



PESIDIOUS - Create Mutated Evasive Malware Using Artificial Intelligence

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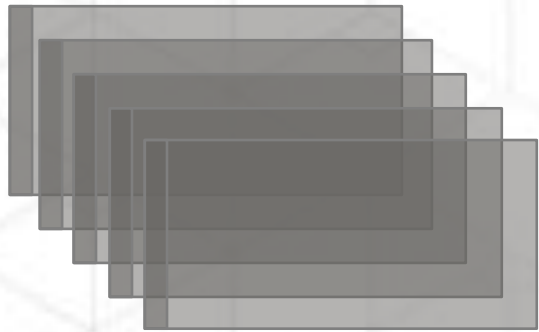
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HITB **LOCKDOWN** ⁰⁰²
livestream



Contents of this Presentation





Contents of this Presentation

Who are we?

Pesidious

Implementation

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Who are **we**?



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Why Pesidious?



PESIDIOUS |

Malware Mutation Using **Reinforcement Learning** and **Generative Adversarial Networks**

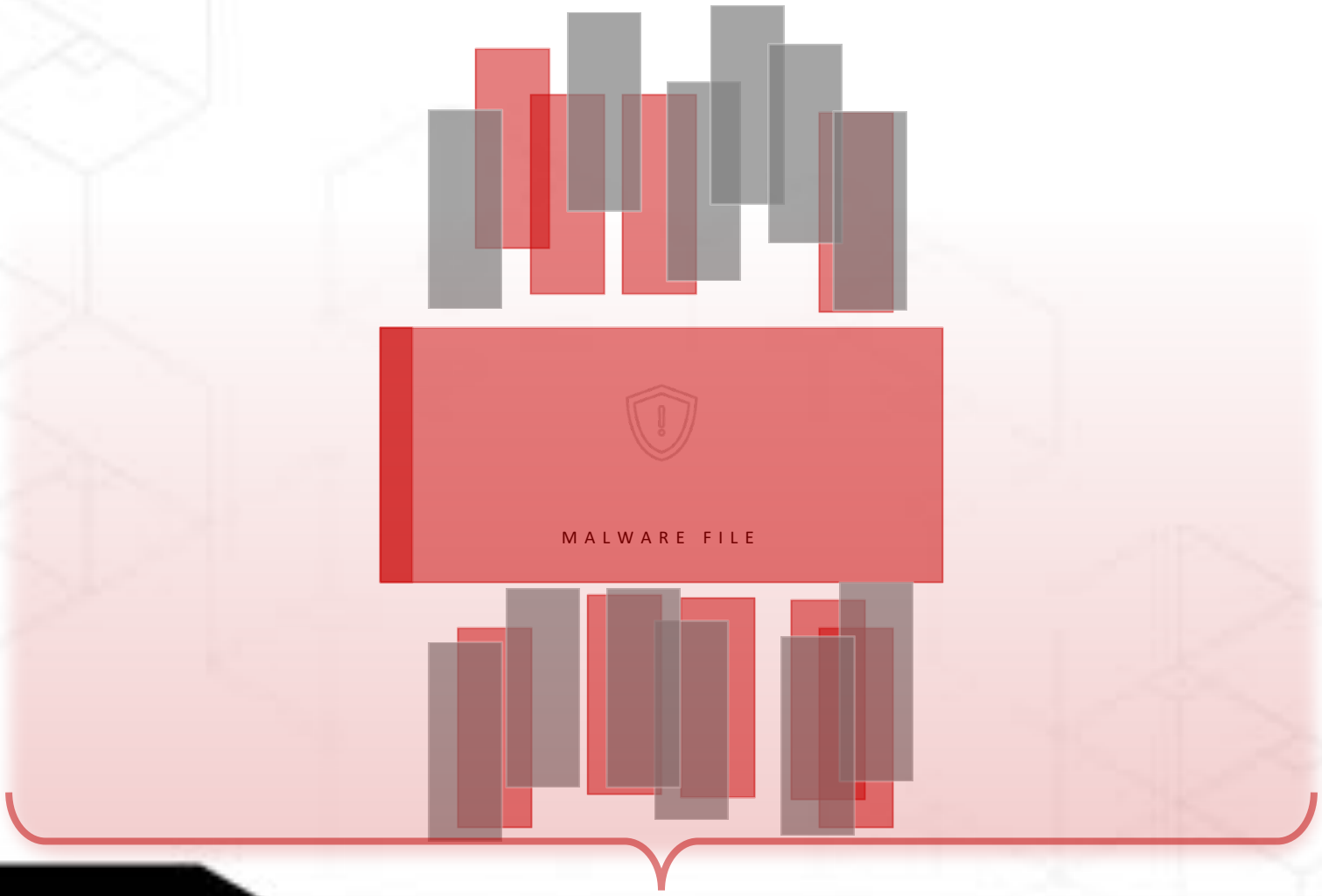
PESIDIOUS AI MUTATION SOLUTION





PESIDIOUS |

Malware Mutation Using **Reinforcement Learning** and **Generative Adversarial Networks**





PESIDIOUS |

Malware Mutation Using **Reinforcement Learning** and **Generative Adversarial Networks**



Benign looking
malware file

What is | Reinforcement Learning



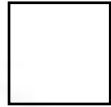
Agent



Environment



Goal



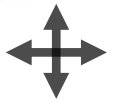
Q-Value (+ve)



State



Q-Value (-ve)



