

# Projectworkphase1(19AI702)

## AI Powered Advanced Web Application Firewall

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2022-2026 Batch

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# Introduction

- **AI-Powered Advanced Web Application Firewall (WAF)** is a security system designed to protect web applications from cyber threats.
- With the rapid growth of web applications, attacks such as **SQL Injection, Cross-Site Scripting (XSS), and DDoS** have become increasingly common.
- Traditional WAFs rely on **static, rule-based mechanisms**, which are often ineffective against evolving or unknown attacks.
- **Artificial Intelligence (AI) and Machine Learning (ML)** enable intelligent traffic analysis and **real-time threat detection**.
- This project integrates AI/ML techniques to create an **adaptive and intelligent WAF** that improves security dynamically.

# Statement of the Problem

Develop an **AI-powered WAF** to protect web applications from cyber threats.

Use **machine learning models** to detect malicious web traffic.

Provide a **real-time monitoring dashboard** with alerts.

Improve **accuracy and efficiency** of threat detection.

**Target Audience:** Web developers, IT security teams, enterprises.

## **Deliverables:**

- Trained AI/ML model for traffic analysis.
- Functional WAF system with monitoring dashboard.
- Documentation of architecture, model, and performance.

# Scope of the project

## Focus:

- AI-powered Web Application Firewall (WAF)
- Real-time detection and blocking of cyber threats
- Dashboard for monitoring traffic and alerts

## Target Audience:

- Web developers & administrators
- Enterprises needing web security
- IT security professionals

## Deliverables:

- AI/ML models detecting SQL Injection, XSS, DDoS
- Functional WAF with monitoring & alerts
- Documentation & scalable deployment

# Scope of the project

## Inclusions:

- Collect and preprocess web traffic logs.
- Extract features from HTTP requests and patterns.
- Build AI/ML detection models.
- Develop a user-friendly dashboard.
- Implement real-time threat blocking and alerts.

## Exclusions:

- No hardware-level network security.
- Not for non-web or offline applications.
- No integration with third-party cloud security services.

# Scope of the project

- **Data Collection:** Use web traffic logs with labeled normal and malicious requests.
- **Preprocessing:** Clean data and extract relevant features for ML analysis.
- **Limitations:**
  - Dataset may not include all types of web attacks.
  - Real-time performance may depend on server capacity.
  - AI model accuracy may vary with unseen traffic patterns.
  - Continuous updates needed for evolving threats.

# Methodology

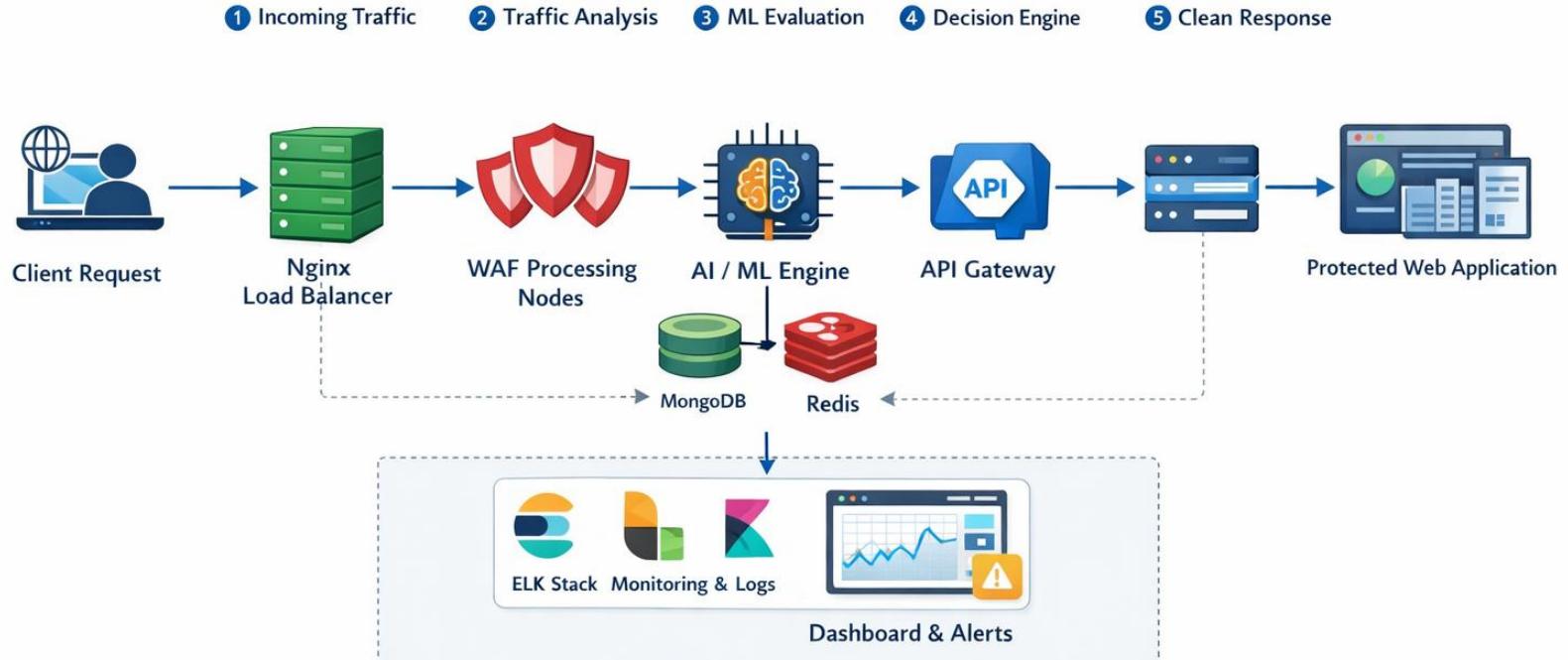
- **Data Collection:** Gather web traffic logs (normal and malicious requests).
- **Data Preprocessing:** Clean data and extract features from HTTP requests and headers.
- **AI/ML Model Development:** Train models to classify requests as safe or malicious.
- **Web Dashboard:** Build a user-friendly interface for monitoring traffic and alerts.
- **Integration & Testing:** Combine AI model with dashboard; validate accuracy and performance.

