# The pending schema of Combinatorial Testing

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Combinatorial testing (CT) aims to detect the failures which are triggered by the interactions of various factors that can influence the behaviour of the system, such as input parameters, and configuration options. Many studies in CT focus on designing an elaborate test suite (called covering array) to reveal such failures. Although covering array can assist testers to systemically check each possible factor interaction, however, it provides weak support to locate the failure-inducing interactions, i.e., the Minimal Failure-causing Schemas (MFS). Recently some elementary researches are proposed to handle the MFS identification problem. However, we argue that many of them are still incomplete in terms of the existence of schemas that cannot be determined to be faulty or not yet. These cannot-be-determined schemas, i.e., the pending schemas, would be hidden dangers to the software under testing. Hence, it is important to obtain these pending schemas for these incomplete MFS identification approaches. In this paper, we proposed several propositions to formulate the set of pending schemas and give three equivalent formulas to obtain them, based on which we reduce the complexity of obtaining pending schemas from  $O(2^n)$  to  $O(\tau^{|FSS^{\perp}|+|HSS^{\perp}|})$ , where n is the number of factors in the software, while  $|FSS^{\perp}|$  and  $|HSS^{\top}|$  are two relatively small numbers and independent on the number of n. We conduct a series empirical studies on some real software systems with various number of parameters and values. Our results shows that the incompleteness in very common in the covering arrays and MFS identification approaches. We also observed that the third proposed formula is the most efficient when compared others in most cases.

# CCS Concepts: • Software defect analysis → Software testing and debugging;

Additional Key Words and Phrases: Pending Schema, Minimal Failure-causing Schema, Combinatorial Testing, Software Testing

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id	Highlight	Status bar	Bookmarks	Smart tags	Outcome
1	On	On	On	On	PASS
2	Off	Off	On	On	PASS
3	On	Off	Off	On	PASS
4	On	Off	On	Off	PASS
5	Off	On	Off	Off	Fail

Table 1. MS word example

#### 1 INTRODUCTION

The behavior of modern software is affected by many factors, such as input parameters, configuration options, and specific events. To test such software system is challenging, as in theory we should test all the possible interaction of these factors to ensure the correctness of the System Under Test (SUT)[16]. When the number of factors is large, the interactions to be checked increase exponentially, which makes exhaustive testing not feasible. Combinatorial testing (CT) is a promising solution to handle the combinatorial explosion problem [6, 7]. Instead of testing all the possible interactions in a system, it focuses on checking those interactions with number of involved factors no more than a prior number. Many studies in CT focus on designing a elaborate test suite (called covering array) to reveal such failures. Although covering array is effective and efficient as a test suite, it provides weak support to distinguish the failure-inducing interactions, i.e., Minimal Failure-causing schemas(MFS), from all the remaining interactions (schemas) [2, 9].

Consider the following example [1], Table 1 presents a pair-wise covering array for testing an MS-Word application in which we want to examine various pair-wise interactions of options for 'Highlight', 'Status Bar', 'Bookmarks' and 'Smart tags'. Assume the last test case failed. We can get five pair-wise suspicious schemas that may be responsible for this failure. They are respectively (Highlight: Off, Status Bar: On), (Highlight: Off, Bookmarks: Off), (Highlight: Off, Smart tags: Off), (Status Bar: On, Bookmarks: Off), (Status Bar: On, Smart tags: Off), and (Bookmarks: Off, Smart tags: Off). Without additional information, it is difficult to figure out the specific schemas in this suspicious set that caused the failure. In fact, considering that the schemas consist of other number of factors could also be MFS, e.g., (Highlight: Off) and (Highlight: Off, Status Bar: On, Smart tags: Off), the problem becomes more complicated. Generally, to definitely determine the MFS in a failing test case of n factors, we need to check all the  $2^n - 1$  interactions in this test case, which is not possible when n is a large number.

To address this problem, prior work [12] specifically studied the properties of MFS in SUT, based on which additional test cases were generated to identify them. Other approaches to identify the MFS in SUT include building a tree model [18], adaptively generating additional test cases according to the outcome of the last test case [22], ranking suspicious schemas based on some rules [5], and using graphic-based deduction [9], among others. These approaches can be partitioned into two categories [2] according to how the additional test cases are generated: *adaptive*–additional test cases are chosen based on the outcomes of the executed tests [5, 8, 12–15, 17, 22]or *nonadaptive*–additional test cases are chosen independently and can be executed in parallel [2, 9, 10, 18, 21].

