

# Towards Automation of Penetration Testing for Web Applications by Deep Reinforcement Learning

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## Introduction

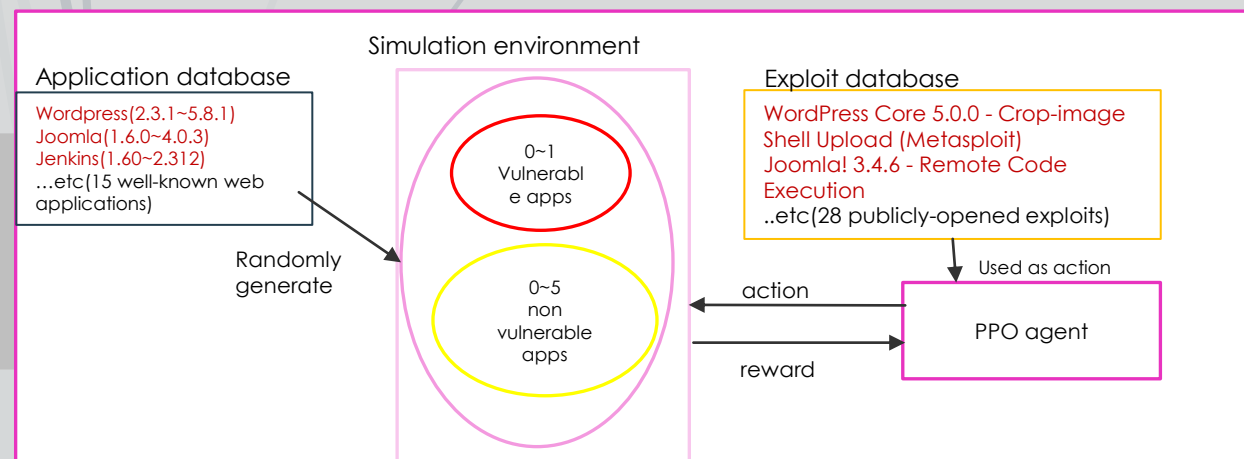
**Penetration testing (PT)** that assesses vulnerabilities by considering and executing all possible attacks is important in security engineering but very expensive due to the need of experienced professionals.

As a countermeasure, there are attempts[1]–[4] to partially automate and improve the efficiency of PT. Such approaches do not embed ML in PT tools, and would not improve the tools themselves.

In this work, we use **deep reinforcement learning** to automate search and exploit executions for various vulnerabilities existing in Web applications so that a wide variety of PT tools can be integrated in an effective manner with embedded ML.

## Experiment 1

Using manually collected application versions from their official webpage and exploits from **exploit-db**, launched simulation environment and **PPO** agents to choose effective exploits from application versions.



**Goal of environment:** success exploit actions or give up action within 10 chance of actions.

**Actions:** 44 type

15 kind of Scan action: Proving corresponding application versions and mapping to observation **reward:** 0(detected information)/ -1(failed)

28 kind of Exploit action: If target version is vulnerable, the episode ends with success.

