## ZTWeb: Cross site scripting detection based on zero trust

《Computers & Security》

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## Introduction

## **Zero Trust**

#### "never trust, always verify" (Rose et al., 2020)

Using strong authentication methods, leveraging network segmentation, preventing lateral movement, providing Layer 7 threat prevention

#### • Users:

User identity, application of "least access" policies, and verification of user device integrity.

#### Applications:

Applications cannot be trusted and continuous monitoring at runtime is necessary to validate their behavior

#### • Infrastructure:

Infrastructure-related—routers, switches, cloud, IoT, and supply chain—must be addressed with a Zero Trust approach

#### XSS attack defense and zero trust

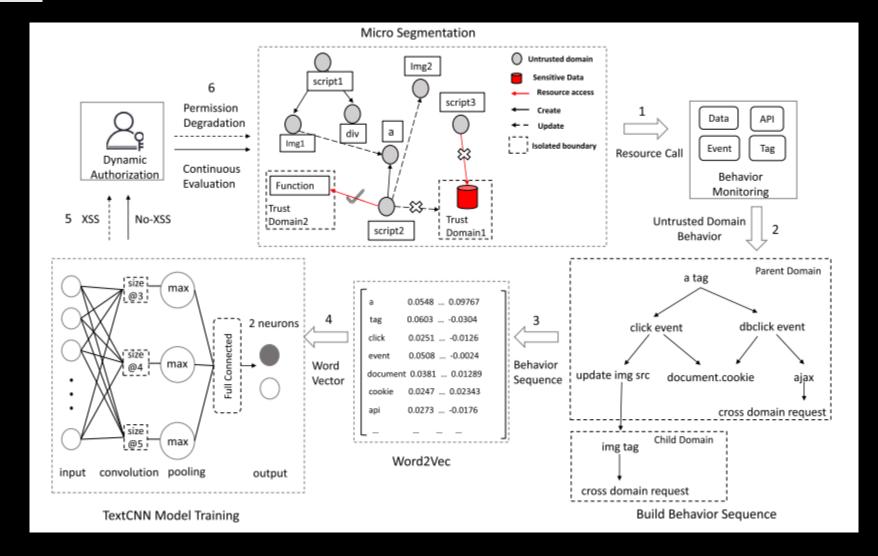
The static defense policy is difficult to adapt to the dynamic change of attack means.

Through continuous monitoring and evaluation of non-protected surface behaviors, and creates a gray "sometimes" area to the traditional blackand-white block-allow access model.

- Differentiation policy:
   Trust domain & untrusted domain.
- Dynamic authorization:
   Behavior sequence based on the untrusted domain & Adjust.
- Extraction of key features TextCNN model.



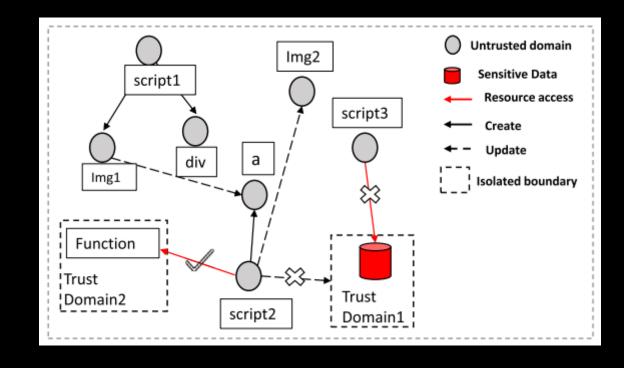
Methods & Implementation



#### 1. Micro-segmentation

#### Trust Domain1:

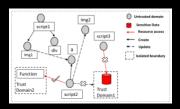
The essence of policy-based XSS defense technology is to authorize the user of sensitive resources.



#### Trust Domain2:

Take differentiated authorization for different isolation domains and intercept illegal lateral movement between isolation domains

- protect surfaces
  - 1) Services based on sensitive data
  - 2) XSS-like code written by the developer



- protect surfaces
  - (1) Unique identification

the script element and the img tag are isolated to the trust domain.

```
<script
accesstoken="050ed93d3ca311ed9871dce9948ef32c">
Javascript Code
</script>
<img
accesstoken="050ed93d3ca311ed9871dce9948ef32c"
src="a.jpg" onclick="Javascript Code"></img>
```

#### (2)Protecting sensitive data

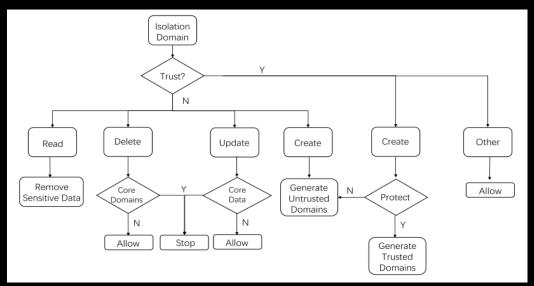
does not allow sensitive data exposure to other domains.