Although many efforts have been devoted to identify the failure-causing schemas from failing test cases, we argue that many of them are still incomplete in terms of the existence of schemas that cannot be determined to be faulty or not yet. Particularly, after identifying the MFS from one failing test case, we wonder that does the schemas other than the identified MFS are guaranteed to be irrelevant to the failure in this failing test case? A related question is that, after identifying the MFS, is there exists any schema in this failing test case that is still cannot be determined to be

faulty or not? To answer these two questions is important, because these cannot-be-determined schemas would be hidden dangers to the SUT. Moreover, we need the measures to evaluate the adequacy of the covering arrays and MFS identification approaches in CT, which is a important key to form the confidence of the developer of the SUT. However, to our best knowledge, no such study has been proposed, especially from a theoretical view.

One simple solution is to exhaustively list all the schemas in one failing test case, and then check them to be faulty or not one by one. However, as we have mentioned before, the complexity of this procedure is  $2^n$ , where n is the number of factors in this test case. Hence, this solution is far from feasible when n is very large.

For all of these, a metric should be proposed to assist in evaluating the completeness of MFS identification approaches, and it should be more efficient than a simple exhaustive testing. In this paper, we proposed the notion of **pending schemas**, which indicates the schemas that cannot be determined to be faulty or not. By calculating the number of pending schemas in one failing test case, we can easily assess the extent to which the MFS identification approaches are incomplete. In fact, by the use of pending schemas, we can also evaluate the incompleteness of traditional covering arrays.

Furthermore, we theoretically analyzed the relationships among schemas by proposing nine novel propositions. Based on them, we gave three equivalent formulas, but with different complexities, to obtain the pending schemas. Among these formulas, Formula 3 helps to reduce the complexity of obtaining pending schemas from  $O(2^n)$  to  $O(\tau^{|FSS^{\perp}|+|HSS^{\top}|})$ , where  $\tau$  is the number of parameter values in the MFS, and  $|FSS^{\perp}|$  and  $|HSS^{\top}|$  are two relatively small numbers and independent on the number of parameters n in one test case. Formula 3 is much more efficient at obtaining pending schemas when compared to the exhaustive methods which consecutively checks schemas in one failing test case, especially when n is large.

We conducted a series empirical studies on some real software systems with various number of parameters and values. We first evaluated the incompleteness of traditional covering arrays and different fault localization approaches in CT. We also compared the efficiency of three formulas in terms of obtaining pending schemas. Our results mainly shows that the incompleteness in very common in the covering arrays and MFS identification approaches. We also observed that Formula 3 is the most efficient formula among others in most cases.

### Contributions of this paper:

- We showed that the traditional covering arrays and the minimal failure-causing schema model are still incomplete in terms of the determination of schemas to be faulty or healthy.
- We introduced the notion of the pending schema to evaluate the incompleteness of these models in combinatorial testing.
- We proposed several propositions to formulate the set of pending schemas and gave three equivalent formulas to obtain the pending schemas, based on which we reduced the complexity of obtaining pending schemas from  $O(2^n)$  to  $O(\tau^{|FSS^{\perp}|+|HSS^{\top}|})$ , where  $|FSS^{\perp}|$  and  $|HSS^{\top}|$  are two relatively small numbers and independent on the number of n.
- We conducted a series of experiments to evaluate the incompleteness of traditional covering arrays and MFS identification approaches. Besides, we also evaluated the efficiency of the three formulas on obtaining pending schemas.

The remainder of this paper is organized as follows: Section 2 describes the motivation for this work. Section 3 introduces some preliminary definitions and propositions. Section 4 proposes several important propositions to formally identify the characteristics of the pending schemas. Section 5 evaluates the incompleteness of MFS identification approaches and compares the effectiveness of different approaches for obtaining pending schemas. Section 6 discusses the findings of our research

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works. Section 7 summarizes the related works. Section 8 concludes this paper and discusses the future works.

#### 2 MOTIVATING EXAMPLES

In this section, we will use several examples to show the incompleteness of traditional covering arrays and well-known MFS identification approaches, respectively. These examples are derived from the MS-Word example listed in the introduction. For simplification, we use integer 0 to represent the state On and 1 to represent the state Off for each option. For example, the second test case listed in Table 1 can be denoted as (1, 1, 0, 0). Also, we use the intuitive notation  $(..., v_{n_i}, v_{n_k}, ..., -)$  to represent the schemas for the system, where  $v_{n_i}$  indicate the value that is assigned to the corresponding factor and '-' indicates that the corresponding factor is not in this schema. For example, (1, 1, -, -) represents the schema (Highlight: Off, Status Bar: Off) in this example. Note that we will introduce a more formal denotation of test case and schema in the following section.

