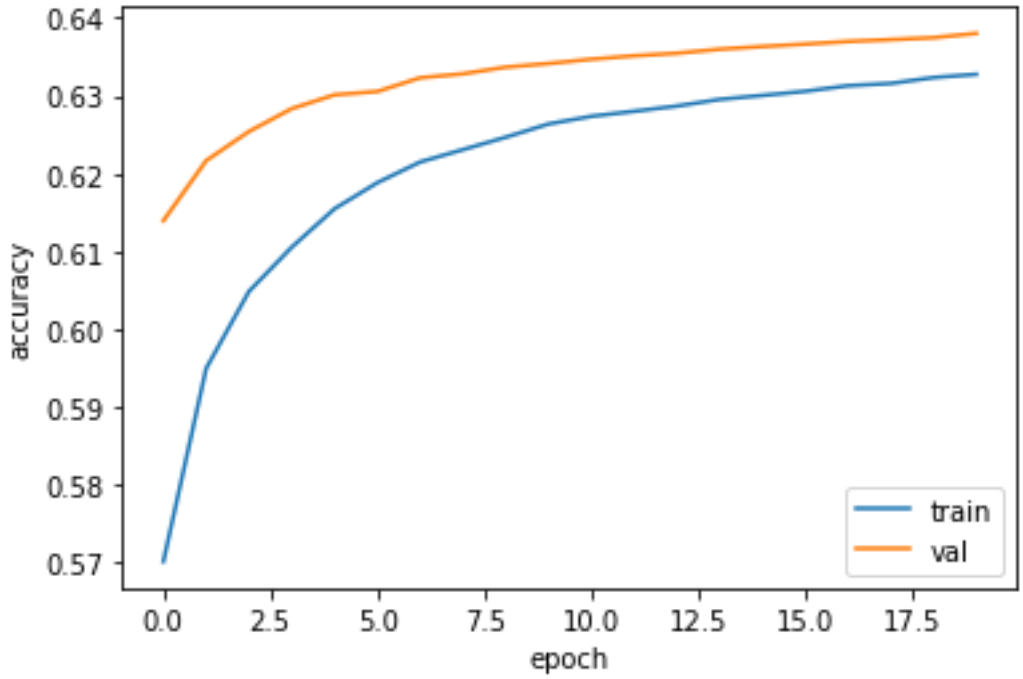
[85] Python

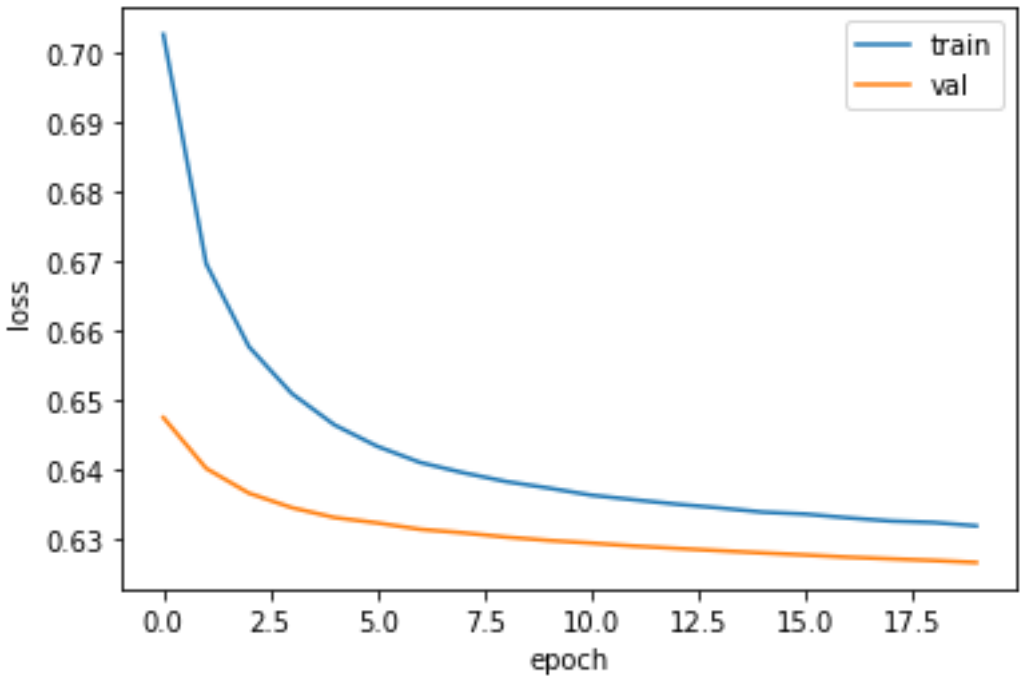
```
# BUILD MODEL
model = Sequential()
model.add(Dense(200,input_dim=len(X_train.columns)))
model.add(Dropout(0.4))
model.add(BatchNormalization())
model.add(Activation('relu'))

model.add(Dense(300))
model.add(Dropout(0.4))
model.add(BatchNormalization())
model.add(Activation('relu'))

model.add(Dense(400))
model.add(Dropout(0.4))
model.add(BatchNormalization())
model.add(Activation('relu'))

model.add(Dense(1, activation='sigmoid'))
model.compile(optimizer=Adam(learning_rate=0.01), loss="binary_crossentropy", metrics=["accuracy"])
annealer = LearningRateScheduler(lambda x: 1e-2 * 0.95 ** x)
```

[93] Python

```
[90]  X_train, X_val, y_train, y_val = train_test_split(X, y, test_size = 0.3,random_state=4)
                                                                                                  Python

[91]  X_train.shape, y_train.shape,X_val.shape
                                                                                                  Python
...   ((70000, 118), (70000,), (30000, 118))
```

6.4 Results





References

Anaconda, 2023. *Anaconda Navigator.* [Online]
Available at: https://www.anaconda.com/download
Studio, V., 2023. *Visual Studio Code.* [Online]
Available at: https://code.visualstudio.com/download