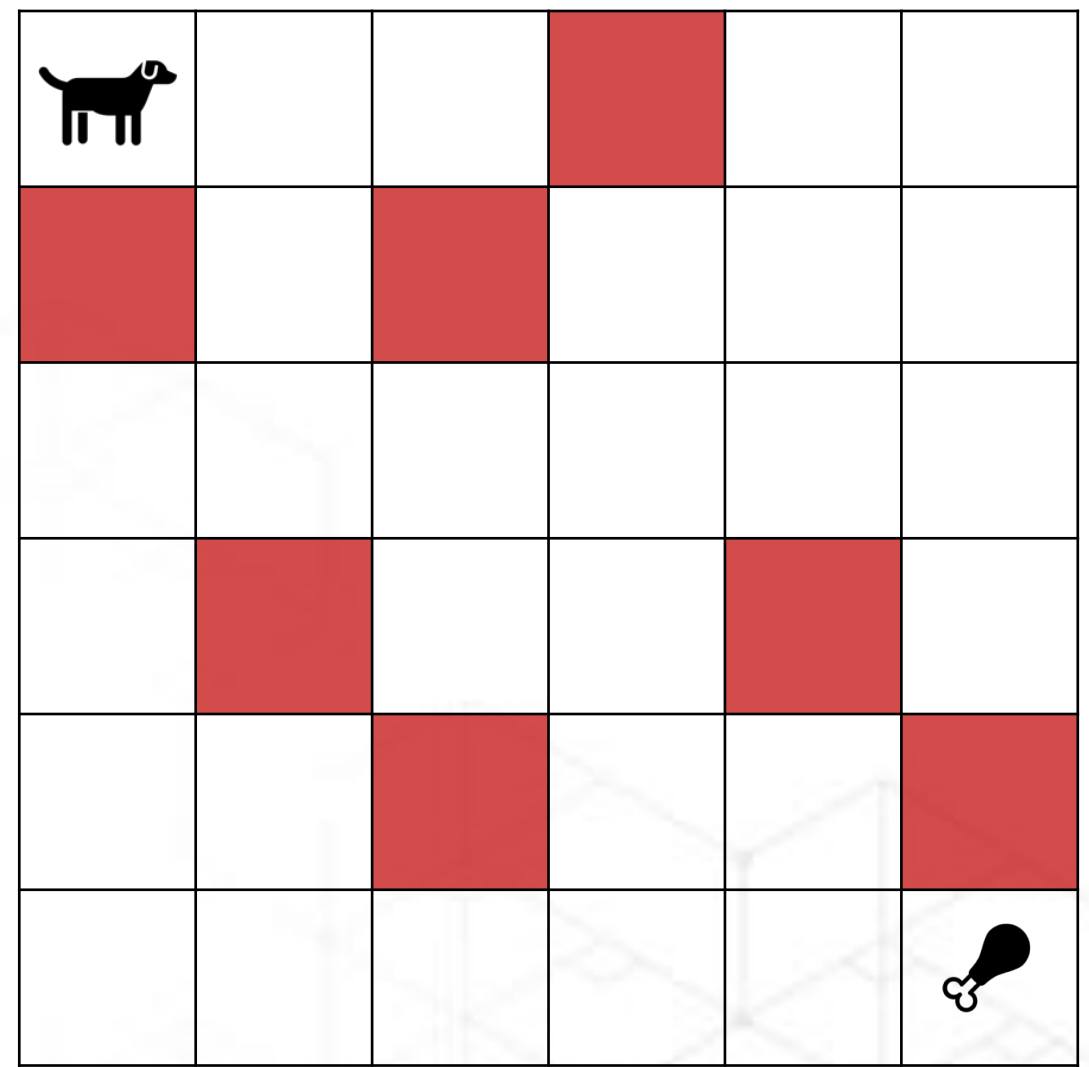
Actions



What is | Reinforcement Learning

	↑	↓	→	←
1	0	-0.51	0.7	0
2	0	0.84	0.5	0.2
4	0	0.87	0.64	0.05
34	0.34	0	0.86	0.21
35	0.55	0	1	0.31

Q-TABLE

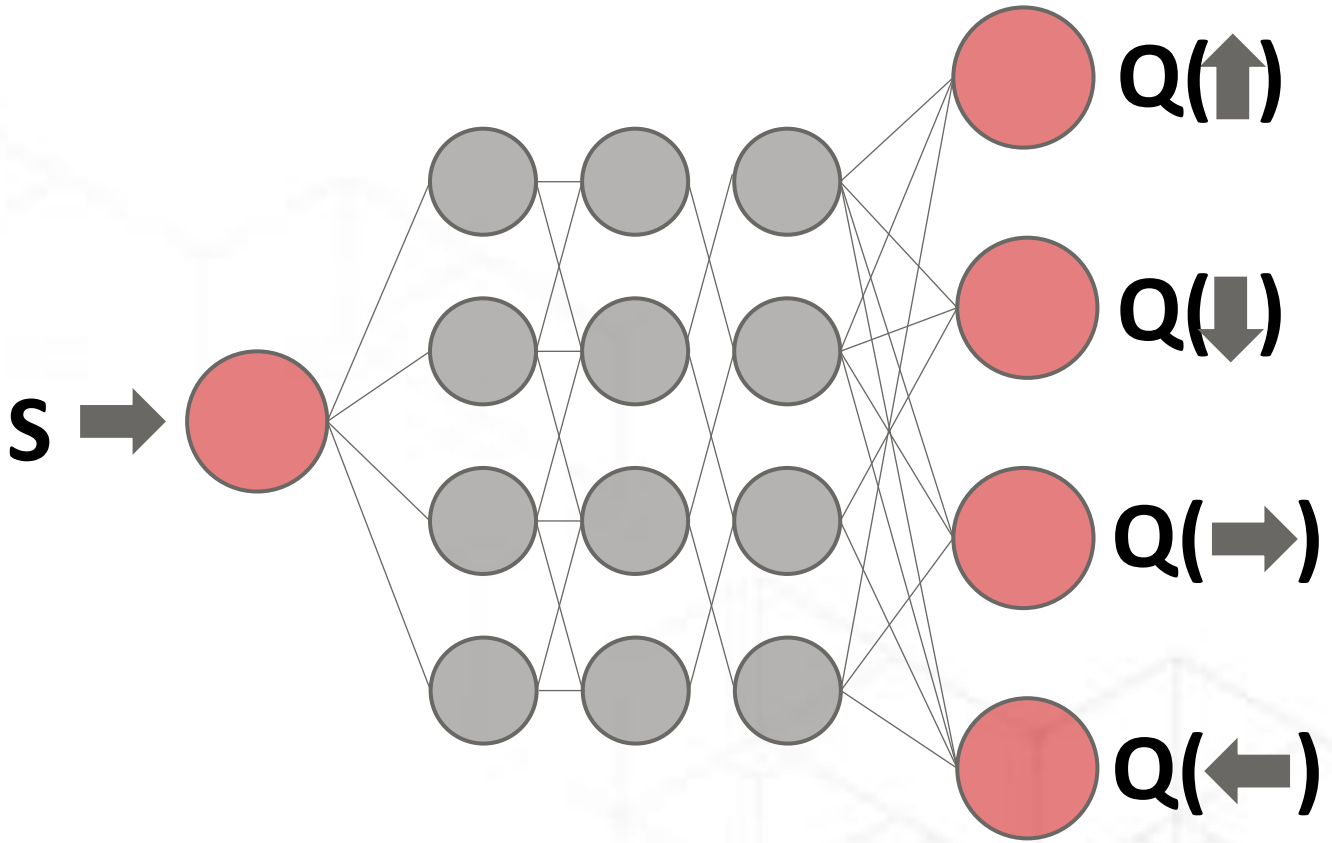




What is | Deep Reinforcement Learning

	↑	↓	→	←
1	0	-0.51	0.7	0
2	0	0.84	0.5	0.2
4	0	0.87	0.64	0.05
34	0.34	0	0.86	0.21
35	0.55	0	1	0.31