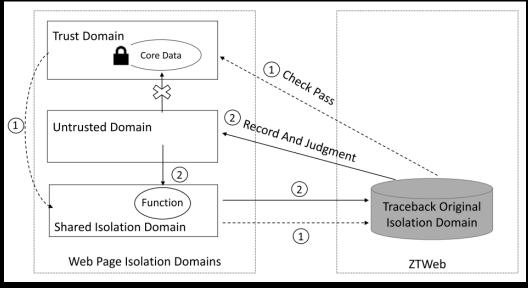
#### (3) Preventing penetration

Add a trust protection mechanism to restrict the untrusted domain. (Protection Module)

#### (4) Data sharing

Isolated domain in which the function caller resides





attack surfaces

<script

Separate the attack surface from the trust domain

(1) resource authorization still requires continuous trust evaluation Script element accept user input.

```
<%! String data="xss \"
onerror=alert(document.cookie) title=\""; %>
<img accesstoken="trust id" src="a.jpg"
alt="<%=data%>"></img>
```

SetAttribute API to load user input data into the attribute.

#### 2. Build Behavior Sequence

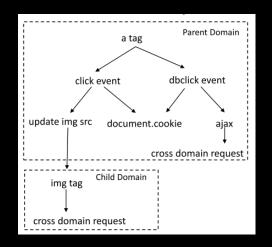
Continuous trust evaluation and dynamic authorization to authorize resources for untrusted domain

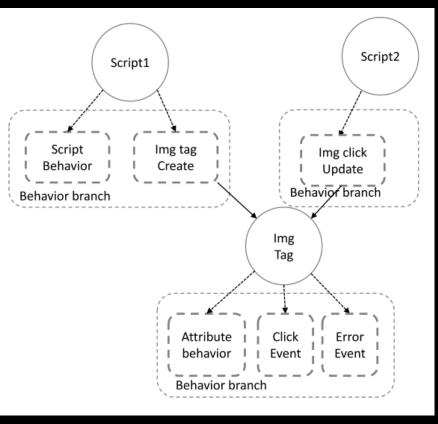
- Untrusted domain as the basic monitoring unit
- Continuously records its resource access behavior

```
<script>
document.write(' <img src=
"http://hackip/xss?cookie=' +document.cookie+' "
width=0 height=0 border=0/>' );
</script>
```

Attribute behavior branch & Event branch divides the script domain into the

- initialization script branch
- tag creation
- tag modification branch





#### 3. Feature extraction based on Word2Vec

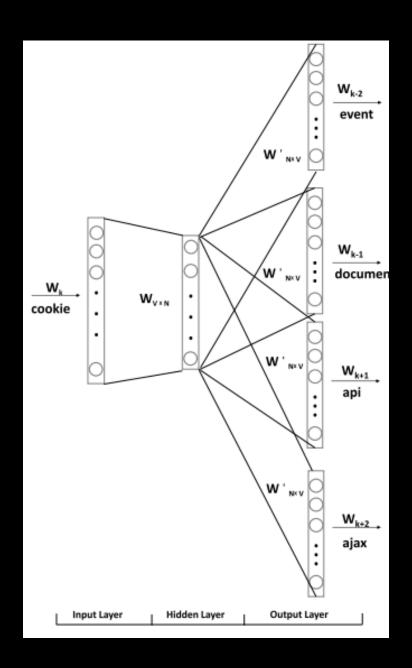
The word embedding model can convert the behavior sequence into a word vector.

Word2Vec (Mikolov et al., 2013) is a language model proposed by Google in 2013.

- CBOW model can predict the probability of central words according to surrounding words.
- Skip-Gram model predict the probability of the surrounding words according to the central word.



- Skip-Gram model
- 1)Input layer
  The one-hot encoding of the central word
- 2)Hidden layer  $WV \times N$ , V represents the vocabulary size in the training sample, and N is the number of neurons.
- 3)Output layer context of its specified window size



XSS detection based on TextCNN

Constructed behavior sequence is a short text and the low-latency scene of detection, TextCNN has a simple structure, fast training, and retains the semantic relationship between word. (Keras & Tensorflow)

1)Word embedding

Each behavior sequence can be represented by a single-channel N\*d matrix.

2)Convolution Layer

$$t_i = f(w \cdot R_{i:i+h-1} + b) \quad t = [t_1, t_2, ..., t_{n-h+1}]$$

three convolution kernels with different window sizes to extract different features of the behavioral sequence.