**reward:** 1(the agent found exploit appropriate for the target version)/ -1(failed)

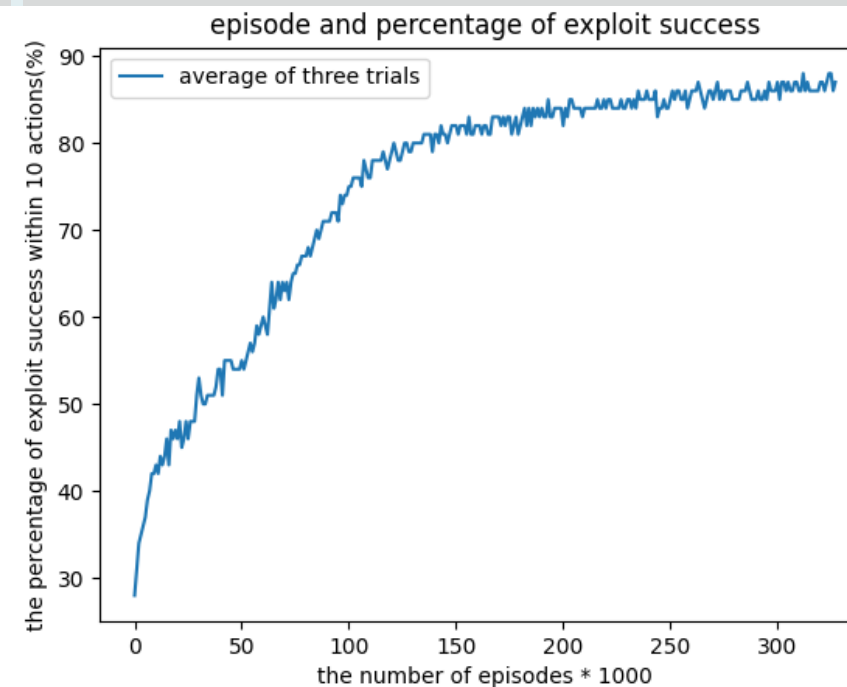
Give up action: declare "there is no vulnerability here" and end the episode.

**reward:** 1(if there is really no vulnerable application): -2(left vulnerable application)

**States:** 15 Application Versions (their range is mapped -1 to 1. (ex.4.8.1→[0.4,0.8,0.1] ) )

## Result 1

The result averaged over three trials is shown below.



As a result, the percentage of success reached nearly 90%.

This shows the task can be handled by deep reinforcement learning.

## Experiment 2

connect the **real-life applications** and exploits to the learned model of Experiment 1 and demonstrate the applicability of our approach in more realistic environment.

### Defined actions:

for wordpress, Jenkins, joomla, phpmyadmin.

Other actions are defined only the shape, and they definitely fails.

Defined Action	Affected Versions
Wordpress scan	All
Wordpress exploit	5.0.0 and <=4.9.8
Jenkins scan	All
Jenkins exploit	1.60-1.658
Joomla scan	All
Joomla exploit	3.0.0-3.4.6
phpmyadmin scan	All
phpmyadmin exploit	4.8.0-1

### Environment 1:

When the exploits correct for this environment have **wider** range of target versions.

Applications	Versions	Vulnerable Action
Wordpress	5.8.1	Not Vulnerable
Jenkins	2.137	Not Vulnerable
Joomla	3.4.3	Joomla exploit
phpmyadmin	5.1.1	Not Vulnerable

### Environment 2:

When the exploits correct for this environment have **relatively narrower** range of target versions.

Applications	Versions	Vulnerable Action
Wordpress	5.8.1	Not Vulnerable
Jenkins	2.137	Not Vulnerable
Joomla	4.0.3	Not Vulnerable
phpmyadmin	4.8.1	phpmyadmin exploit

## Result 2

As a result of 30 demonstrations in each of the two states:

### • Environment 1

Found valid exploits :26

Gave up :4

Could find correct exploits many times.

### • Environment 2

Time up :22

Gave up :8

Couldn't find correct exploits, and wasted time to useless exploits.

So the target range of exploit affects the performance of PT.

## Future works

- exploits of RCE only in specific versions are not the only means of attack.

→ Improvements based on such logic would make our proposed approach more realistic and easier to use.

- The agent should correctly determine that the environment is harmless if the exploit unexpectedly fails because of DBMS, OS, and other causes

→ adjust the algorithm parameters so that the time series data can be learned correctly.

## Reference

[1] <https://github.com/13o-bbrbbq/machine-learning-security/tree/master/DeepExploit>, accessed on October 25, 2021.

[2] <https://github.com/rapid7/metasploitframework/wiki>, accessed on October 31, 2021.

[3] Mohamed C. Ghanem and Thomas M. Chen, "Reinforcement Learning for Intelligent Penetration Testing", 2018 Second World Conference on Smart Trends in Systems, Security and Sustainability (WorldS4), 2018.

[4] Ovidiu Valea and Ciprian Opris, "Towards Pentesting Automation Using the Metasploit Framework", 2020 IEEE 16th International Conference on Intelligent Computer Communication and Processing (ICCP), 2020