# Methodology

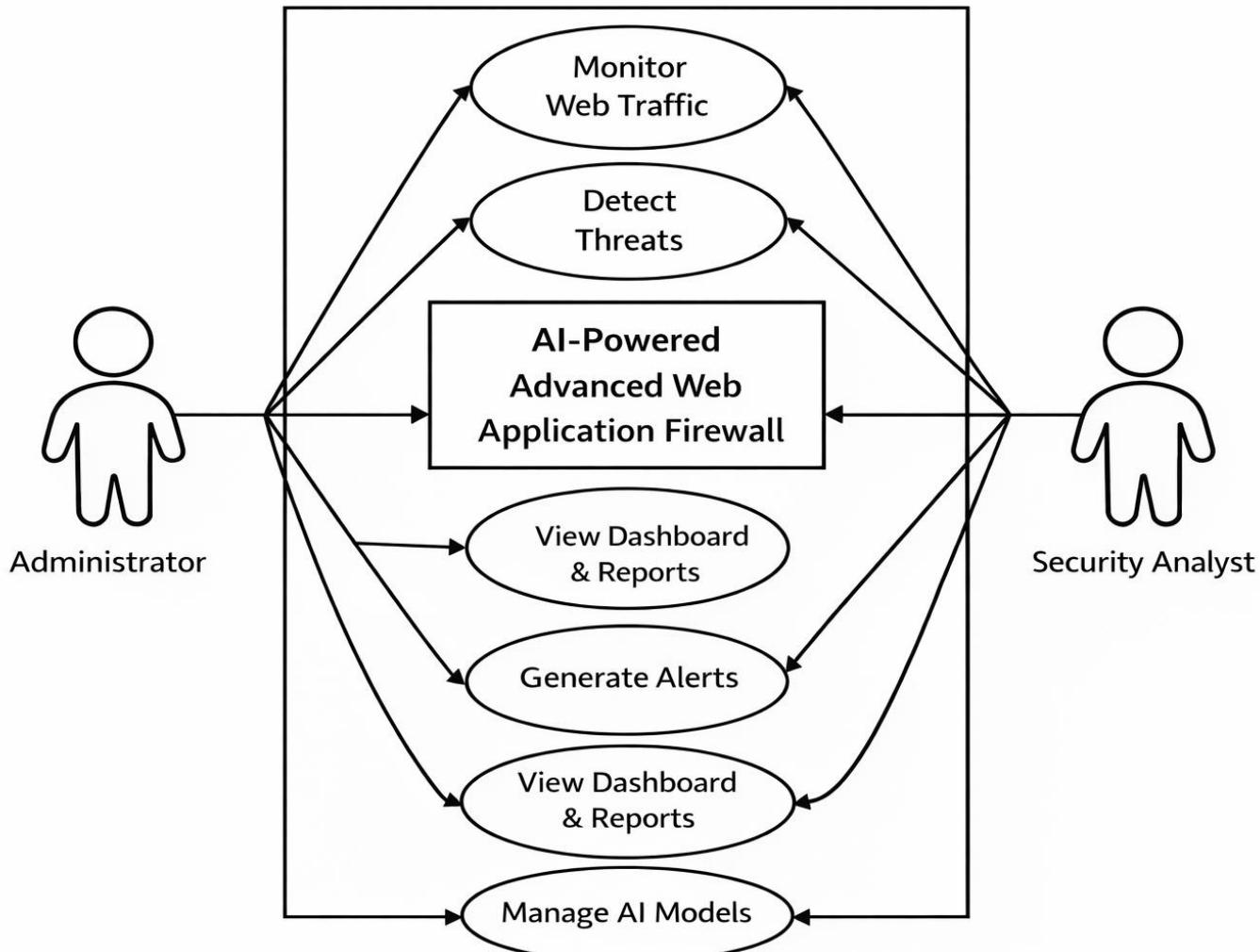
## Traffic Analysis & AI Integration:

- **Collect Traffic Data:** Gather normal and malicious web requests.
- **Preprocess Data:** Clean, normalize, and extract features from HTTP requests.
- **Develop AI/ML Model:** Train models to classify requests as safe or malicious.
- **Web Dashboard:** Create a user-friendly interface for monitoring and alerts.
- **Integration & Testing:** Combine AI model with dashboard; test for accuracy and performance.

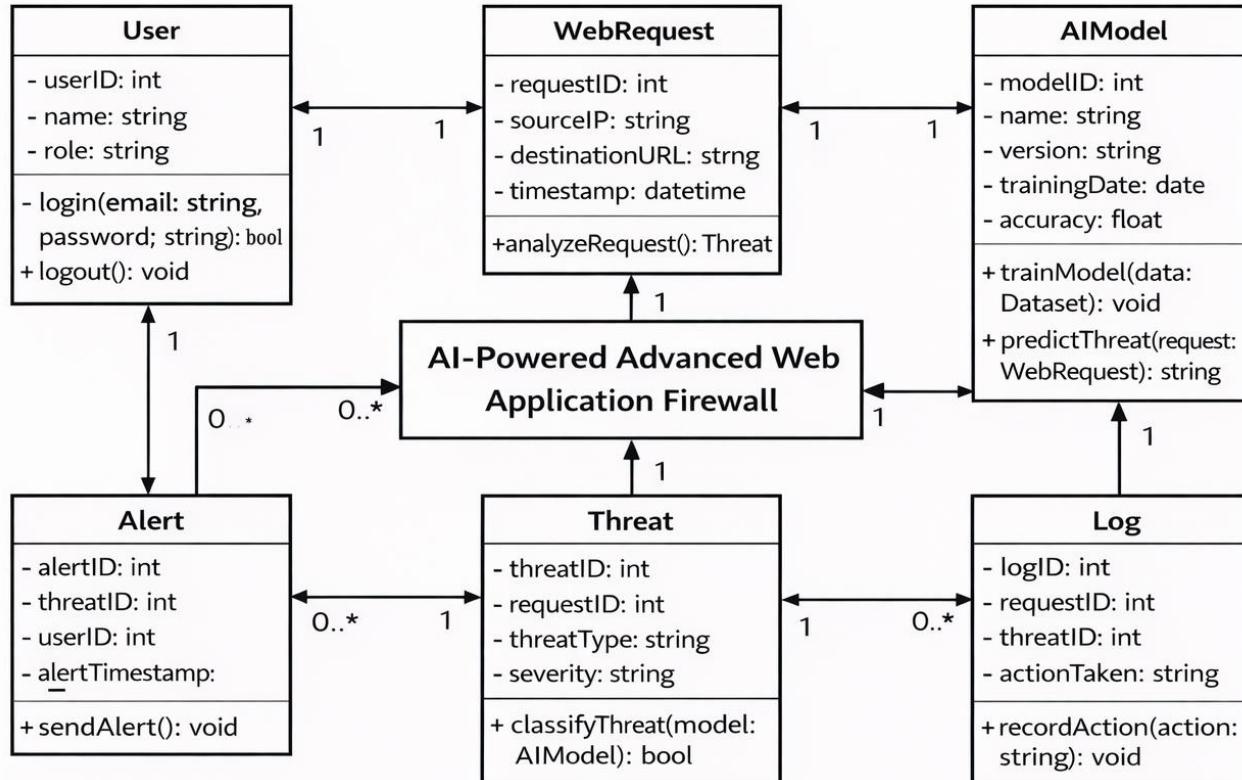
# Architecture Diagram/Flow



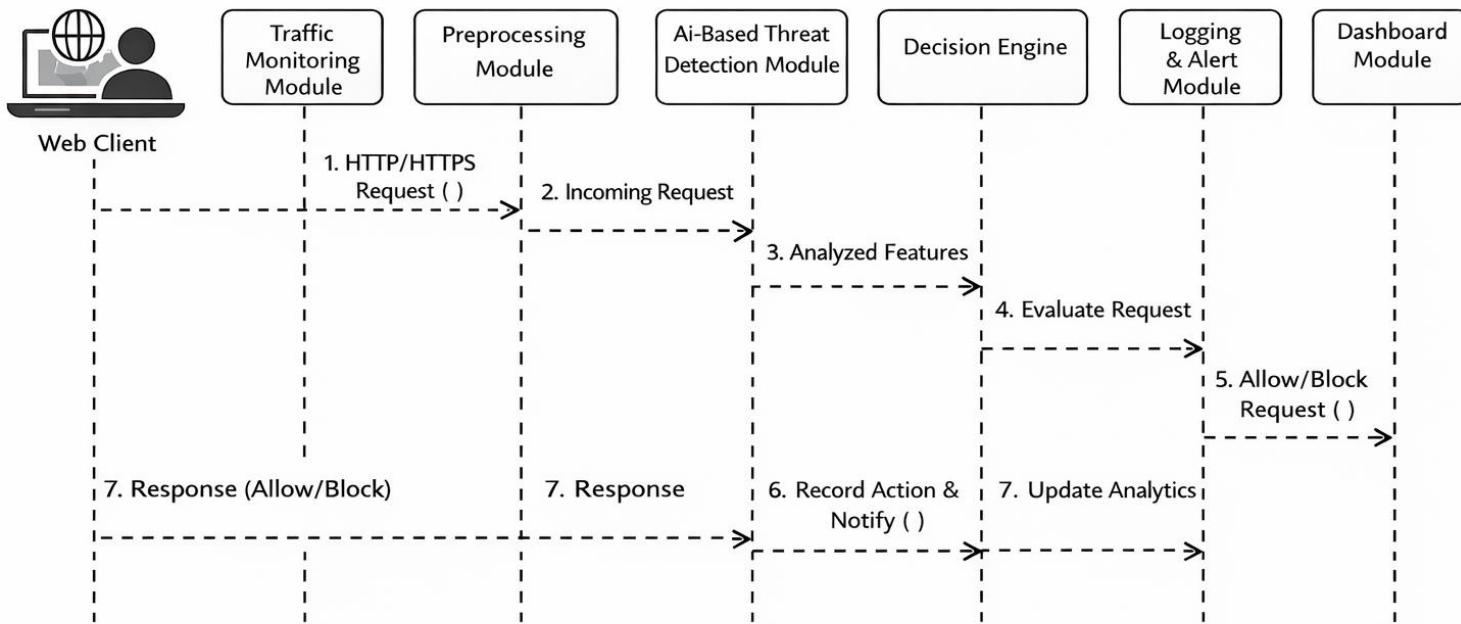
# Design-Use Case Diagram



# Design-Class Diagram



# Sequence Diagram



# Algorithms used

## Machine Learning–Based Threat Detection Model:

- The Machine Learning–based Threat Detection Model is used to analyze incoming web traffic in real time.
- It classifies HTTP requests as normal or malicious based on extracted features from headers, URLs, and payloads.
- It is trained using labeled web traffic data containing both legitimate and attack requests.
- It improves detection accuracy without increasing response latency.
- It enables adaptive learning by identifying new attack patterns and reducing false positives during request filtering.

