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

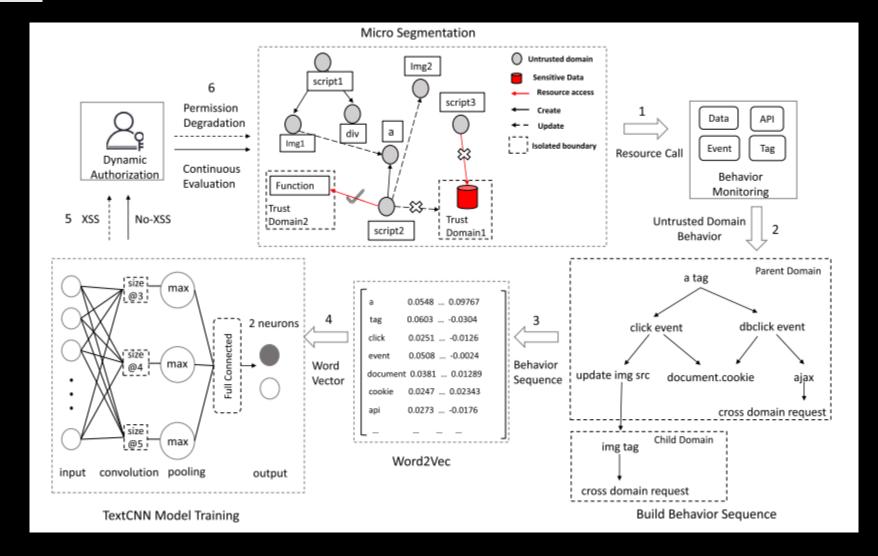
The static defense policy is difficult to adapt to the dynamic change of attack means.

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.



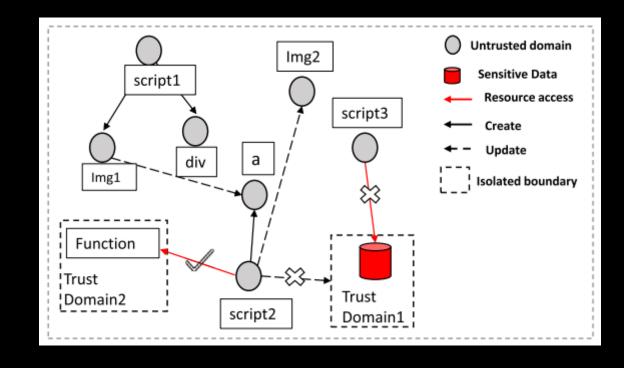
Methods & Implementation



1. Micro-segmentation

Trust Domain1:

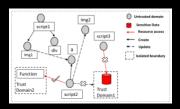
The essence of policy-based XSS defense technology is to authorize the user of sensitive resources.



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

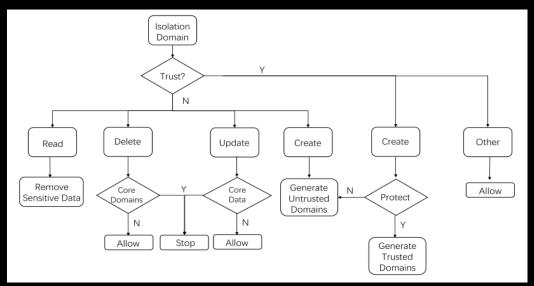
does not allow sensitive data exposure to other domains.

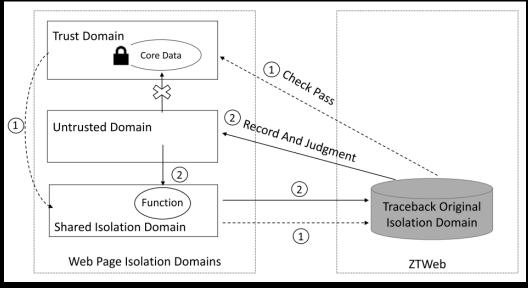
(3) Preventing penetration

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

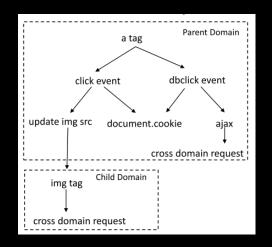
Continuous trust evaluation and dynamic authorization to authorize resources for untrusted domain