Also, to understand the following examples, we first give two rules. The first rule is that all the schemas in a passing test case are non-faulty, i.e., will not cause failure. The second rule is that any schema contain a MFS is a faulty schema, i.e., will also cause the failure. These two rules are wildly used in the MFS identification approaches [5, 12, 13, 22]. We will discuss the justifications of these two rules later, as well as some issues if these two rules are not hold.

### 2.1 The incompleteness of covering array

We first consider the traditional covering arrays. To understand the incompleteness of the covering array, we need to check each schema in the test case of the covering array. As we said before, we use the same MS-Word example in the first section. Figure 1 lists the detail of this example. The test cases  $t_1$  to  $t_5$  constitute the covering array shown in Table 1. The complete set of the schemas of each test case is attached at the right side of the corresponding test case. For example, for test case  $t_1$ , i.e., (0, 0, 0, 0), all the possible schemas (0, -, -, -), (0, 0, -, -), ..., are listed at the right side of  $t_1$ . There are  $2^4 - 1 = 15$  schemas in total for each test case in this example.

In this figure, the test case with dark color represents a failing test case, while the test case with white color is a passing test case. The schema with white color is non-faulty, i.e., will not cause a failure, while the schema with dark color is faulty, i.e., any test case contain this schema would fail after testing. At last, the schema with light dark color and dashed outline is the pending schema.

In this figure, we can first observe that all the schemas in the passing test case is non-faulty. This result is according to the first rule we mentioned before. The second observation is that the schema with the maximal number of factors (4 factors) in a failing test case is a faulty schema. In fact, this schema is the failing test case itself, i.e., (1, 1, 0, 1). This is because the failing test case must contain at least one MFS (otherwise, it will not fail). Hence, the schema which is the test case itself must also contain at least one MFS. As a result, it must be faulty schema according to the second rule we mentioned before. The last observation is that other schemas in this failing test case are not guaranteed to be faulty schemas. In fact, if we assume this test case only contains one MFS (1, 1, 0, 1), then all the other schemas can be non-faulty schemas. Hence, these schemas cannot be determined to be faulty or not if we focus on this failing test case alone.

# 2.2 The incompleteness of error locating array

Since that , many approaches give a alleviation of how to compute the remaining schemas, which is called the fault localization. Error locating array is one of them. However, this time, we also give .

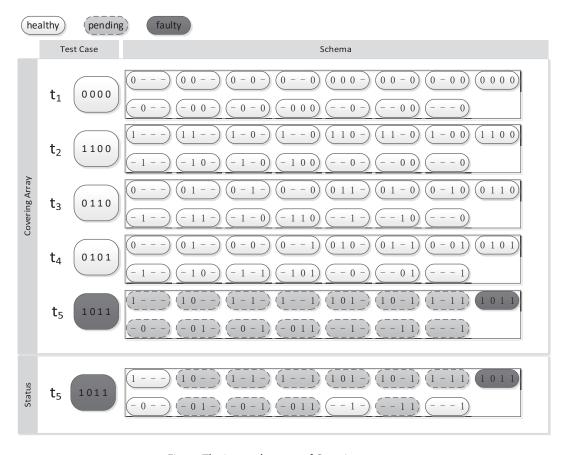


Fig. 1. The incompleteness of Covering array

# 2.3 The incompleteness of OFOT and SOFOT

Why minimal failure-causing schema is not effective? This example shows that . Because there still existing pending schemas.

# 2.4 The incompleteness of FIC and FIC\_BS

Since we cannot obtain a completeness of . It is important to evaluate how incompleteness of covering array, fault localization.

### 2.5 A complete process

# 3 PRELIMINARY

This section presents some definitions and propositions to give a formal model for CT.

Assume that the Software Under Test (SUT) is influenced by a set of parameters P, which contains n parameters, and each parameter  $p_i \in P$  can take the values from the finite set  $V_i$  (i = 1,2,..n).

Definition 3.1. A test case of the SUT is a tuple of n values, one for each parameter of the SUT. It is denoted as  $(v_1, v_2,...,v_n)$ , where  $v_1 \in V_1, v_2 \in V_2 \dots v_n \in V_n$ .