# Algorithms used

## Anomaly Detection and Dynamic WAF Rule Generation Algorithm:

- This algorithm identifies abnormal behavior by comparing incoming traffic patterns with normal baseline traffic.
- It detects attacks such as SQL Injection, Cross-Site Scripting (XSS), and brute-force attempts.
- Based on detected threats, it dynamically generates and updates Web Application Firewall rules.
- It automatically blocks malicious requests and IP addresses in real time.
- It ensures continuous protection by adapting firewall rules according to evolving attack behaviors.

# Hardware and software selection

## Hardware:

- The project uses the following hardware components:
- Server or cloud system for hosting the WAF
- Load balancer (Nginx) for traffic distribution
- Storage for logs and ML models
- Network infrastructure for HTTP/HTTPS traffic

## Software:

- The project uses the following software components:
- Python programming language
- Flask / FastAPI framework
- Machine learning using Scikit-learn
- Nginx web server and load balancer
- Docker and Docker Compose
- Prometheus and Grafana for monitoring
- Linux operating system

# Implementation

- The system is implemented using Python-based web technologies and containerized services.
- Nginx is used as a load balancer to route incoming traffic to multiple WAF nodes.
- HTTP requests are intercepted and analyzed using a security middleware.
- Relevant features are extracted and passed to the machine learning engine.
- The ML model classifies requests as normal or malicious.
- Based on the prediction, dynamic WAF rules are generated to allow or block traffic.
- All security events and metrics are logged and visualized using monitoring tools.

# Important Code segments

## Dynamic WAF Rule Generation (AI Driven)

```
def generate_rules_from_threats(self, threats):  
    if len(threat_list) >= 3:  
        rule = self._generate_rule_for_threat_type()
```

### Description:

- Uses ML threat predictions to create **dynamic WAF rules**
- Generates rules for **SQL Injection, XSS, Bot attacks**
- Automatically blocks **IPs, URLs, User-Agents**
- Prevents duplicate rules and manages rule expiry

# Important Code segments

## Nginx Rule Conversion & Validation

```
def to_nginx_config(self):  
    return f"deny {self.condition};"  
def _validate_nginx_config(self, config):  
    return syntax_check and security_check
```

### Description:

- Converts AI-generated rules into **valid Nginx configuration**
- Validates:
  - Syntax correctness
  - Regex patterns
  - Security directives
- Ensures **safe deployment without server crash**

# Important Code segments

## Secure Rule Deployment Across Nginx Nodes

```
async def deploy_rules_to_all_nodes(self, config):  
    await self._deploy_to_node(node, config)
```

### Description:

- Deploys WAF rules to **multiple Nginx nodes**
- Supports **Docker, API, and SSH-based deployment**
- Includes **backup, rollback, and verification**
- Ensures **high availability and fault tolerance**

# Output

**WAF AI**  
CONTROL CENTER

## Dashboard

System Healthy

### System Overview

Real-time monitoring and control of your WAF AI system

**Traffic Control**  
HTTP traffic collection and monitoring

Running

Start Collection Stop Collection

Status

**54500** Requests **0** Rate/min

**ML Engine**  
Machine learning model training and inference

Ready

Train Model Model Info

**0%** Accuracy **0** Threats

**Threat Detection**  
Real-time threat analysis and blocking

Active

Start Detection Stop Detection

Status

**67** Detected **1** Blocked

**WAF Rules**  
Dynamic rule generation and deployment

Active

Generate Rules Deploy Rules

Rule Stats Cleanup

**1** Active **1** Deployed

**System Health**  
Overall system monitoring

Healthy

Health Check All Metrics

**0h** Uptime **0MB** Memory

**Master Control**  
Quick system-wide operations

Start All Stop All Restart

**System Logs**

# Output

**docker desktop** PERSONAL

Containers [Give feedback](#)

