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Wi-Fi Network Security Using AI Bot

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ABSTRACT: In today's interconnected world, the security of wireless networks is of paramount importance. This paper presents a novel approach to bolstering Wi-Fi network security by leveraging advanced technologies, specifically deep reinforcement learning, to create an AI-powered bot capable of adapting to its environment. Implemented on a Raspberry Pi Zero W, the proposed system aims to enhance security by efficiently collecting diverse handshake data for analysis and prediction of potential cyber threats. By integrating cutting-edge technologies and adopting a proactive approach to security, the proposed model lays the foundation for robust Wi-Fi network protection.

KEYWORDS: IoT, Wi-Fi security, AI bot, deep reinforcement learning, Raspberry Pi, handshake data, cyber threats.

I. INTRODUCTION

In an era dominated by wireless connectivity, ensuring the security of Wi-Fi networks has become increasingly challenging. With the proliferation of IoT devices and the rise of sophisticated cyber threats, traditional security measures are often inadequate. This paper presents a novel approach to addressing this challenge by harnessing the power of artificial intelligence (AI) and IoT technologies to fortify Wi-Fi network defenses.

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II. LITERATURE REVIEW

Research in Wi-Fi network security encompasses various approaches such as encryption protocols like WPA2 and WPA3, intrusion detection systems (IDS), anomaly detection algorithms, wireless intrusion prevention systems (WIPS), software-defined networking (SDN), and blockchain technology. WPA3 strengthens encryption, while IDS and anomaly detection algorithms monitor and identify suspicious activities. WIPS actively prevents unauthorized access, with ongoing research focusing on integrating machine learning for improved threat detection. SDN centralizes network management for dynamic control and fine-grained access policies, while blockchain technology offers tamper-resistant records and decentralized authentication methods. These advancements collectively aim to enhance the security posture of Wi-Fi networks against evolving cyber.

EXISTING METHOD:

The existing method entails the creation of a sophisticated Smart Bot on a Raspberry Pi Zero W, equipped with the capability for adaptive security measures. The bot undergoes a learning process by assimilating passive and active responses

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from its environment, thereby gaining insights into potential threats. The primary focus lies in the efficient collection of diverse handshake data essential for analysis and optimization.

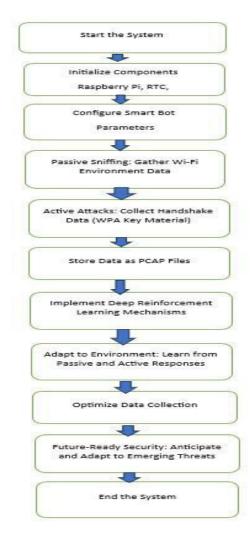
The bot utilizes passive sniffing techniques to gather information about the Wi-Fi environment. Through active attacks, it strategically acquires crackable WPA key material, encompassing various types supported by hash cat. The collected data is stored as PCAP files, ensuring a comprehensive and diverse range of handshake data.

The integration of cutting-edge technology involves the incorporation of deep learning mechanisms, specifically deep reinforcement learning, into the bot's architecture. This infusion of advanced technology enhances the bot's ability to adapt to evolving threats and challenges in real-time.

The proposed model is future-ready, aiming to anticipate and counteract emerging cyber threats. By combining adaptive learning, efficient data collection, and cutting-edge technology, the project envisions establishing robust Wi-Fi network security. The ultimate goal is to provide a sense of security and peace of mind to both end-users and network administrators, ensuring the continuous protection of Wi-Fi networks against evolving cyber threats.

III. PROPOSED METHOD

The proposed model aims to enhance Wi-Fi network security by leveraging AI-powered reinforcement learning techniques. Implemented on a Raspberry Pi Zero W, the model involves the creation of an intelligent bot capable of adapting to its environment through passive sniffing and active attacks. The bot's primary objective is to collect diverse handshake data, including various types supported by hash cat, stored as PCAP files. By optimizing the acquisition of crackable WPA key material and continuously learning from its interactions with the network, the bot seeks to anticipate and mitigate potential cyber threats in real-time. The integration of cutting-edge technologies such as deep learning and reinforcement learning enables the model to adapt to evolving threats, laying the foundation for robust and future-ready Wi-Fi network security.



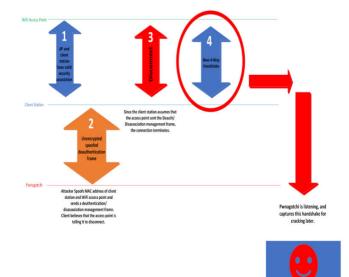
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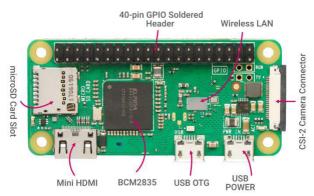
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BLOCK DIAGRAM:



1. RASPBERRY PI ZERO W:



Raspberry Pi Zero W is a miniature single-board computer equipped with wireless connectivity capabilities, including Wi-Fi and Bluetooth. With headers soldered, it facilitates easy connection to other components.