Q-TABLE



Neural Network



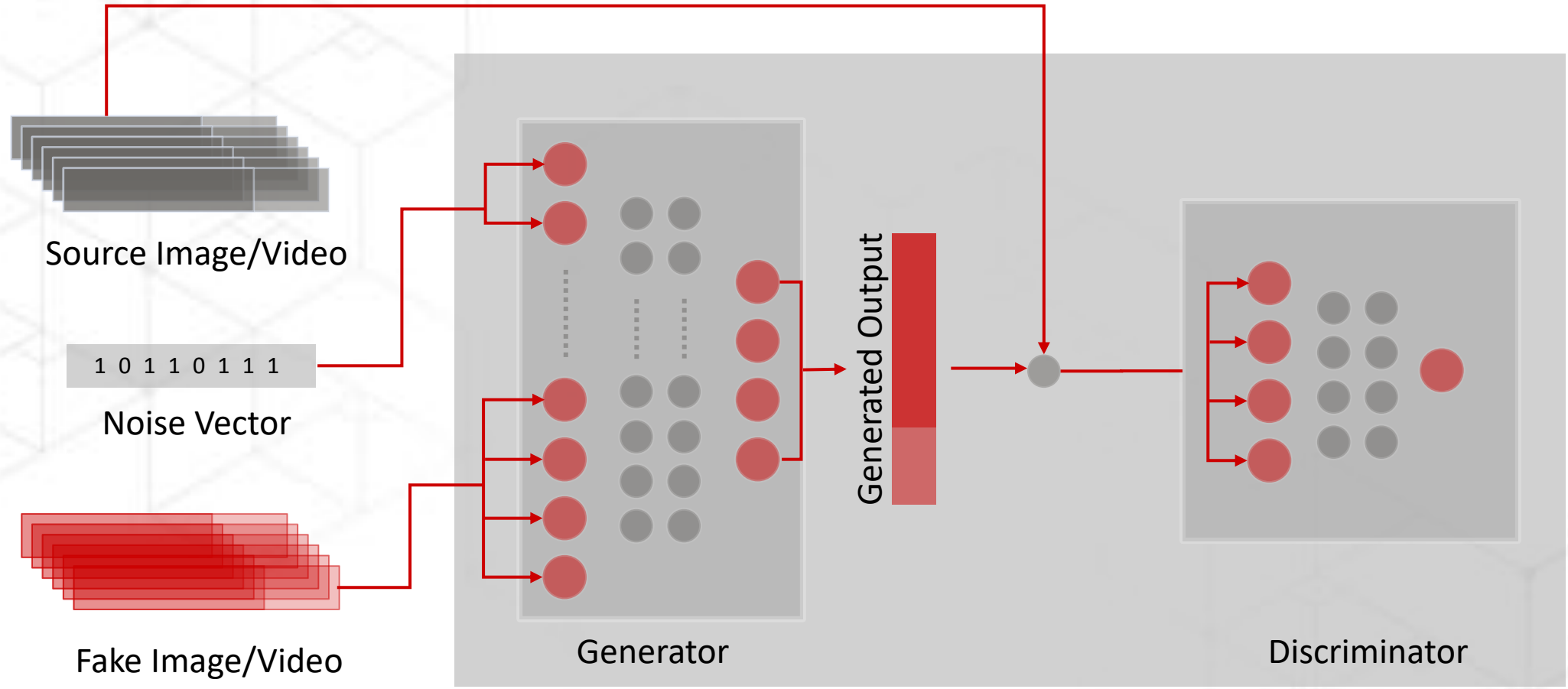
What is |

Generative Adversarial Networks





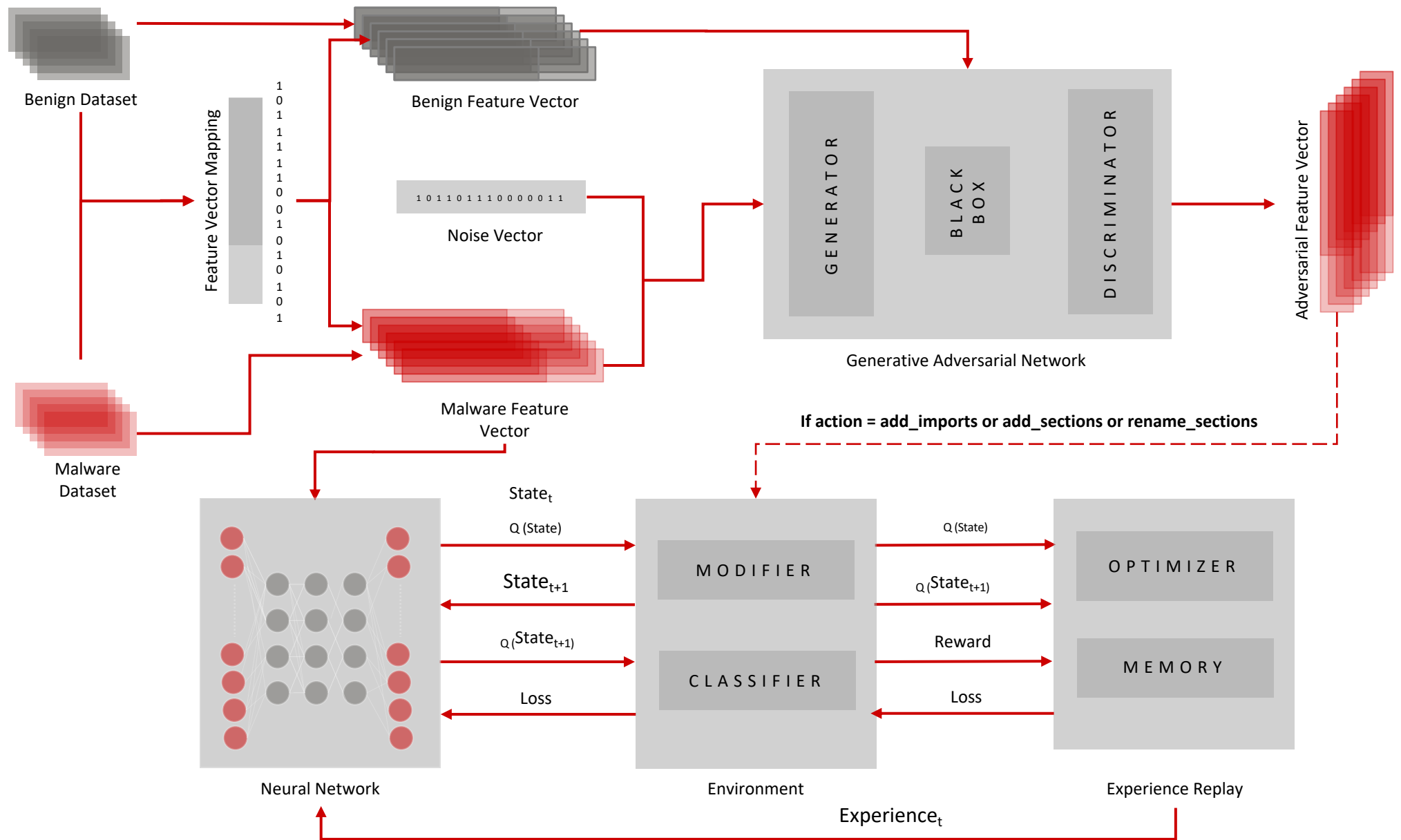
What is | Generative Adversarial Networks