Container CPU usage (i) Container memory usage (i) Show charts

7.82% / 2000% (20 CPUs available) 643.04MB / 7.39GB

Search  Only show running containers

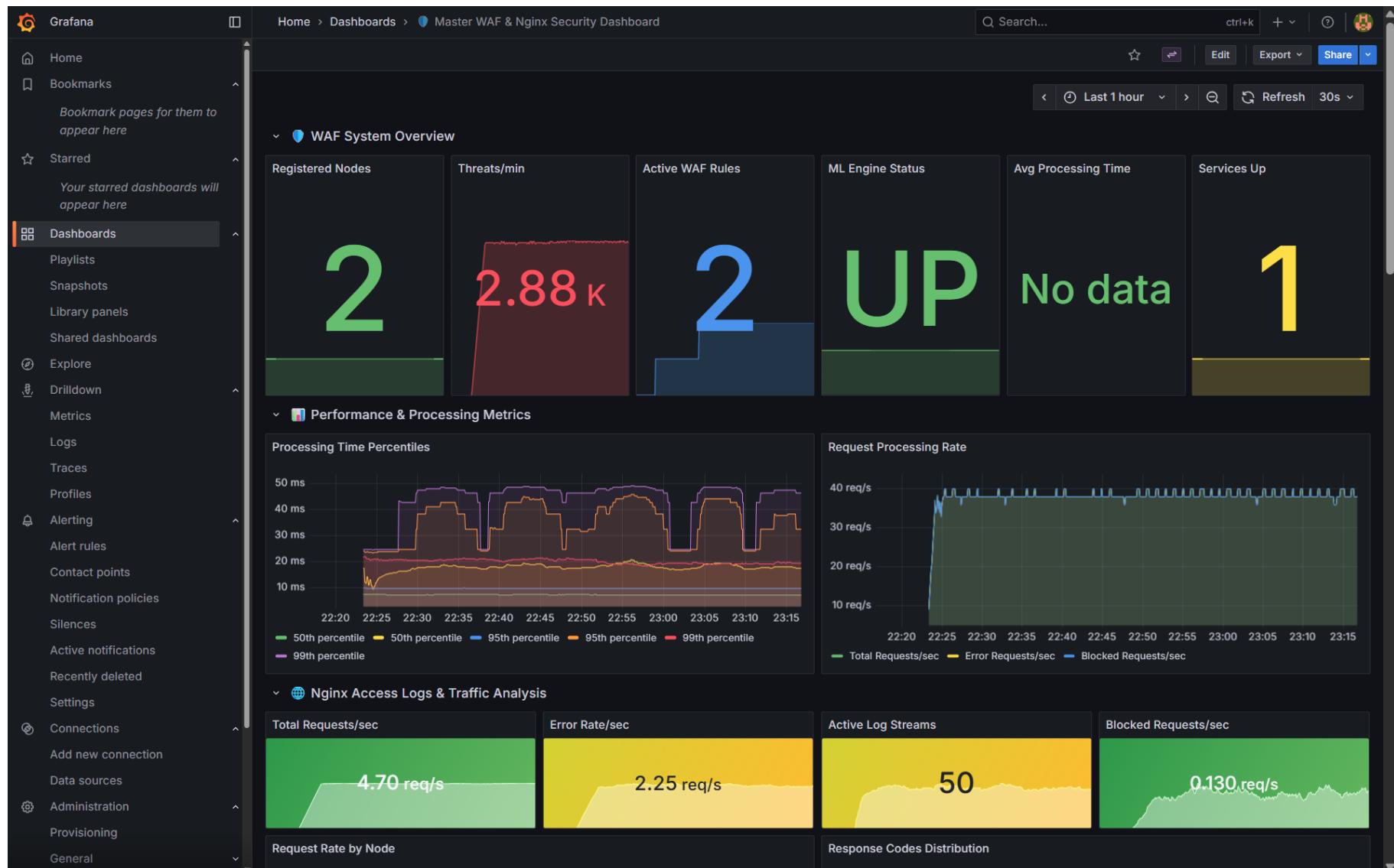
	Name	Container ID	Image	Port(s)	CPU (%)	Last started	Actions
<input type="checkbox"/>	cyberguard-ai	-	-	-	7.82%	32 minutes ago	
<input type="checkbox"/>	nginx-node-2	c8baddb30f45	nginx:alpine	8082:80	0.12%	32 minutes ago	
<input type="checkbox"/>	nginx-node-1	43e18ed00a98	nginx:alpine	8081:80	0.09%	32 minutes ago	
<input type="checkbox"/>	promtail-node-1	6e699ed8b920	grafana/promtail:latest	-	1.28%	1 hour ago	
<input type="checkbox"/>	promtail-node-2	a5333b0c7e6e	grafana/promtail:latest	-	1.53%	1 hour ago	
<input type="checkbox"/>	waf-grafana	669cb2fcc53f	grafana/grafana:latest	3080:3000	0.32%	1 hour ago	
<input type="checkbox"/>	waf-loki	9fcfa145f2e3f	grafana/loki:latest	3100:3100	1.88%	1 hour ago	
<input type="checkbox"/>	waf-redis	d1d0c84bef0d	redis:7-alpine	6379:6379	0.7%	1 hour ago	
<input type="checkbox"/>	waf-prometheus	7ff69d35381f	prom/prometheus:latest	9090:9090	0%	1 hour ago	
<input type="checkbox"/>	waf-api	fd413bf5a600	cyberguard-ai-waf-api	8000:8000	0.17%	60 minutes ago	
<input type="checkbox"/>	log-server-2	c7979f794413	cyberguard-ai-log-server	8083:8080	0.22%	60 minutes ago	
<input type="checkbox"/>	log-server-1	598e55c0d475	cyberguard-ai-log-server	8080:8080	0.2%	60 minutes ago	
<input type="checkbox"/>	traffic-generator	7d266dcfd27a	cyberguard-ai-traffic-ger	-	1.31%	60 minutes ago	