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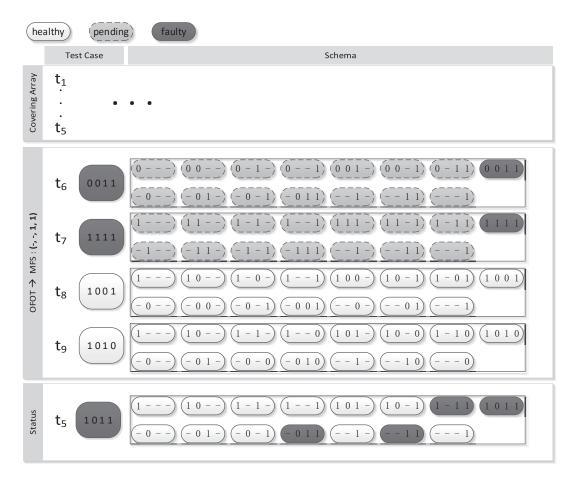


Fig. 2. OFOT with single MFS

In practice, these parameters in the test case can represent many factors, such as input variables, run-time options, building options or various combination of them. We need to execute the SUT with these test cases to ensure the correctness of the behaviour of the SUT.

We consider any abnormally executing test case as a *fault*. It can be a thrown exception, compilation error, assertion failure or constraint violation. When faults are triggered by some test cases, it is desired to figure out the cause of these faults.

Definition 3.2. For the SUT, the  $\tau$ -set  $\{(p_{x_1}, v_{x_1}), (p_{x_2}, v_{x_2}), ..., (p_{x_\tau}, v_{x_\tau})\}$ , where  $0 \le x_i \le n$ ,  $p_{x_i} \in P$ , and  $v_{x_i} \in V_{x_i}$ , is called a  $\tau$ -degree schema  $(0 < \tau \le n)$ , when a set of  $\tau$  values assigned to  $\tau$  distinct parameters.

For example, the interactions (Highlight: Off, Status Bar: On, Smart tags: Off) appearing in Section 1 is a 3-degree schema, where three parameters are assigned to corresponding values. In effect a test case itself is a n-degree *schema*, which can be described as  $\{(p_1, v_1), (p_2, v_2), ..., (p_n, v_n)\}$ .

Note that the schema is a formal description of the interaction between parameter values we discussed before.

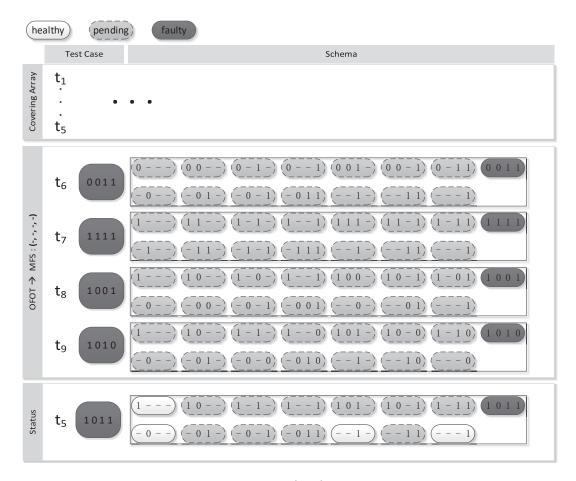


Fig. 3. OFOT with multi MFS

Definition 3.3. Let  $c_1$  be a l-degree schema,  $c_2$  be an m-degree schema in SUT and l < m. If  $\forall e \in c_1, e \in c_2$ , then  $c_1$  is the sub-schema of  $c_2$ , and  $c_2$  the super-schema of  $c_2$ , which can be denoted as  $c_1 < c_2$ .

For example, the 2-degree schema {(Highlight, Off), (Status Bar, On)} is a sub-schema of the 3-degree schema {(Highlight, Off), (Status Bar, On), (Smart tags, Off)}.

Definition 3.4. If for any test cases that contain a schema, say c, it will trigger a failure, then we call this schema c the faulty schema. Additionally, if none of sub-schema of c is a faulty schema, we then call the schema c the minimal failure-causing schema (MFS) [12].

Note that MFS is identical to the failure-inducing interaction discussed previously. In this paper, the terms *failure-inducing interactions* and *MFS* are used interchangeably. Figuring the MFS helps to identify the root cause of a failure and thus facilitate the debugging process.