Generative Adversarial Network

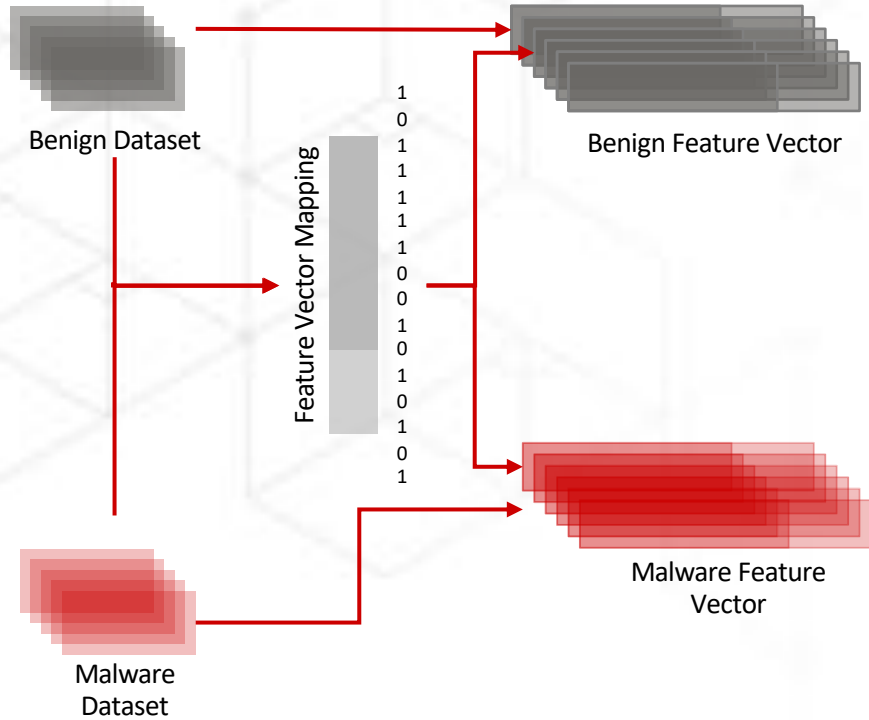
PESIDIOUS

Malware Mutation Using Reinforcement Learning and Generative Adversarial Networks



Implementation |

Extracting Features into Feature Vector Maps



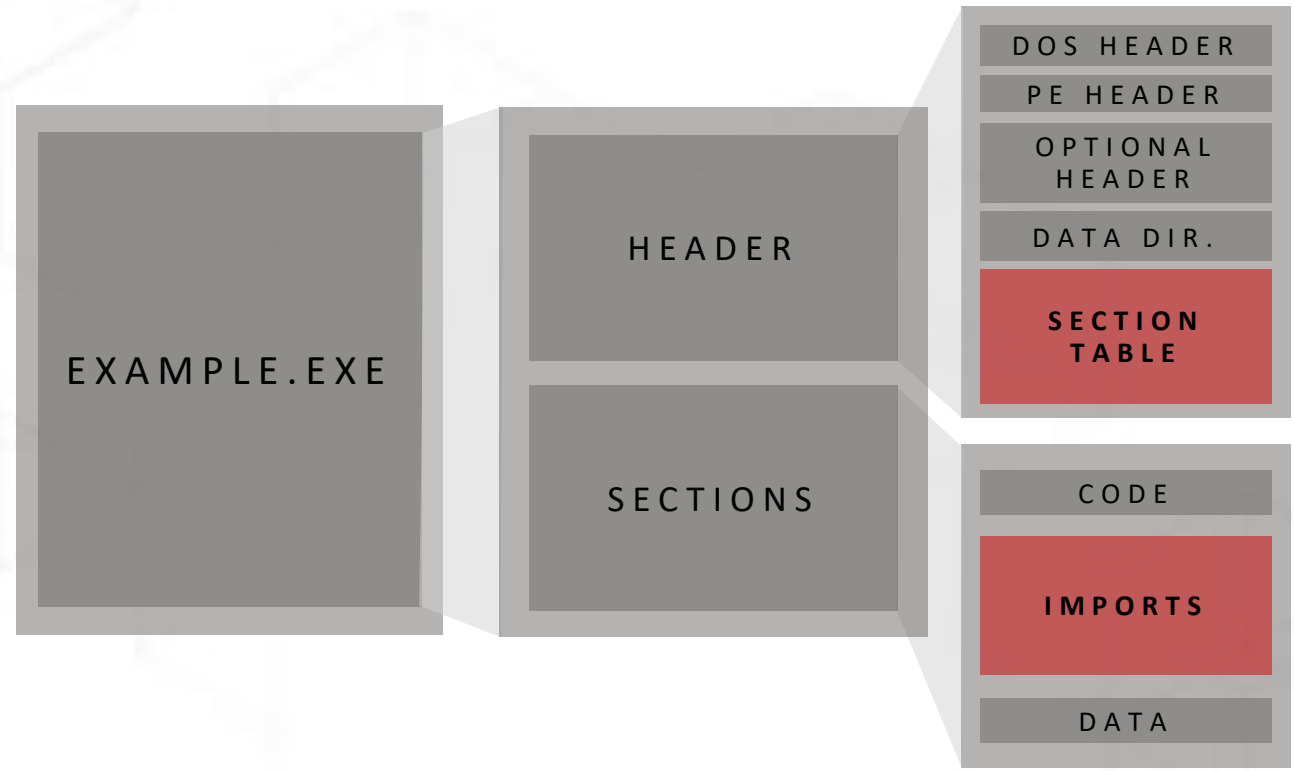
- 1.** Collect the malicious and benign binary dataset.
- 2.** Extract all the features into a single feature vector map
- 3.** Generate feature vectors for each binary data using the feature vector map