Showing 14 items

Engine running | RAM 4.79 GB CPU 4.01% Disk: 11.14 GB used (limit 1006.85 GB) | Terminal | Update version

Search 01:31 AM 19-12-2025

# Output



# Output

**WAF AI**  
CONTROL CENTER

OVERVIEW

- Dashboard
- System Status

SERVICES

- ML Engine
- Traffic Control
- Threat Detection
- WAF Rules

EXTERNAL

- Grafana
- Prometheus

ML Engine

ML Engine

Machine learning model training, evaluation and management

**Model Training**  
Train and manage ML models

Ready

**Start Training** **Model Info**

**Generate Data**

**Model Performance**  
Accuracy and performance metrics

0% Accuracy	0% Precision
0% Recall	0% F1 Score

**Feature Analysis**  
Feature importance and analysis

URL Length	0.85
Suspicious Patterns	0.75
Request Rate	0.60

**Model Information**  
Current model details and status

Model Type:	Not loaded
Training Data:	0 samples
Last Trained:	Never
Model Version:	N/A

# Test Cases

S.No	Module	Test Case	Input	Expected Output	Actual Output	Status
1	Authentication	Valid login	Admin username & password	Returns JWT token	JWT token returned, e.g., eyJhbGciOiJIUzI1NilsInR5cCl...	Pass
2	Authentication	Invalid login	Wrong credentials	401 Unauthorized	401 Unauthorized	Pass
3	User Management	Add new user	Admin token + user data	201 Created	201 Created	Pass
4	User Management	List users	Admin token	List of users	[{"id":1,"name":"John"}, {"id":2,"name":"Alice"}]	Pass
5	ML Engine	Train model	Sample training JSON	Training completes, model saved	Training completed successfully, model saved to models/	Pass
6	ML Engine	Predict threat	Test HTTP request data	Correct threat label	Threat label: SQL Injection	Pass
7	ML Engine	Incremental training	New threat data	Model updated	Model updated with new threat data	Pass
8	Traffic Collector	Start collection	Node URLs	Collector starts, logs traffic	Collector started on nodes: [node1, node2]	Pass
9	Traffic Collector	Stop collection	Active nodes	Collection stops	Collection stopped for all nodes	Pass
10	Traffic Stats	Fetch stats	Node ID	Traffic metrics JSON {"requests":1200,"blocked":35,"anomalies":5}		Pass
11	WAF Rules	Generate rules	Threat prediction data	Nginx rules file generated	nginx_rules.conf generated	Pass
12	WAF Rules	Deploy rules	Node IDs + rules	Rules deployed to nodes	Rules deployed successfully to nodes [node1, node2]	Pass