Definition 3.5. A schema, say, c, is called a *healthy schema* when we find at least one passing test case that contains this schema. In addition, if none of super-schema of c is the *healthy schema*, we then call the schema c the *maximal healthy schema* (MHS).

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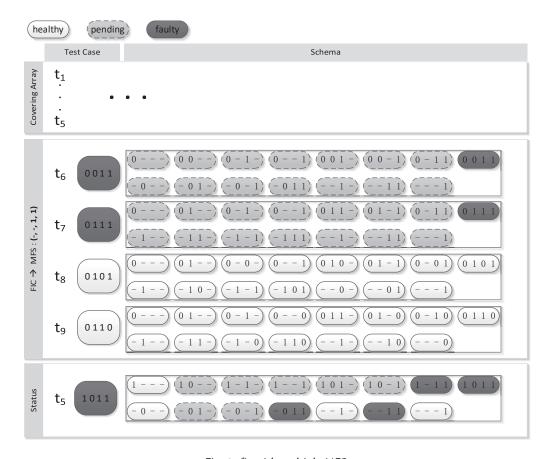


Fig. 4. fic with multiple MFS

These two type of schemas, i.e., MFS and MHS, are the keys to our approach, as they are essentially representations of the healthy schemas and faulty schemas in a test case. As shown later, other schemas can be determined to be healthy or faulty by these two type of schemas. As a result, we just need to record these two types of schemas (normally a small amount) instead of recording all the schemas in a test case (up to  $2^n$ ) when identifying MFS.

Definition 3.6. A schema is called a *pending* schema, if it is not yet determined to be *healthy* schema or *faulty* schema.

The *pending* schema is actually the *un-determined* schema discussed in Section 1. In effect, to identify the MFS in a failing test case is to figure out all the *pending* schemas, and then try to classify each of them into *healthy* or *faulty* schema. After that, the MFS can be selected from those *faulty* schemas by definition.

To facilitate our discussion, we introduce the following assumptions that will be used throughout this paper:

Assumption 1. The execution result of a test case is deterministic.

This assumption is a common assumption of CT[5, 13, 22]. It indicates that the outcome of executing a test case is reproducible and will not be affected by some random events. Some

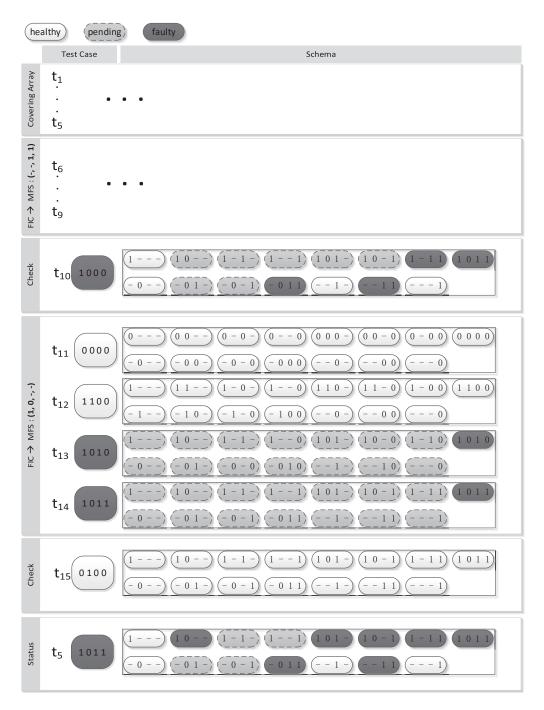


Fig. 5. finovlp with multiple MFS

approaches have already proposed measures to handle this problem, e.g., studies in [3, 18] use multiple covering arrays to avoid this problem.

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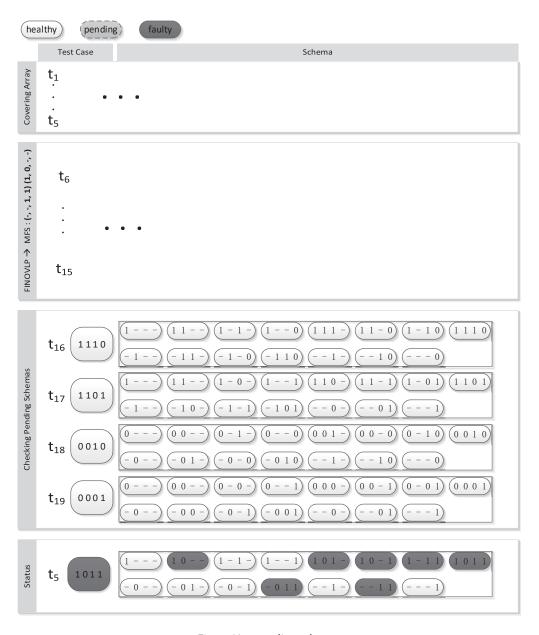


Fig. 6. Non pending schema

Assumption 2. If a test case contains a MFS, it must fail as expected.