1



Implementation |

Extracting Features into Feature Vector Maps



1

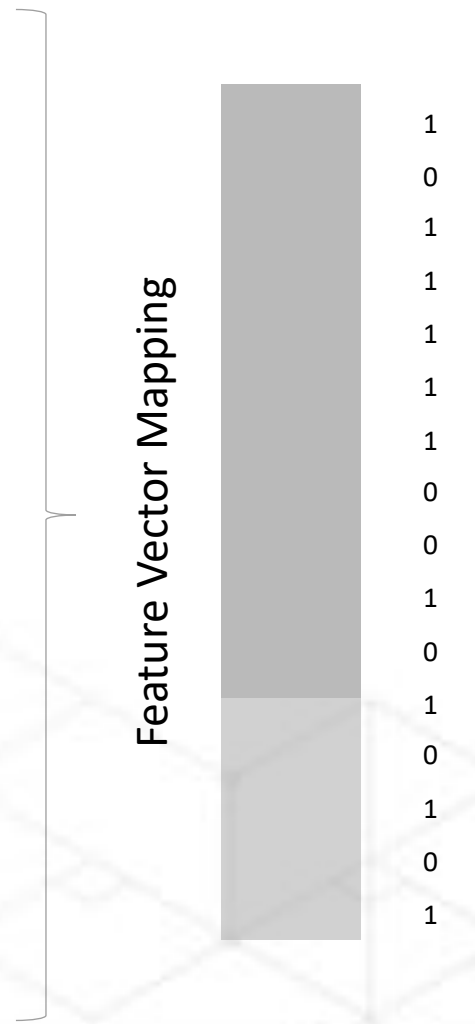
Implementation |

Extracting Features into Feature Vector Maps



- SECTION INFORMATION
- PROPERTY OF ENTRY POINTS
- IMPORTS INFORMATION
- EXPORT INFORMATION
- MACHINE ARCHITECTURE OS
- OTHER HEADER INFORMATION

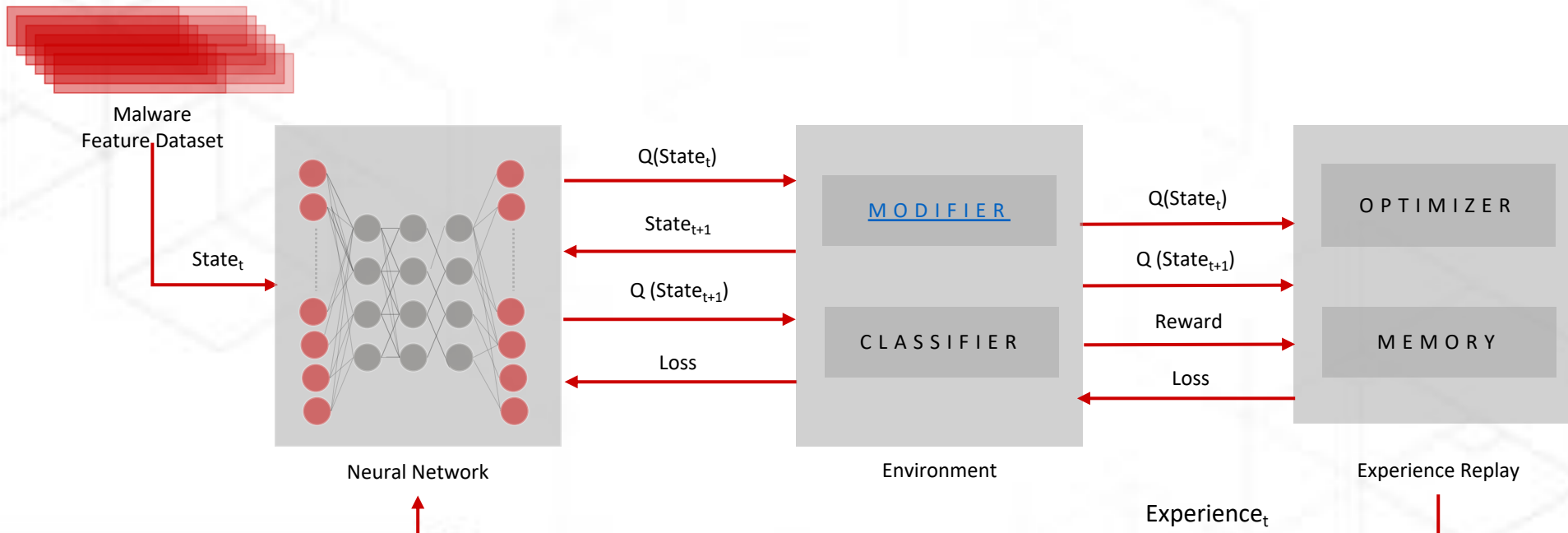
- Number of sections
- Number of sections with non zero value or empty name
- No of sections with different characteristics
- DLL
- Functions
- Virtual size
- Debug flag
- Relocation
- Resources
- Signature



Implementation |

Training a Deep Reinforcement Learning Agent

1. Implement the environment for the agent to learn.
2. Design a Deep learning model to select the actions based on the current state of the malware.
3. Use experience replay with prioritized replay buffer.