2. DS3231 RTC MODULE:



The DS3231 is a highly accurate real-time clock (RTC) module that provides precise timekeeping functionality for the Raspberry Pi. It ensures accurate timestamps for network activities and scheduling of security tasks.

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3. WAVE SHARE 2.13INCH E-PAPER HAT V2:



The Wave Share e-Paper HAT is a display module featuring an electronic paper (e-ink) display. It provides a low-power and easy-to-read display for visualizing system status, network statistics, and other relevant information.

4. SANDISK 32GB ULTRA MICROSD UHS-I MEMORY CARD:



The SanDisk microSD memory card provides ample storage capacity for storing system files, software, and collected data. Its high-speed UHS-I interface ensures efficient data transfer, crucial for handling network traffic and storing PCAP files.

5. HEADER STRIP:



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The pins are typically spaced at a standard interval (e.g., 2.54mm or 0.1 inch) and can be male (soldered to the board) or female (sockets that accept other connectors). They provide a way to easily connect electronic components like microcontrollers, sensors, and displays to a circuit board.

IMPLEMENTATION:

Implementation of the IoT-based Wi-Fi network security system involves several steps, including hardware setup, training the AI-powered reinforcement learning bot, establishing connections, and using the system in real-time operations.

1. HARDWARE SETUP:

Begin by assembling the hardware components, including the Raspberry Pi Zero W, RTC module, UPS-Lite, e-Paper display, and 3D printed case. Connect the RTC module and UPS-Lite to the Raspberry Pi Zero W via GPIO pins, ensuring proper alignment and secure connections.

Install the e-Paper display on top of the Raspberry Pi Zero W, aligning the GPIO pins with the corresponding headers. Place the assembled components inside the 3D printed case, ensuring that all connections are secure and the components are properly housed.

2. TRAINING THE AI BOT:

Develop and train the AI-powered reinforcement learning bot using Python. Collect diverse handshake data by passively sniffing Wi-Fi traffic and actively performing attacks. Implement deep reinforcement learning algorithms to optimize the acquisition of crackable WPA key material.

Train the bot to adapt to its environment by learning from passive observations and active interactions with the Wi-Fi network.

3. ESTABLISHING CONNECTIONS:

Connect the Raspberry Pi Zero W to the local Wi-Fi network using the built-in Wi-Fi module. Ensure that the RTC module is properly synchronized with the current time to timestamp network events accurately.

Establish communication between the Raspberry Pi Zero W and the e-paper display to visualize system status and network statistics.

4. REAL-TIME OPERATIONS:

Deploy the IoT-based Wi-Fi network security system in a real-world environment. Monitor network traffic and security events in real-time using the AI-powered reinforcement learning bot.

Collect handshake data and analyze network traffic to identify potential cyber threats. Adapt security measures based on the current network environment and anticipated threats.

Display system status, security alerts, and network statistics on the e-Paper display for real-time monitoring

WORKING:

ESAPPS starts by collecting real-time data from the onboard sensors, including vehicle speed, acceleration, steering angle, and environmental conditions.

The collected data is processed by the deployed machine learning models, which analyze the input features and predict the likelihood of potential accidents.

Based on the predictions, ESAPPS activates pre-alert mechanisms if hazardous conditions are detected. These mechanisms alert the driver and relevant authorities to take necessary precautions and preventive actions.

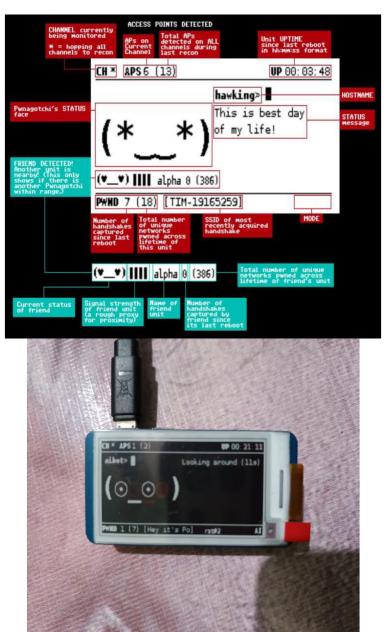
ESAPPS continuously updates its models using incoming data, ensuring adaptability to changing road conditions and driver behaviors.

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The above figure shows that the AI Bot is looking around the surroundings to capture networks that are available in the region.



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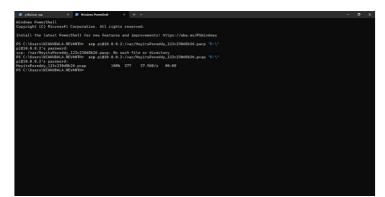
Here the figure describes that the bot does not make any new friends and does not get any handshakes later it will analyze the network to make new friends and captures the handshakes between the network.



Cool, the bot makes a new handshake and stores the handshake in its memory for cracking the password.