This assumption shows that we can always observe the failure caused by the MFS. In practice, some issues may prevent this observation. For example, the coincidental correctness problem [11] may happen through testing, when the faulty-code is executed but the failure doesn't propagate to the output. Masking effect [19] may also make the failure-observation difficult, as other failure may triggered and stop the program to go on discovering the remaining failures.

We will later discuss the impacts on MFS identification from these two assumptions, as well as how to alleviate them. Based on these definitions and assumptions, we can get several propositions as following. These propositions are the foundation of our approach, and their proofs are omitted due to their simplicity.

PROPOSITION 3.7. Given schemas  $s_1$ ,  $s_2$ , and  $s_3$ , if  $s_1 < s_2$ ,  $s_2 < s_3$ , then  $s_1 < s_3$ .

Proposition 3.8. Given a faulty schema  $s_1$ , then  $\forall s_2, s_1 < s_2, s_2$  is a faulty schema.

Proposition 3.9. Given a healthy schema  $s_1$ , then  $\forall s_2, s_2 \prec s_1, s_2$  is a healthy schema.

You should give normal sentences description and examples of them (if possible, give some simple proofs)

#### 4 PENDING SCHEMA

In this section, we will describe our approach to identify the failure-inducing schemas in the SUT. To give a better description, we will give several propositions which provide theoretical supports for our approach. The proofs of these propositions are given in the Appendix.

# 4.1 What is pending schema

Through the motivating examples, we can learn that the pending schema is not checked to be healthy and faulty. That is,

However, to emulate all the faulty schemas and all the healthy schemas is very large. By the propostios 3 and 4,

More formally, we can learn that it should not be any super-schemas of existing faulty schemas, and any sub-shemas of existing healthy schemas.

PROPOSITION 4.1. Given two pending schemas  $s_1$ ,  $s_2$ , and  $s_1 < s_2$ . Then  $\forall s_3, s_1 < s_3 < s_2$ ,  $s_3$  is a pending schema.

# 4.2 Obtaining pending schema

To identify the MFS in a failing test case, we need to figure out all the pending schemas and then classify them into faulty or healthy schemas. For this, we firstly need to be able to find one pending schema and check it. Then, we repeat this procedure until no more pending schema can be obtained.

To be general, we formalize the problem of obtaining one pending schema as the following problem:

Given a failing test case  $t = \{(p_1, v_1), (p_2, v_2), ..., (p_n, v_n)\}$ , a set of faulty schemas  $FSS = \{c_1, c_2, ...c_i, ...\}$ , where  $c_i < t$  or  $c_i = t$  (That is,  $c_i \le t$ ), a set of healthy schemas  $HSS = \{c_1, c_2, ...c_i, ...\}$ , where  $c_i \le t$ , and  $FSS \cap HSS = \emptyset$ . The goal is to find one pending schema.

Note that at the beginning of MFS identification, if there is no additional information, *FSS* will be initialized to have one element, i.e., *t* itself, and *HSS* is an empty set.

We will settle this problem step by step. According to the Propositions 3.8 and 3.9, it is easy to find that, the pending schemas set PSS should be the following set

$$PSS = \{c | c < t \&\& \forall c_f \in FSS, c_f \nleq c\&\& \forall c_h \in HSS, c \nleq c_h\}. \tag{1}$$

To obtain one pending schema, we just need to select one schema which satisfies  $c \in PSS$ . However, to directly utilize Formula 1 is not practical to obtain one pending schema, because in the worst case it needs to check every schema in a test case t, of which the complexity is  $O(2^n)$ . Hence, we need to find another formula which is equivalent to Formula 1, but with much lower complexity.

For this purpose, we defined the following two sets, i.e. CMXS and CMNS.