Implementation |

Training a Deep Reinforcement Learning Agent

MODIFIER

RANDOMLY ADDING IMPORT FUNCTIONS AND DLLS

RANDOMLY ADDING SECTIONS AND

RENAMING SECTIONS

APPENDING TO EXSITING SECTION

APPENDING RANDOM BYTES

REMOVING/ADDING SIGNATURE

REMOVING DEBUG FLAG

UPX PACK/UNPACK

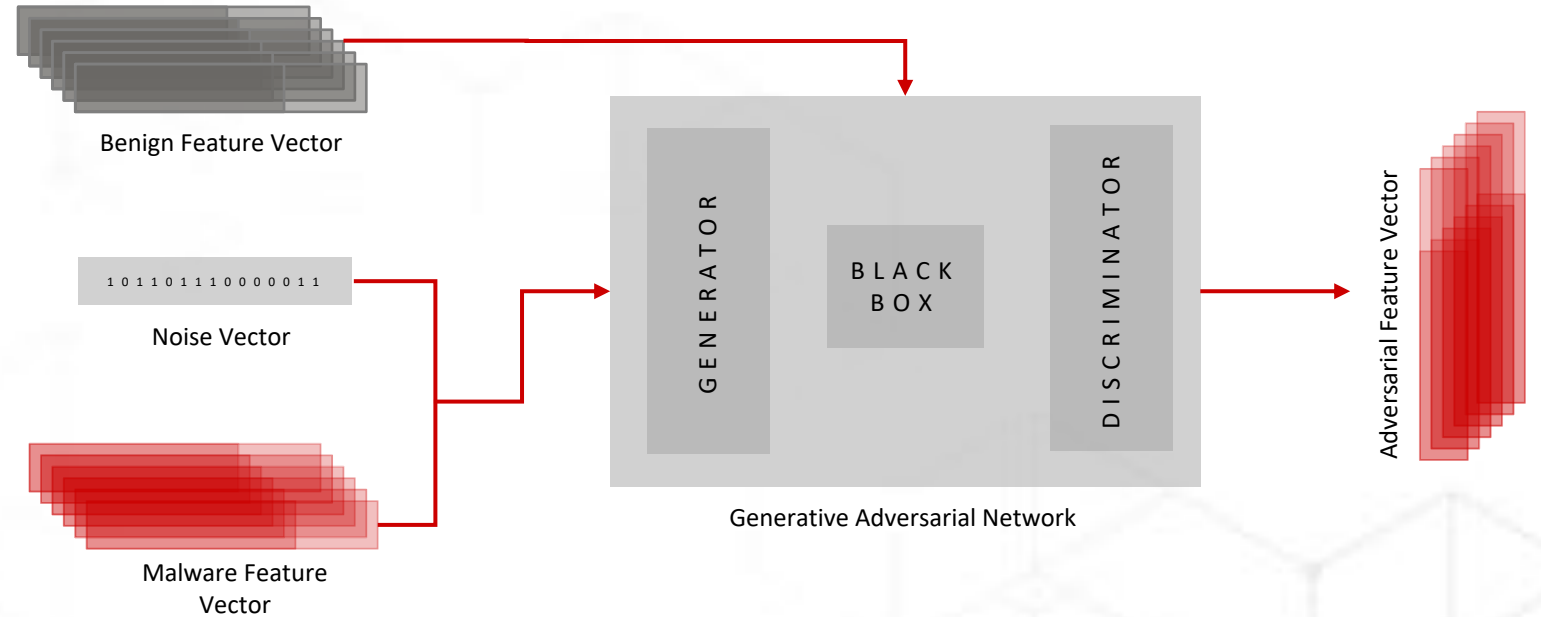
2

Implementation |

Generating Adversarial Feature Samples with Generative Adversarial Networks

1. Feature vectors are concatenated with noise and fed to the GAN.

2. The GAN generates adversarial feature vectors.



Implementation |

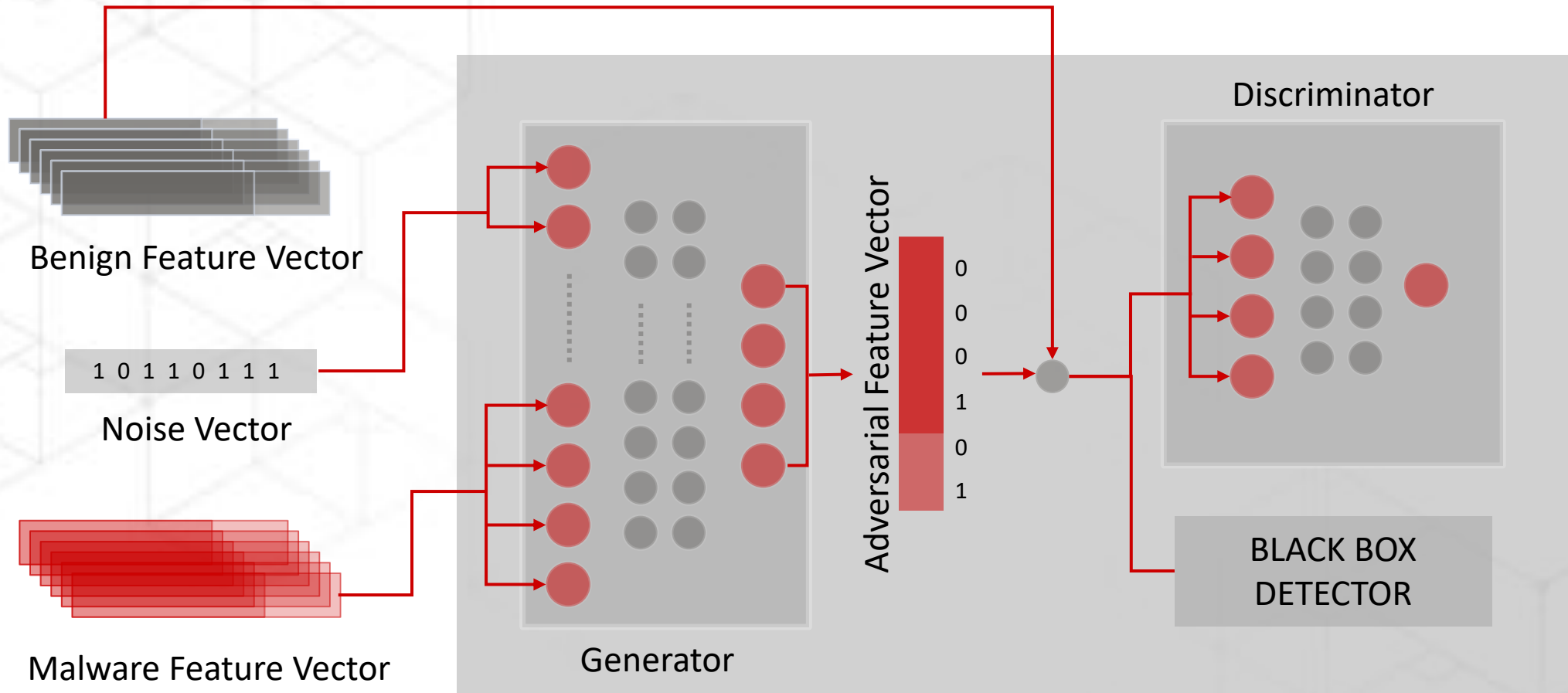
Generating Adversarial Feature Samples with Generative Adversarial Networks

