Conditional Probability Voting Algorithm Based on Heterogeneity of Mimic Defense System

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About the paper

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Abstract

- In recent years network attacks have been increasing rapidly, and it is difficult to defend against these attacks, especially attacks at unknown vulnerabilities or backdoors.
- As a novel method, Mimic defense architecture has been proposed to solve these cyberspace security problems by using the principle of dynamic heterogeneous redundant variants.
- Choosing appropriate variants and voting algorithm according to heterogeneities of these variants become the key issue of designing mimic defense architecture.
- This paper analyzes the system failure probability and scalability of 3 different voting algorithms- MHA, MVA, CPVA, and decide which one is the best.

Variants

Mimic defense system can be considered a restrict version of N-variant systems, because it adopts the basic idea of running multiple variants of the same program in parallel.

- Variants are usually composed of a series of components, such as CPU, operating system, middleware, application, etc.
- Each component is composed of several modules. For example, the application can be divided into module 1, module 2,..., module M, etc.
- A module is atomic, and its implementations are different from each other.
- **3** Each variant can be represented by a module implementation vector $z^i = (g_1{}^i g_2{}^i ... g_N{}^i)$, where N is the number of modules contained in a variant and each module may have several implementations.

Variants

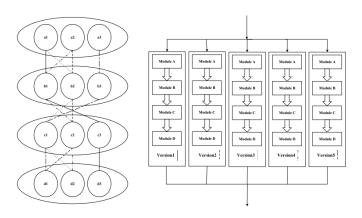


Figure: A typical mimic defense instance

Variants

The mimic defense system shown in above figure can be described by a matrix as shown below.

$$\begin{pmatrix} z^1 \\ z^2 \\ z^3 \\ z^4 \\ z^5 \end{pmatrix} = \begin{pmatrix} a_1 & b_1 & c_3 & d_3 \\ a_2 & b_2 & c_2 & d_1 \\ a_1 & b_2 & c_1 & d_2 \\ a_2 & b_1 & c_2 & d_2 \\ a_3 & b_3 & c_1 & d_1 \end{pmatrix}$$

Heterogeneous Variants

- Mimic defense system requires the variants to be heterogeneous to each other, not just applications, but also including CPU, OS, middleware and so on.
- For a large system which is common in mimic defense system, it is hard to realize totally heterogeneous.

There are mainly three kinds of algorithms to choose variants:

- maximum heterogeneous algorithm (MHA)
- optimal mean distance algorithm (OMDA)
- 3 random seeds scheduling method

Heterogeneous Variants

2-level similarity

Suppose there are three variants (1, 2, 3), the 2-level similarities are referred to the similarities for 1&2,1&3,2&3. The sum of similarities is lower, the system is considered to be safer.

- As the number of working variants grows, 2-level similarity become less important.
- Heterogeneity will reach max when there are 3 variants in the mimic system, and the heterogeneity will drop as variants number increase more than 3.

Classic Model

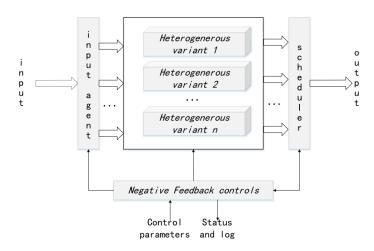


Figure: Classic model of mimic defense architecture

Assumptions

- Input agent, Scheduler, and Feedback controller are safe from network attacks.
- Each vulnerability/backdoor of the system have the same probability being attacked.
- When the high-level vulnerability/backdoor is attacked, the variants who share the vulnerability will generate the same output.
- Only one vulnerability / back door can be attacked at a time, and the attacked variant will be cleaned in a short time.

The probability of successfully attacking each module is β_i , which should satisfy $\sum_i \sum_{\psi_i} \beta_i = 1$, ψ_i is the number of implementations for module i.

Binary Division Vectors

Module diversity ψ_i

The implementation number of the module g_i , which can be calculated by formula $|\bigcup_i g_k{}^i|$. Implementation set is $(a_1a_1a_1a_2a_3)^T$, then union the contents and get the set $\{a_1a_2a_3\}$, which contain 3 elements, so the diversity of module a is 3.

Binary division vector η_i^k

- In a module, divide the same implementations into one group and other implementations into another group, and use a vector to represent. Assign corresponding values in the vector of the same implementations to 1 and others to 0.
- ② For example, the implementation of g_1 is $(a_1a_1a_1a_2a_3)^T$, then there are 3 binary division vectors for module g_1 , the binary division vector of a_2 is $(00010)^T$, a_3 is $(00001)^T$, a_1 is $(11100)^T$.

Binary Division Vectors

Complement binary division vector $\sim \eta_i^k$

which is reversing every element in the binary division vector. Based on the binary division vector $(11100)^T$ of module implementation a_1 , reverse all the elements in it, and its complement vector will be $(00011)^T$.

Isomorphic number of binary division vector $\lambda_i^{\ k}$

The number of elements whose value is equal to 1. There are three 1 in the binary division vector of module implementation a_1 , which is $(11100)^T$, then there are three a_1 , and Isomorphic number of $(11100)^T$ is 3.

Maximum heterogeneous algorithm

- 1 It is based on 2-level similarity.
- 2 system failure probability = $\frac{\text{no. of similarities}}{\text{no. of vulnerabilities}}$.

