

Configuration Manual

MSc Research Project
MSc in Cloud Computing

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MSc Project Submission Sheet
School of Computing



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Programme: MSc in Cloud Computing
..... **Year:** 2023-2024
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Module: Research in Computing
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Lecturer: Shreyas Setlur Arun
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Submission Due Date: 16, September 2024
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Project Title: Assessing Machine learning Algorithms for Cloud Computing Threat Detection
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Word Count: 413
..... **Page Count:** 4
.....

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Zaid Ali Khan

Signature:
16, September 2024
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Configuration Manual

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1. Introduction

The study demonstrates the utilization of machine learning for threat detection. The study considers the usage of CICIDS 2017 Dataset for training CNN, RNN and Reinforcement learning model for prediction of threats.

2. Hardware Requirements

2.1 For macOS (High-Performance Mac Configuration)

- Model: Mac Pro or MacBook Pro
- Processor: Apple M1 Max or Intel Core i7/i9 processor
- Memory: 32 GB or 64 GB of RAM (depending on workload intensity)
- Storage: 2 TB SSD
- Graphics: Integrated Apple GPU or AMD Radeon Pro (for Mac Pro)
- Operating System: macOS Ventura or later
- Network: High-speed Ethernet or Wi-Fi 6 for data-intensive tasks

2.2 For Windows(High-Performance Windows Configuration)

- Model: Workstation PC (e.g., Dell Precision, HP Z Workstation) or high-end laptop (e.g., Dell)
- Processor: Intel Core i7/i9 or AMD Ryzen 9
- Memory: 32 GB or 64 GB of RAM
- Storage: 2 TB SSD
- Graphics: NVIDIA RTX 3060 or higher for workstation: Integrated Intel Iris Xe for laptops
- Operating System: Windows 11 Pro
- Network: Gigabit Ethernet or Wi-Fi 6

3. Software Requirements

3.1 Operating System:

- Windows 11 Pro or macOS Ventura or later

3.2 Programming Environment:

- **Python:** Version 3.8 or later
- **Anaconda:** For managing Python packages and environments
- **Jupyter Notebook:** For interactive development and documentation

3.3 Machine Learning Libraries:

- **TensorFlow:** For deep learning model development
- **Keras:** High-level neural networks API for TensorFlow
- **Scikit-learn:** For traditional machine learning algorithms
- **Pandas:** For data manipulation and analysis
- **NumPy:** For numerical computations
- **Matplotlib:** For plotting and visualization

3.4 Integrated Development Environment (IDE):

- **Visual Studio Code:** Lightweight editor with Python support

4. Dataset Used

The project makes use of the CICIDS2017 dataset provided by the Canadian Institute, for Cybersecurity which's a benchmark dataset specifically created for assessing network intrusion detection systems (NIDS). This dataset contains a range of network traffic data covering both activities and various types of cyber-attacks like DDoS and Brute Force making it well suited for evaluating threat detection models within cloud environments. The dataset, which is accessible in formats such as CSV and PCAP underwent pre-processing steps to prepare it for use in the project. These steps included handling missing data detecting outliers normalizing the data and extracting features to ensure that it was properly prepared for training machine learning algorithms aimed at enhancing security, in cloud environments.

5. Follow below steps for the code demonstration

Step 1: Open sage maker for the configured model. The graph has a –ve axis for better visualizations of the binary actions.

Step 2: Run sage maker notebook to save the model to s3 bucket.

```

In [87]: # Create a dictionary to encapsulate the environment, Q-table, and rewards
import boto3, joblib, os
reinforcement_model = {
    'environment': env,
    'q_table': q_table,
    'rewards_per_epoch': rewards_per_epoch
}

# Save the entire model to a file using joblib
joblib.dump(reinforcement_model, 'reinforcement_model.pkl')

# Initialize S3 client
s3 = boto3.client('s3')

# Define your bucket name
bucket_name = 'mybuc143'

# Function to upload a file to S3
def upload_to_s3(file_name, bucket, object_name=None):
    if object_name is None:
        object_name = os.path.basename(file_name)

    try:
        response = s3.upload_file(file_name, bucket, object_name)
    except Exception as e:
        print(f"Error uploading {file_name}: {e}")
    else:
        print(f"Successfully uploaded {file_name} to {bucket}/{object_name}")

# Upload the model to S3
upload_to_s3('reinforcement_model.pkl', bucket_name, 'models/reinforcement_model.pkl')

# Clean up local file if needed
os.remove('reinforcement_model.pkl')

Successfully uploaded reinforcement_model.pkl to mybuc143/models/reinforcement_model.pkl

```

Figure 1: Save Model to S3

Step 3: Run python app.py and browse to local host <http://localhost:5000> for the frontend