Here we are extracting the handshakes that are captured by the bot by using the command prompt in the form of PCAP files.



Hence the PCAP file is converted to the hc22000 file for the decoding of the password.

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Nicrosoft Windows (Version 18.8.22631.3185) (c) Microsoft Corporation. All rights reserved.		
C:\Users\NIHKUDALA.REVANTH\Downloads\Dut\hashcat-6.2.6>hashcat.exe -m 22008 2114181_1708967851.hc22006 7d7d7d7d7d7d7d7d7d7d7d7d7d7d7d7d7d7d7d		
Started: Wed Mar 06 17:31:57 2024 Stopped: Wed Mar 06 17:31:57 2024		
C:\Users\NIKKUDALA.HEYANTH\Downloads\Out\hashcat-6.2.6>hashcat.exe -m 22000 2114101_17000067851.hc22000 -a 3 ?d?d?d?d?d hashcat (v6.2.6) starting		
OpenCL APT (OpenCL 3.8) - Platform #1 [Intel(R) Corporation]		
 Device #1: Intel(8) UHD Graphics, 1536/3194 HB (798 HB allocatable), 32HCU 		
Minimum password length supported by kernel: 8 Maximum password length supported by kernel: 63		
Hashes: 2 digests; 2 unique digests, 1 unique salts Bitmaps: 16 bits, 65536 entries, 0x0000ffff mask, 262144 bytes, 5/13 rotates		
0g tisizers applied: • Acrespte • Single-Sint • Sint • Sint		
Natohdog: Hardmare monitoring interface not found on your system. Matchdog: Temperature abort trigger disabled.		
Host memory required for this attack: 368 MB		
693;48bc5232297c60594r0069643588;72c381cc2532;986864722287;1qoo zó pro:12345078 A5#74444877453zeethaka7661757c3+277c181+e2532;185A65464471;rgon zó pro:12345678		
Sessioni hashcat		

The hc22000 file is decrypted by using hash cat software in the bot.

527580522275655167595535817521814252217506858722371508288 gav11208878 52759127752248848776459159221752218142232198058598977140828 gav11208878 5645912775224884877647552281752218142232198058598977140828 gav11208878 5645912752286 (dos-Supergraventintintintinti) 5645912752286 (dos-Supergraventintintintintinti) 564591275288 gav1120828 gav1120828 gav112088 56459127528 gav1120828 gav1120828 gav112088 56459127528 gav1120828 gav112088 gav112088 gav112088 56459127528 gav112088 gav112088 gav112088 56459127528 gav112088 gav112088 gav112088 56459127528 gav112088 gav112088 gav112088 56559128 gav11208 gav112088 56559128 gav112088 gav112088 56559128 gav112088 56559128 gav11208 56559128 gav11	
Becovered: 2/2 (100.00%) Digests (total), 2/2 (100.00%) Digests (new) Progress: 1638/1/00000000 (0.02%)	
Rejected	
Restore.Sub.#1: Salt:0 Amplifier:0-1 Iteration:1-3	
Candidate.Engine.: Device Generator Candidate.A: 1240676 >> 1966678	
Startod: Wed Mar 06 17:35:85 2024 Stopped: Wed Mar 06 17:36:37 2024	
C:\Users\NEH&UDALA.REVANTh\Downloads\Out\hashcat-6.2.6>hashcat.exe -m 22800 2114181_1708867851.hc22800 -a 3 rockyou.txt hashcat (v6.2.6) starting	
OpenCL API (OpenCL 3.0) - Platform #1 [Intel(R) Corporation]	
 Device #1: Intel(R) UHD Graphics, 1536/3194 MB (798 MB allocatable), 32MCU 	
Birismu passmord length supported by Wernel: 8 Maximum passmord length supported by Wernel: 63	
1NFO: All hashes found as potfile and/or empty entries! Useshow to display them.	
Started: Wed Mar 06 17:46:42 2024 Stopped: Wed Mar 06 17:46:47 2024	

Finally, the password of the network is cracked by the bot after checking with multiple combinations of passwords.

ADVANTAGES:

- Adaptive Security Measures
- Enhanced Detection Capabilities
- Improved Efficiency
- Real-time Monitoring
- Scalability

APPLICATIONS:

- Home Wi-Fi Security
- Enterprise Network Security
- Public Wi-Fi Hotspots
- IoT Device Security
- Critical Infrastructure Protection

IV. CONCLUSION

The IoT-based Wi-Fi network security system presents a cutting-edge solution for addressing the evolving challenges of wireless network security. By integrating advanced technologies such as deep reinforcement learning, the system offers adaptive security measures that continuously learn and adapt to the network environment. Through passive sniffing, active attacks, and real-time analysis of network traffic, the system effectively detects and mitigates potential cyber threats, providing enhanced protection for users and network administrators. With its scalability, efficiency, and real-time monitoring capabilities, the system represents a significant step towards ensuring robust and future-ready Wi-Fi network security in various applications and environments.

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