

# Configuration Manual

MSc Research Project  
MSc Cybersecurity

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**MSc Project Submission Sheet**



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**Programme:** MSc Cybersecurity

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**Lecturer:** Vikas Sahni

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**Project Title:** Improving API Security in the Cloud with AI/ML

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# Configuration Manual

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**Project Title:** Improving API Security in the Cloud with AI/ML

**Environment:** AWS EC2 (Ubuntu 24.04 LTS, t3.micro)

**Main Components:** Flask APIs, Random Forest, Autoencoder, JWT + Rate Limit + Regex Signatures

## Tools and Technologies Used

Component	Tool/Service	Version
Programming Language	Python	3.12
Framework	Flask	3.03
ML Libraries	Scikit-learn , Keras , TensorFlow	1.5.2, 3.3.3, 2.17.0
Data Handling	Pandas , NumPy	2.2.2. 1.26.4
API Testing	curl , Postman , OWASP ZAP	8.9.1, v11.10, 2.15.0
Deployment	AWS EC2 (Ubuntu 24.04 LTS, t3.micro instance)	
Monitoring	htop , logging, manual timing	3.3.0
Security Add-ons	Flask-Limiter , PyJWT , regex-based pattern match (re module Python)	3.8.0, 2.9.0

## EC2 INSTANCE SETUP

### Launch EC2 Instance

- **AMI:** Ubuntu 24.04 LTS (64-bit)
- **Instance type:** t3.micro
- **Security Group Rules:**
  - Allow **SSH** (Port 22)
  - Allow **HTTP** (Port 80)
  - Allow **Custom TCP Rule** (Port 5000 or 8000 for API)

### Connect to Instance via SSH

```
ssh -i "your_key.pem" ubuntu@your-ec2-public-ip
```

## Python Environment Setup

```
sudo apt update && sudo apt install python3-venv python3-pip -y
python3 -m venv venv
source venv/bin/activate
```

## Install Required Libraries

```
pip install flask scikit-learn pandas numpy keras tensorflow flask-limiter pyjwt gunicorn
```

## PROJECT STRUCTURE

```
api_security_ml_project/
├── app/ # API logic
│   ├── api.py # Flask app
│   ├── utils.py # Signature + JWT + Rate Limiter
│   ├── rf_predictor.py # Random Forest logic
│   └── autoencoder_predictor.py # Autoencoder logic
├── models/
│   ├── rf_model.joblib
│   └── autoencoder_model.h5
├── data/
│   ├── api_logs.csv
│   └── labeled_api_logs.csv
├── test_payloads/
│   ├── normal_login.json
│   ├── malicious_delete_user.json
│   ├── large_payload_ddos.json
│   └── xss_payload.json
├── evaluation_summary.txt
├── detected_anomalies.json
└── requirements.txt
```

## MODEL TRAINING (Optional if models already saved)

### Random Forest

```
from sklearn.ensemble import RandomForestClassifier
from sklearn.preprocessing import LabelEncoder
import pandas as pd, joblib

df = pd.read_csv("data/labeled_api_logs.csv")
X = df[['endpoint', 'method', 'payload_size', 'status_code']]
y = df['label']
```

```
# Encode categorical
for col in ['endpoint', 'method']:
    X[col] = LabelEncoder().fit_transform(X[col])

rf = RandomForestClassifier(n_estimators=100)
rf.fit(X, y)

joblib.dump(rf, 'models/rf_model.joblib')
```

## Autoencoder (Keras)

```
from keras.models import Sequential
from keras.layers import Dense
import numpy as np

X_train = np.loadtxt('data/normalized_cicids.csv', delimiter=',')

model = Sequential([
    Dense(64, activation='relu', input_shape=(X_train.shape[1],)),
    Dense(32, activation='relu'),
    Dense(64, activation='relu'),
    Dense(X_train.shape[1])
])
model.compile(optimizer='adam', loss='mse')
model.fit(X_train, X_train, epochs=20, batch_size=32)

model.save('models/autoencoder_model.h5')
```

## RUNNING THE APPLICATION

### Start the Flask API Server

```
gunicorn -b 127.0.0.1:8000 app.api:app
```

If using Nginx as a reverse proxy, map port 80 to 8000 internally.

## TESTING VIA CURL

### Traditional Detection

#### Valid Request:

```
curl -X POST http:// your-ec2-ip /traditional_check \
-H "Content-Type: application/json" \
-d @test_payloads/normal_login.json
```

#### Missing/Invalid Token:

```
curl -X POST http:// your-ec2-ip /traditional_check \ -H "Content-Type: application/json" \ -d
'{"user_id": "123", "endpoint": "/data", "method": "GET"}'
```

### **XSS Attack:**

```
curl -X POST http:// your-ec2-ip /traditional_check \  
-H "Content-Type: application/json" \  
-d '{"payload": "<script>alert(\"XSS\")</script>", "token_used":  
"valid_token_123"}'
```

### **Random Forest Endpoint**

#### **Normal:**

```
curl -X POST http:// your-ec2-ip /predict \ -H "Content-Type: application/json" \ -d  
@test_payloads/normal_login.json
```

#### **Malicious Delete User Request:**

```
curl -X POST http://your-ec2-ip/predict \  
-H "Content-Type: application/json" \  
-d @test_payloads/malicious_delete_user.json
```

#### **Large Payload DDoS Simulation :**

```
curl -X POST http:// your-ec2-ip /predict \ -H "Content-Type: application/json" \ -d  
@test_payloads/large_payload_ddos.json
```

### **Autoencoder Detection**

```
curl -X POST http://your-ec2-ip/detect_autoencoder \  
-H "Content-Type: application/json" \  
-d @test_payloads/large_payload_ddos.json
```

## **MONITORING & LOGGING**

### **View Logs**

```
tail -f data/api_logs.csv
```

### **System Monitoring**

```
htop # Monitor CPU/RAM
```

## **COMMON ISSUES & FIXES**

<b>Problem</b>	<b>Fix</b>
Autoencoder crash on EC2	Reduce dataset size or use t3.medium instance
RF model misclassifying new endpoints	Retrain with more diverse data
No response from API	Check gunicorn port binding or Nginx proxy config
Memory overflow	Use batch processing or split data during inference

## PACKAGING FOR DEPLOYMENT

- Zip the following folders and send to client/supervisor:
  - /models/
  - /test\_payloads/
  - /app/
  - run\_server.sh (startup script)

Example run\_server.sh:

```
#!/bin/bash
source venv/bin/activate
gunicorn -b 127.0.0.1:8000 app.api:app
```