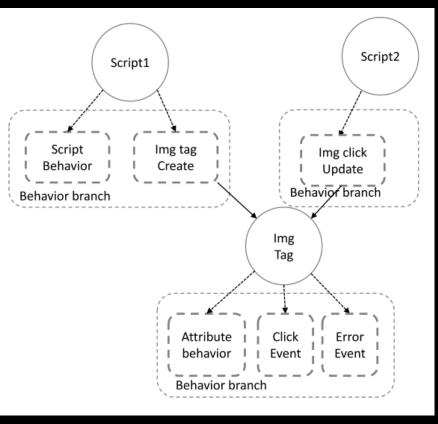
- Untrusted domain as the basic monitoring unit
- Continuously records its resource access behavior

```
<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

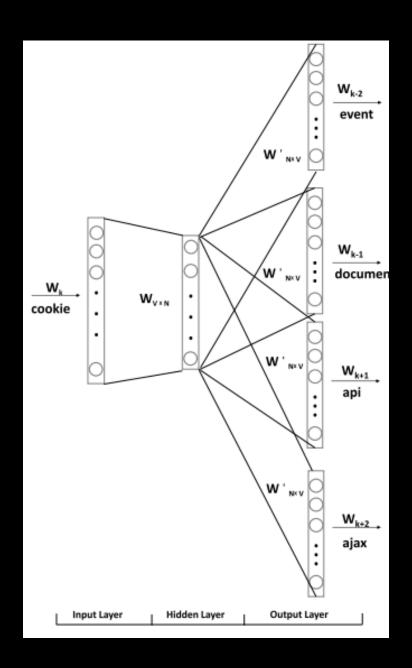
The word embedding model can convert the behavior sequence into a word vector.

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.

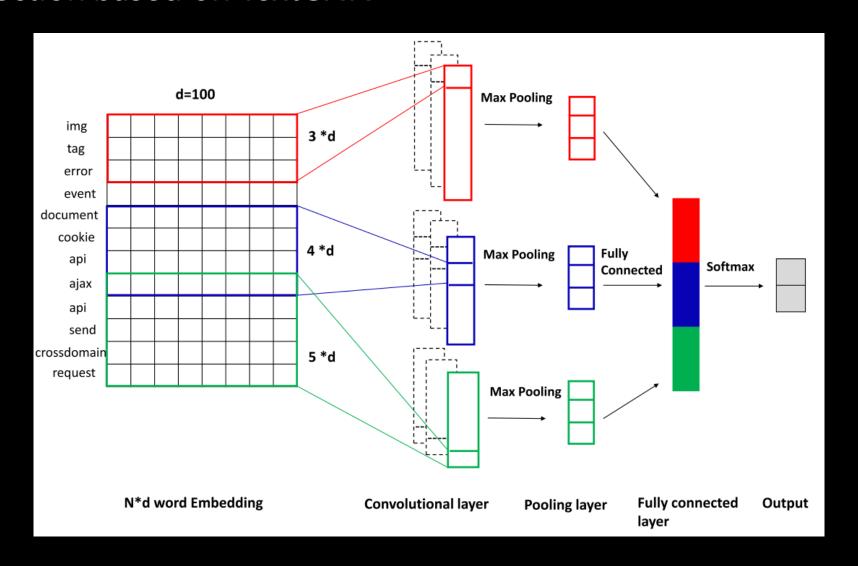
3)Pooling layer and fully connected layer

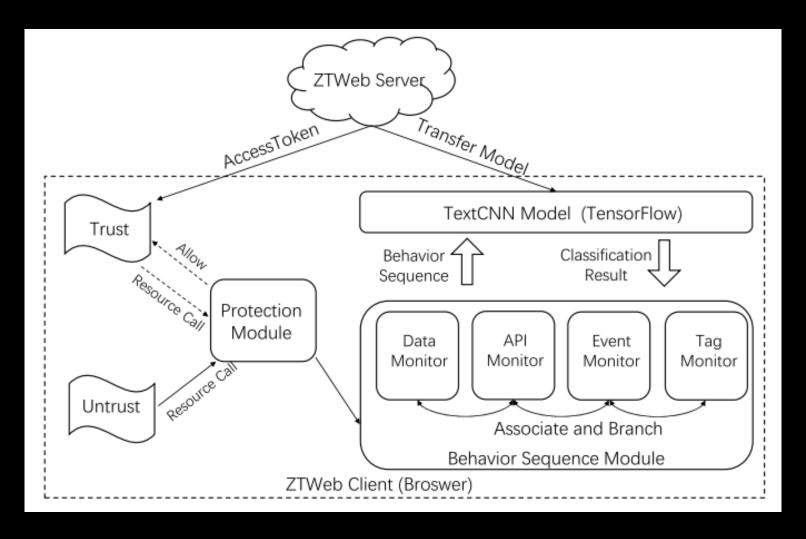
$$t' = \max(t)$$

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