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Definition 4.2. For a k-degree faulty schema  $c = \{(p_{x_1}, v_{x_1}), (p_{x_2}, v_{x_2}), ..., (p_{x_k}, v_{x_k})\}$ , a failing test case  $t = \{(p_1, v_1), (p_2, v_2), ..., (p_n, v_n)\}$ , and  $c \le t$ . We denote the candidate maximal pending schema set as CMXS(c,t)=  $\{t \setminus (p_{x_i}, v_{x_i}) \mid (p_{x_i}, v_{x_i}) \in c\}$ .

Note that CMXS is the set of schemas that remove one distinct factor value in c, such that all these schemas will not be the super-schema of c. For example assume the failing test case  $\{(p_1, v_1), (p_2, v_2), (p_3, v_3), (p_4, v_4)\}$ , and a faulty schema  $\{(p_1, v_1), (p_3, v_3)\}$ . Then the CMXS set is  $\{(p_2, v_2), (p_3, v_3), (p_4, v_4)\}$ ,  $\{(p_1, v_1), (p_2, v_2), (p_4, v_4)\}$ . Obviously, the complexity of obtaining CMXS of one faulty schema is  $O(\tau)$ , where  $\tau$  is the number of parameter values in this faulty schema, i.e., the degree of this schema.

With respect to the CMXS set of a single faulty schema, we can get the following proposition:

PROPOSITION 4.3. Given a faulty schema  $c_1$ , a failing test case t, where  $c_1 \le t$ , we have  $\{c \mid c < t \&\& c_1 \not \le c\} = \{c \mid \exists c_1' \in CMXS(c_1, t), c \le c_1'\}$ .

This proposition means that the pending schema must be the sub-schema or equal to the schemas in CMXS, otherwise, it is a faulty schema.

In the equation of Proposition 4.3, the schemas of the left side, i.e.,  $\{c \mid c \leq t \&\& c_1 \nleq c\}$ , are the sub-schemas of test case t, but not the super-schemas of faulty schema  $c_1$  nor equal to  $c_1$ . Note that the pending schemas can only appear in this set, because any schema that is not belong to this set is faulty schema according to Proposition 3.8. The right side set in this equation, i.e.,  $\{c \mid \exists c_1' \in CMXS(c_1,t), c \leq c_1'\}$ , are schemas which are sub-schemas of or equal to at least one schema in  $CMXS(c_1,t)$ . Proposition 4.3 indicates that these two schema sets are equivalent. As an example, considering a failing test case  $t - \{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}$ , and a faulty schema  $c_f - \{(p_3, 1)\}$ . Table 2 shows the schema set  $\{c \mid c < t \&\& c_f \nleq c\}$ ,  $CMXS(c_f, t)$  and  $\{c \mid \exists c_1' \in CMXS(c_f, t), c \leq c_1'\}$ .

Test case t		$CMXS(c_f, t)$	$\left\{c \mid \exists c_1' \in CMXS(c_f, t), c \leq c_1'\right\}$
$\{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}$	$\{(p_1, 1), (p_2, 1), (p_4, 1)\}$	$\{(p_1, 1), (p_2, 1), (p_4, 1)\}$	$\{(p_1, 1), (p_2, 1), (p_4, 1)\}$
Faulty schema c <sub>f</sub>	$\{(p_1, 1), (p_2, 1)\}$		$\{(p_1, 1), (p_2, 1)\}$
$\{(p_3, 1)\}$	$\{(p_1, 1), (p_4, 1)\}$		$\{(p_1, 1), (p_4, 1)\}$
	$\{(p_2, 1), (p_4, 1)\}$		$\{(p_2, 1), (p_4, 1)\}$
	$\{(p_1, 1)\}$		$\{(p_1, 1)\}$
	$\{(p_2, 1)\}$		$\{(p_2, 1)\}$
	$\{(p_4, 1)\}$		$\{(p_4, 1)\}$

Table 2. An example of Proposition 4.3

We can extend this conclusion to a set of faulty schemas. For this, we need the following notation: For two faulty schemas  $c_1, c_2$ , and a failing test case t ( $c_1 \le t, c_2 \le t$ ), let  $CMXS(c_1, t) \land CMXS(c_2, t) = \{c \mid c = c'_1 \cap c'_2, \text{ where } c'_1 \in CMXS(c_1, t), \text{ and } c'_2 \in CMXS(c_2, t)\}.$ 