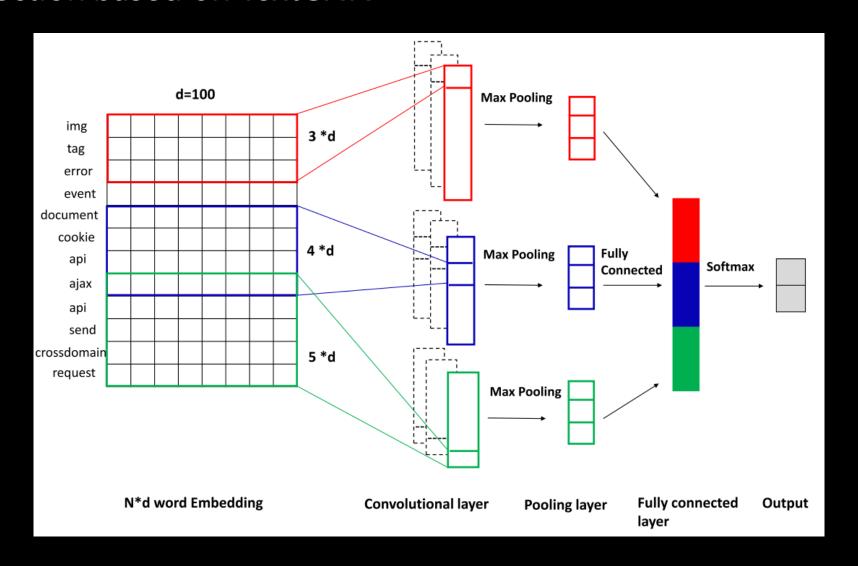
3)Pooling layer and fully connected layer

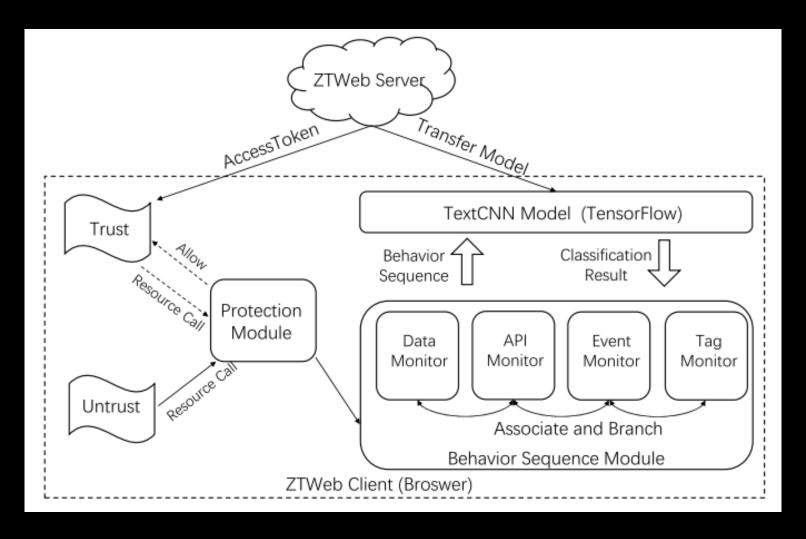
$$t' = \max(t)$$

pooling layer adopts the max pooling, and fully connected layer splices each feature into a length of 3 \* m.

4)Output layer: Softmax function

XSS detection based on TextCNN







# Experiment

## Enviroment

- Win10 64-bit operating system
- i5-1135G7,8-core processor
- 16GB RAM
- Python 3.9
- keras2.9.0

## Experiment

- Dataset
   (XSS attack behavior sequences)
  - 1. XSS Filter Evasion Cheat Sheet (PortSwigger Research, 2022)
  - 2. XSS Payload Dataset (Payloadbox, 2020)

The dataset includes 954 benign and 3365 malicious samples 80% of which are used to train the model and 20% to test.

| Confusion matrix. |            |                |  |  |
|-------------------|------------|----------------|--|--|
|                   | Actual XSS | Actual Non-XSS |  |  |
| Predicted XSS     | TP         | FP             |  |  |
| Predicted Non-XSS | FN         | TN             |  |  |
|                   |            | _              |  |  |

| Result of detection by TextCNN. |                  |               |           |  |
|---------------------------------|------------------|---------------|-----------|--|
| Accuracy                        | Weight Precision | Weight Recall | Weight F1 |  |
| 0.997093                        | 0.997107         | 0.997093      | 0.997091  |  |

## Experiment

Call 1000 times in the trust and untrusted domains, respectively, and calculate the average time.

| Script delay.  |              |                  |               |  |  |  |
|----------------|--------------|------------------|---------------|--|--|--|
| Type           | ZTWeb        |                  | Without ZTWeb |  |  |  |
|                | Trust Domain | Untrusted Domain |               |  |  |  |
| innerHTML      | 0.123 ms     | 0.198 ms         | 0.014 ms      |  |  |  |
| read cookie    | 0.109 ms     | 0.186 ms         | 0.039 ms      |  |  |  |
| document.write | 0.181 ms     | 0.332 ms         | 0.023 ms      |  |  |  |
| ajax           | 2.139 ms     | 11.139 ms        | 1.313 ms      |  |  |  |
|                |              |                  |               |  |  |  |



## Conclusions

## Conclusion

- 1. The model intercepts the attack surfaces penetration by isolating the protected surface, which guarantees the developer codes resource authorization
- 2. TextCNN model to identify whether the behavior sequence is an XSS attack to adjust the resource authorization of untrusted domains dynamically.
- 3. (Future) modify the browser kernel to monitor the behavior in the domain more comprehensively and accurately

# Q & A