# Test Cases

13 WAF Rules	Rollback rules	Node IDs	Rules reverted	Rules rolled back successfully	Pass
14 Nginx Node	Add node	Node JSON	Node added, status OK	Node added: node3, status OK	Pass
15 Nginx Node	Node status	Node ID	Node connectivity and health	Node node1 is UP, latency 20ms	Pass
16 Control Panel	Service status	Access panel	Real-time status shown	Dashboard shows all services UP	Pass
17 Control Panel	Start/Stop service	Service name	Service starts/stops successfully	Service WAF started successfully	Pass
18 Monitoring	Grafana metrics	Access Grafana	Metrics displayed	Grafana dashboard displays CPU, memory, requests	Pass
19 Monitoring	Prometheus metrics	/metrics endpoint	Correct metrics output	{"http_requests_total":1200,"threats_detected":35}	Pass
20 Alerts	Slack/email alert	Threat detected	Alert sent	Alert sent to Slack/email for SQL Injection	Pass
21 Security	Emergency shutdown	Admin token	All services stop	All services stopped, nodes [node1,node2]	Pass
22 Security	Block IP range	CIDR + reason	IP blocked in WAF	IP range 192.168.1.0/24 blocked	Pass
23 API	Rate limiting	High request volume	Requests limited	429 Too Many Requests	Pass
24 API	Unauthorized access	No token	401 Unauthorized	401 Unauthorized	Pass
25 Logging	Access logs	HTTP requests	Logs stored and filtered	Logs stored in MongoDB, filtered by threat type	Pass

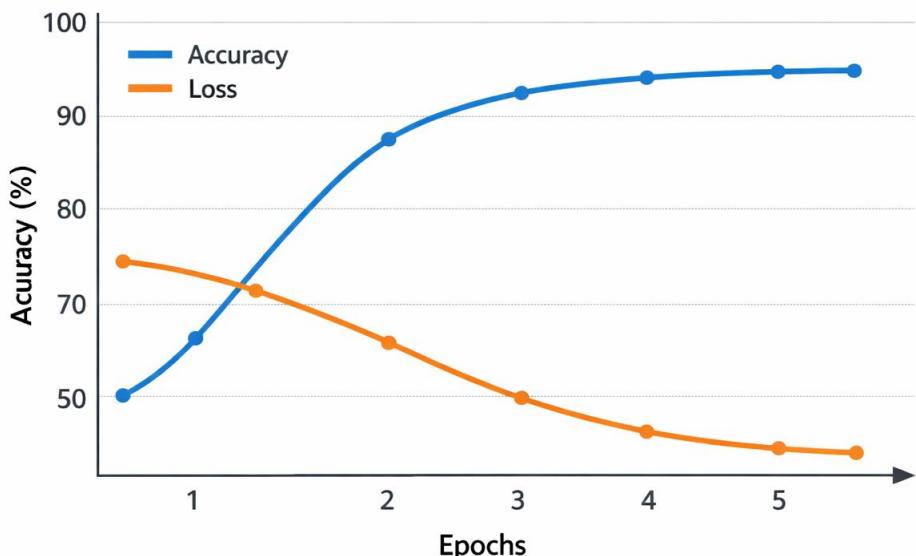
# Results

## Machine Learning Performance

- **Accuracy:** 90%
- **Precision:** 93%
- **Recall:** 92%

## Inference

- Fast detection
- Reliable predictions
- Low false positives
- Real-time analysis
- Improved system security



# Conclusion

- The **AI-Powered Advanced Web Application Firewall** successfully demonstrates the use of machine learning to monitor and analyze web traffic in real time.
- The system intelligently detects malicious requests, blocks potential threats, and dynamically generates security rules.
- By combining automation with AI, the firewall reduces manual intervention, improves detection accuracy, and enhances overall web application security.
- This project proves that AI-based WAF solutions are effective, scalable, and suitable for modern web environments.

# Future Work

- In future, the system can be enhanced by training the machine learning model with larger and more diverse real-world traffic datasets to improve detection accuracy.
- Advanced deep learning techniques can be integrated to identify zero-day and evolving cyber attacks more effectively.
- The firewall can be extended to support cloud-native and distributed environments with auto-scaling capabilities.
- Integration with threat intelligence feeds and SIEM tools can further strengthen real-time threat response and security analytics.

# References

1. A. Patcha and J. M. Park,  
*An Overview of Anomaly Detection Techniques*, Computer Networks, 2007.
2. N. Moustafa and J. Slay,  
*UNSW-NB15: A Comprehensive Data Set for Network Intrusion Detection*, IEEE, 2015.
3. Z. Alom et al.,  
*Intrusion Detection Systems Using Deep Learning Techniques*, IEEE Access, 2019.
4. Project GitHub Repository:  
<https://github.com/Darkwebnew/AI-Powered-Advanced-Web-Application-Firewall.git>
5. **Project Demo Video:**  
Demonstration of AI-Powered Advanced Web Application Firewall (Self-Recorded)

# Questions



# Thank You