DECISION TREE
LOGISTIC REGRESSION
MULTI LAYER PERCEPTRON
RANDOM FOREST
SVM

BLACK BOX

Implementation |

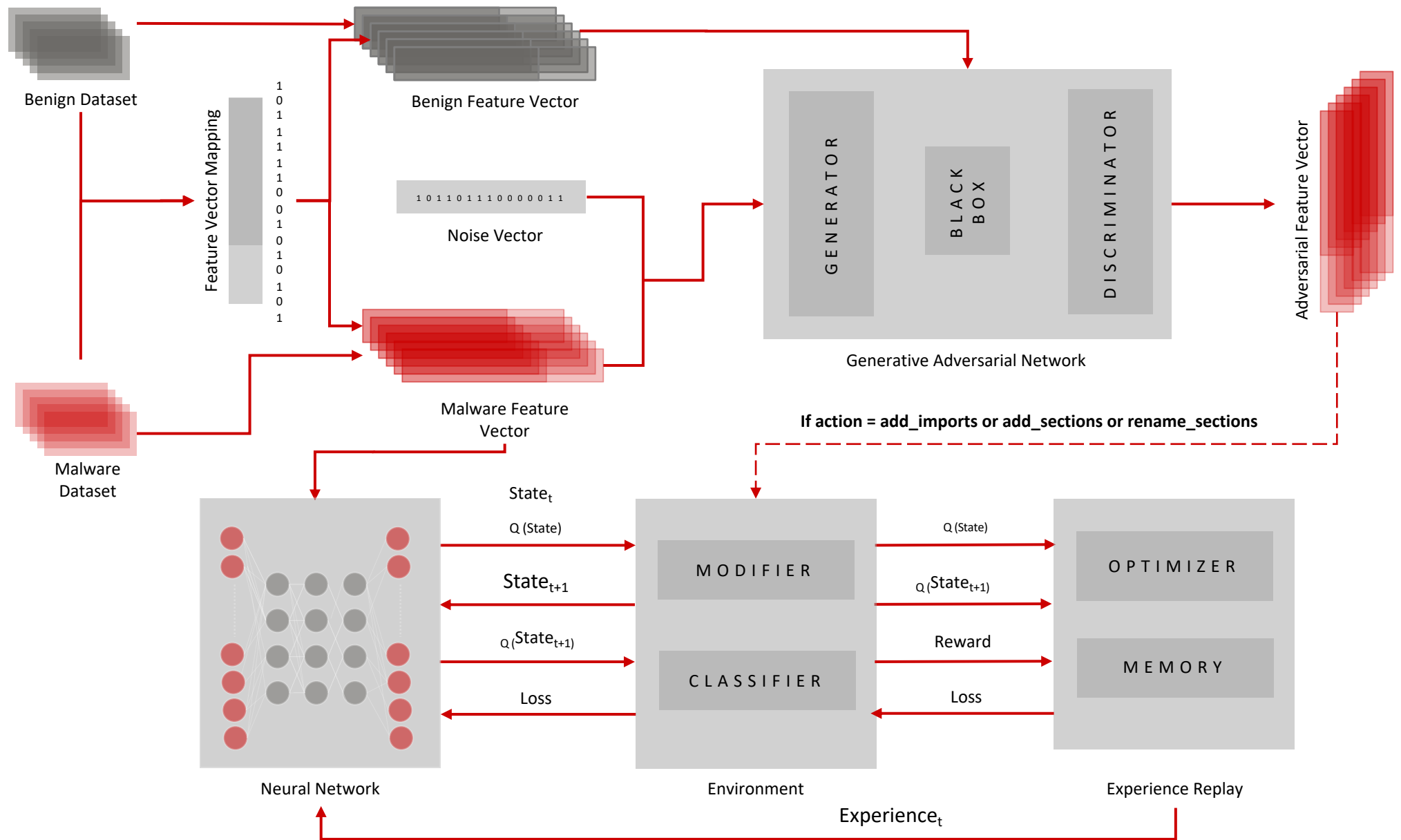
Understanding the Generative Adversarial Network