3 system failure probability = $\frac{0+1+1+0+1+2+1+1+1+0}{9} = \frac{8}{12} = \frac{3}{4}$

Important Property

- **1** If there are T executions in the mimic defense system, there will be at least 1 implementation whose isomorphic number is not less than $\left|\frac{T+1}{2}\right|$.
- ② In order to reduce the failure probability of mimic defense system, add different implementation of a module or balance the same implementation of a module so that its maximum implementation is less than $\left|\frac{T+1}{2}\right|$.

Majority voting algorithm

- ① Divide the variants by their results, put variants with the same result into a group G_k . According to the hypothesis only one vulnerability/backdoor is attacked at a time, so there are usually 2 groups, suppose they are G_1 and G_2 .
- ② If $|G_1| > |G_2|$, then select the result of G1 as the final output; otherwise, select the result of G2 as the final output.
- 3 Clean the variants which have been arbitrated to be abnormal.

Majority voting algorithm

- $\mathbf{0} V_{g} = \mathsf{NULL}$
- \bigcirc for i = 1: N
- \bullet for k=1: ψ_i
- \odot add i in V_g
- endfor
- endfor
- $oldsymbol{0}$ for each index in U_g
- $\mathbf{0}$ msum = msum + β_k
- endfor

Conditional probability voting algorithm

- **1** The variants which generated the same results are divided into one group G_k , generally there are only two groups, assumed as G_1 and G_2 .
- ② If there is the same implementation of one module in both G_1 and G_2 , the same module implementation in the G_1 and G_2 need to be removed, then we can get two eliminated sets G_1' and G_2' , that is $(z^i \bigcap_{i \in G_1} z^j)$ and $(z^i \bigcap_{i \in G_1} z^j)$.
- If there are multiple implementations of one module in G_1' or G_2' , we use intersection to eliminate different module implementations, we can get two eliminated sets G_1'' and G_2'' , that is $\bigcap\limits_{i\subset G_1}(z^i-\bigcap\limits_{j\subset G_2}z^j)$ and

$$\bigcap_{i\subset G_2}(z^i-\bigcap_{j\subset G_1}z^j).$$



Conditional probability voting algorithm

- **3** Calculation β_{G_1} and β_{G_2} , $\beta_{G_1} = \sum_k \beta_k$, if $k \in \bigcap_{i \subset G_1} (z^i \bigcap_{j \subset G_2} z^j)$, $\beta_{G_2} = \sum_k \beta_k$, if $k \in \bigcap_{i \subset G_2} (z^i \bigcap_{j \subset G_1} z^j)$.
- **1** If $\beta_{G_1} > \beta_{G_2}$, the result of G_2 shall be used, otherwise, the result of G_1 shall be used.
- 6 Clean the abnormal variants which have generated wrong result.

Conditional probability voting algorithm

$$\mathbf{0}$$
 $G_{\prime\prime} = \text{NULL}$

$$\bigcirc$$
 for $i = 1$: N

$$\square$$
 Isum = Isum + β_k

$$\odot$$
 for k = 1: ψ_i

$$\bullet$$
 for each k in G_S

$$G_{\mu} = \text{union}(G_{\mu}, \eta_i^k)$$

add i in
$$G_L$$
solution else if(\sim vctorl== η_i^k)

ssum = ssum +
$$\beta_k$$
 and endfor

$$\mathbf{G}_{u} = \operatorname{dinon}(\mathbf{G}_{u}, \eta_{i})$$

$$\mathbf{G}_{u} = \operatorname{dinon}(\mathbf{G}_{u}, \eta_{i})$$

$$\odot$$
 add i in G_S

$$\circ$$
 csum = 0

$$\mathbf{0}$$
 csum = csum + ssum

$$\mathbf{G}_L = \mathbf{G}_S = \mathsf{NULL}$$

• for each vctorl in
$$G_u$$
 • for each k in G_L

Results for 3-variants experiment

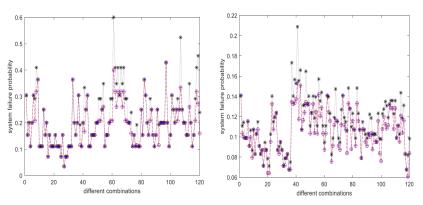


Figure: System failure probabilities of MHA, MVA and CPVA when N=10 M=5 and N=100 M=10

Results for 5-variants experiment

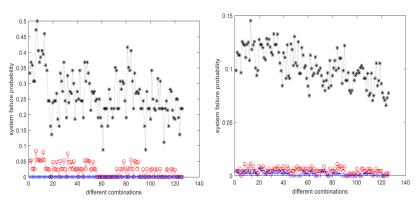


Figure: System failure probabilities of MHA, MVA and CPVA when N=10 M=10 and N=100 M=10

Results for scalability experiment

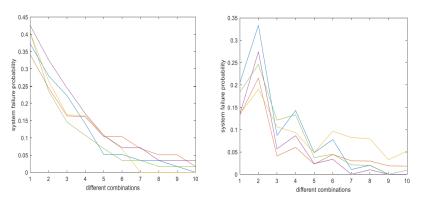


Figure: System failure probabilities of CPVA and MVA with variants increase

Performance analysis

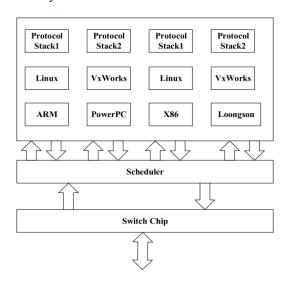


Figure: Mimic defense architecture for control panel of ethernet switch.