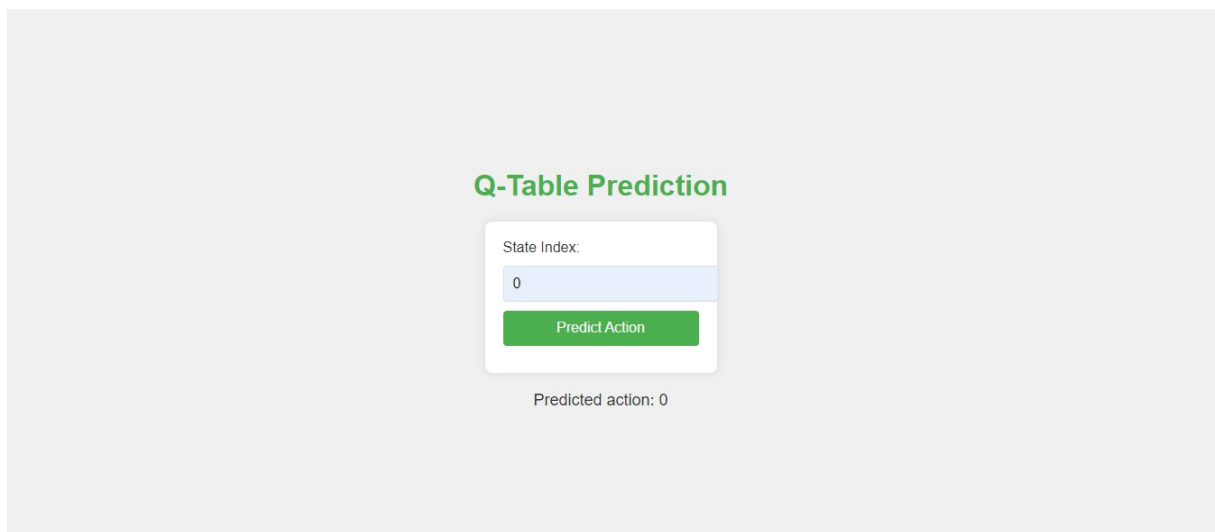
Q-Table Prediction

State Index:

Predict Action

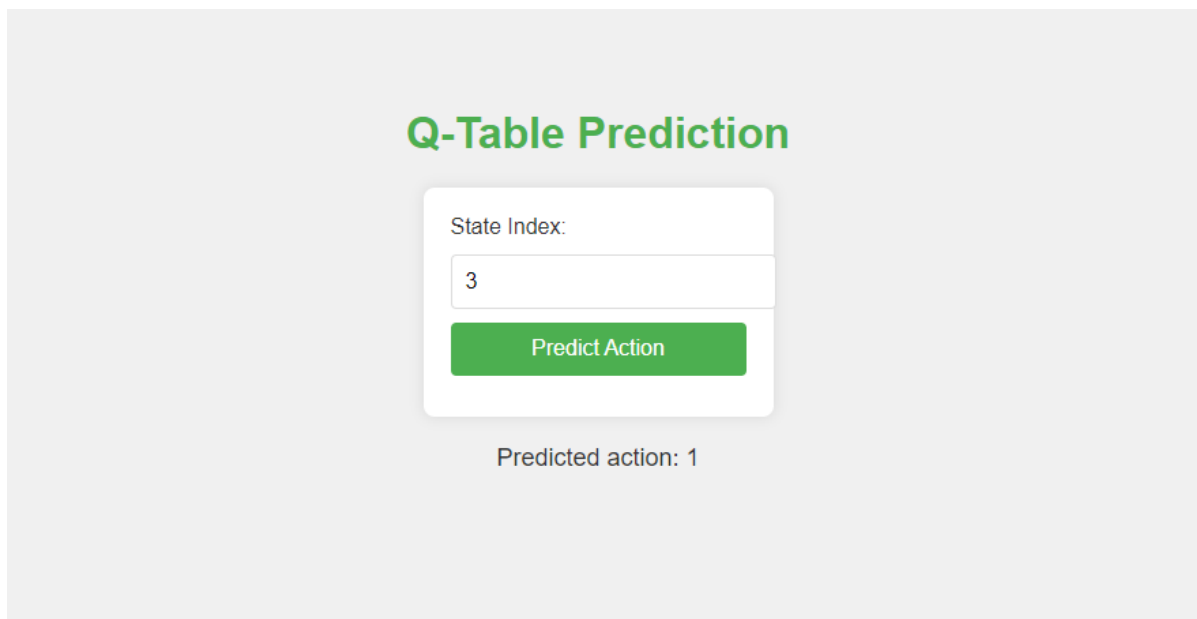
Figure 2: User-Interface of RL Model

Step 4: Enter values to predict Action.



The interface is titled "Q-Table Prediction" in green. It features a white rounded rectangle containing a "State Index:" label, a text input field with the value "0", and a green "Predict Action" button. Below the button, the text "Predicted action: 0" is displayed.

Figure 3: User-Interface of RL Model (Insert State)



The interface is titled "Q-Table Prediction" in green. It features a white rounded rectangle containing a "State Index:" label, a text input field with the value "3", and a green "Predict Action" button. Below the button, the text "Predicted action: 1" is displayed.

Figure 4: User-Interface of RL Model (Predict Action)