Generative Adversarial Network

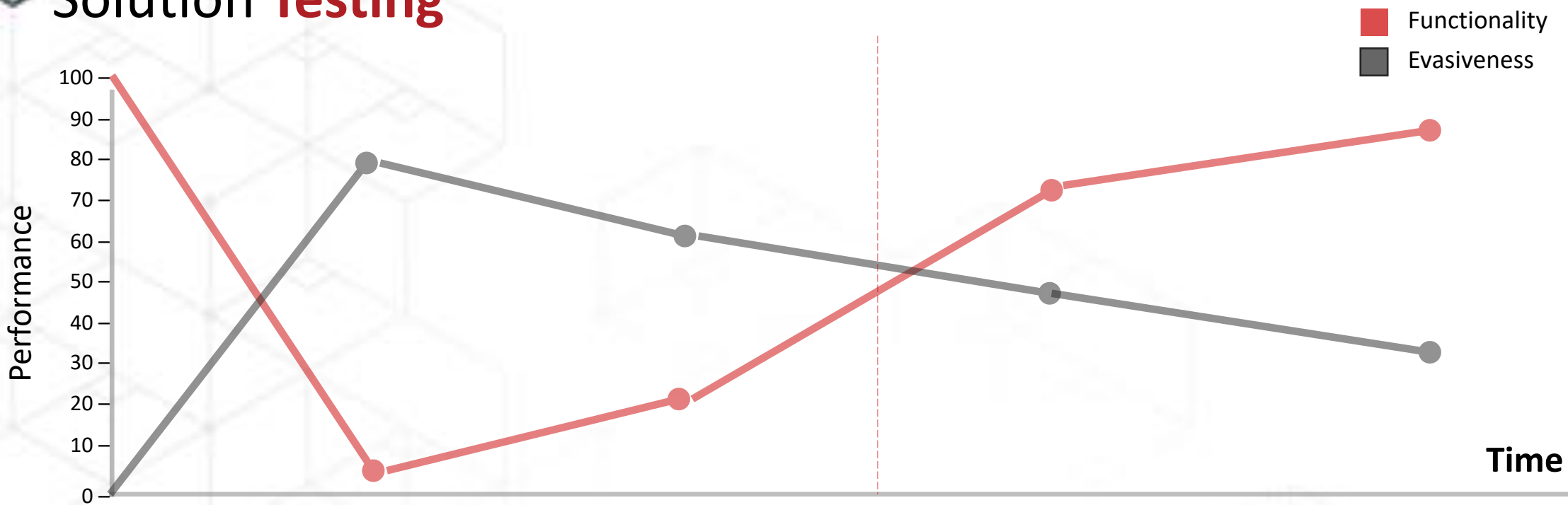
PESIDIOUS

Malware Mutation Using Reinforcement Learning and Generative Adversarial Networks





Solution Testing



- Maintaining the functionality:

- Filtering out the PE32 files based on 32 bit
- Filtering out DLLs and sections
- Using C++ instead of Python for the malware reconstruction

- Improving performance:

- Using a combination of the machine learning models scores
- Initially we trained it with backdoors; now we are giving it more diverse malwares
- For testing we made a comparison between AI and human



Project Demo

The screenshot shows a presentation slide on the left and a terminal window on the right. The slide features the logo for PESIDIOUS, which includes a stylized skull icon above the word "PESIDIOUS" in a bold, sans-serif font. Below the name is the tagline "Creating Chaos with Mutated Evasive Malware." The terminal window on the right shows a shell prompt with the user "bedang_sen@cuckoo" and the current directory "Documents/tAZchi". The terminal content is mostly obscured by a dark, textured overlay.



Project **Demo**

Run Our Mutated Malware In A Cloud Based **Secure Sandboxed Environment.**

Variant.Ransom.Cerber.171:

66 detected

<https://bit.ly/2DaxtVz>

Mutated Variant.Ransom.Cerber.171:

40% more evasive

100% functionality

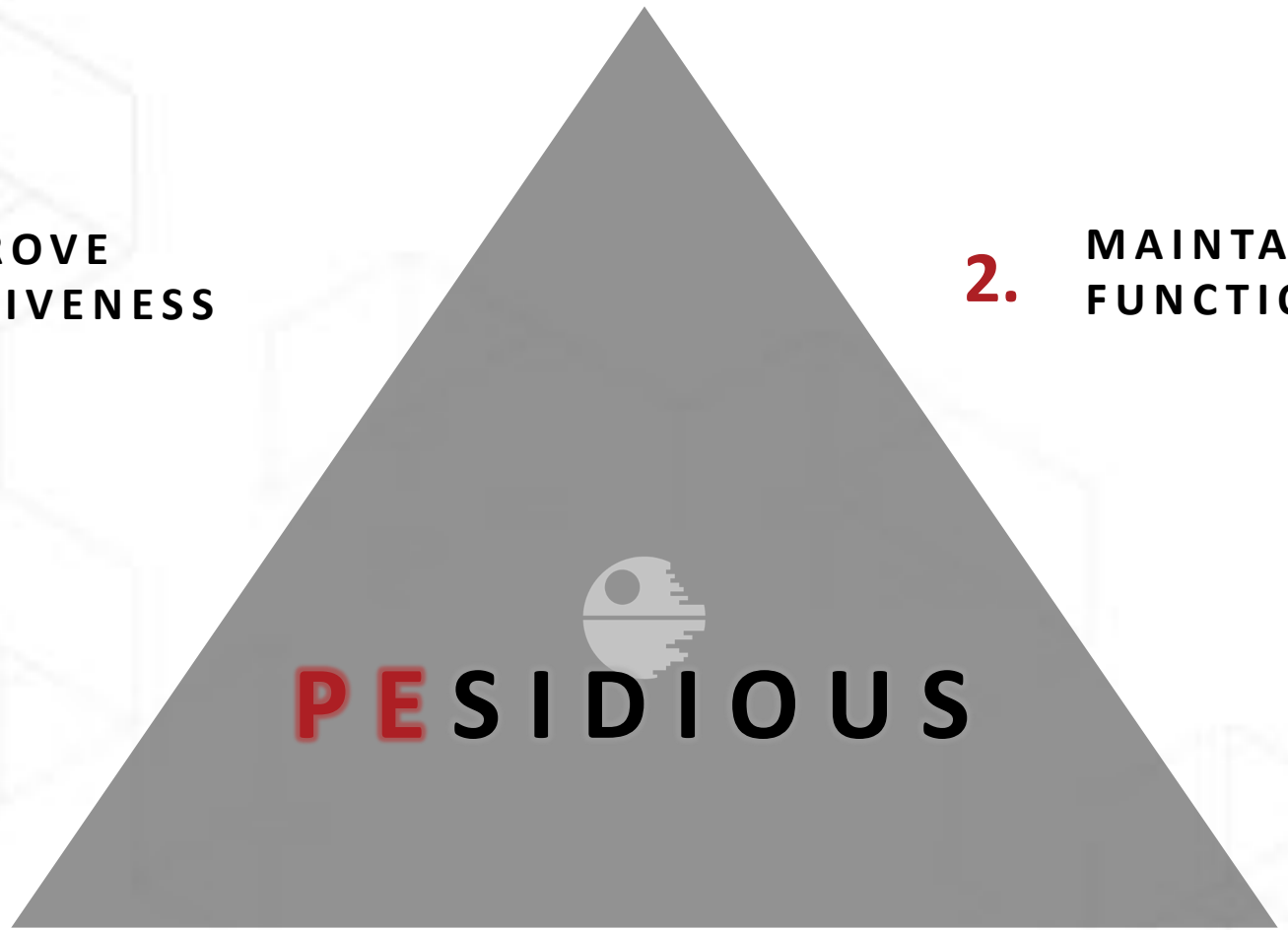
<https://bit.ly/32TFTLU>



Future Work

1. IMPROVE
EVASIVENESS

2. MAINTAIN
FUNCTIONALITY



PESIDIOUS

3. HELP THE NEXT-GEN
ANTI-VIRUS SYSTEMS



Thank You!

HITB LOCKDOWN⁰⁰²
livestream