For example, let  $t = \{(p_1, v_1), (p_2, v_2), (p_3, v_3)\}$ ,  $c_1 = \{(p_1, v_1), (p_2, v_2)\}$ ,  $c_2 = \{(p_2, v_2), (p_3, v_3)\}$ . Then we have CMXS  $(c_1, t) = \{\{(p_1, v_1), (p_3, v_3)\}, \{(p_2, v_2), (p_3, v_3)\}\}$ ,  $CMXS(c_2, t) = \{\{(p_1, v_1), (p_2, v_2)\}, \{(p_1, v_1), (p_3, v_3)\}\}$ , and  $CMXS(c_1, t) \land CMXS(c_2, t) = \{\{(p_1, v_1)\}, \{(p_1, v_1), (p_3, v_3)\}\}$ . It is easy to know the complexity of obtaining CMXS of two faulty schemas is  $O(\tau^2)$ , where  $\tau$  is the number of parameter values in the faulty schema. Based on this, we denote CMXS(FSS, t) for a set of faulty schemas.

Definition 4.4. Given a failing test case  $t = \{(p_1, v_1), (p_2, v_2), ..., (p_n, v_n)\}$ , and a set of faulty schemas FSS =  $\{c_1, c_2, ...c_i, ...\}$ , where  $c_i \le t$ , we denote the candidate maximal pending schema of this set as CMXS(FSS,t)=  $\bigwedge_{c_i \in FSS} CMXS(c_i, t)$ .

Note to compute the CMXS of a set of faulty schema, we just need to sequentially compute the CMXS of two faulty schemas until the last schema in this set is computed. Hence, the complexity of obtaining CMXS of a set of faulty schema is  $O(\tau^{|FSS|})$ , where |FSS| is the number of faulty schemas in the schema set, and  $\tau$  is the degree of the schema. According to Proposition 4.3, we have:

PROPOSITION 4.5. Given a failing test case  $t = \{(p_1, v_1), (p_2, v_2), ..., (p_n, v_n)\}$ , and a set of faulty schemas  $FSS = \{c_1, c_2, ...c_i, ...\}$ , where  $c_i \leq t$ , we have  $\{c \mid c \leq t \&\& \forall c_i \in FSS, c_i \nleq c\} = \{c \mid \exists c_1' \in CMXS(FSS, t), c \leq c_1'\}$ .

Proposition 4.5 extends Proposition 4.3 from a single faulty schema to a set of faulty schemas. As an example, considering a failing test case  $t - \{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}$ , and a set of faulty schemas  $FSS - \{\{(p_3, 1)\}, \{(p_1, 1), (p_2, 1)\}\}$ . Table 3 shows the schema set  $\{c \mid c < t \&\& \forall c_i \in FSS, c_i \nleq c\}$ .

Test case t	$\{\mathbf{c} \mid \mathbf{c} \prec \mathbf{t} \&\& \forall \mathbf{c_i} \in FSS, \mathbf{c_i} \nleq \mathbf{c}\}\$	CMXS(FSS, t)	$\{c \mid \exists c'_1 \in CMXS(FSS, t), c \leq c'_1\}$	
$\{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}$	$\{(p_1, 1), (p_4, 1)\}$	$\{(p_1, 1), (p_4, 1)\}$	$\{(p_1, 1), (p_4, 1)\}$	
Faulty schema set FSS	$\{(p_2, 1), (p_4, 1)\}$	$\{(p_2, 1), (p_4, 1)\}$	$\{(p_2, 1), (p_4, 1)\}$	
$\{(p_3, 1)\}$	$\{(p_1, 1)\}$		$\{(p_1, 1)\}$	
$\{(p_1, 1), (p_2, 1)\}$	$\{(p_2, 1)\}$		$\{(p_2, 1)\}$	
	$\{(p_4, 1)\}$		$\{(p_4, 1)\}$	

Table 3. An example of Proposition 4.5

Next, we give the definition of CMNS.

Definition 4.6. For a k-degree healthy schema c =  $\{(p_{x_1}, v_{x_1}), (p_{x_2}, v_{x_2}), ..., (p_{x_k}, v_{x_k})\}$ , a failing test case t =  $\{(p_1, v_1), (p_2, v_2), ..., (p_n, v_n)\}$ , and c < t. We denote the candidate minimal pending schema set as CMNS(c,t)= $\{\{(p_{x_i}, v_{x_i})\} \mid (p_{x_i}, v_{x_i}) \in t \setminus c\}$ .

Note that CMNS is the set of schemas that are assigned to one distinct factor value that is not in c, such that all these schemas will not be the sub-schema of c. For example assume the failing test case  $\{(p_1, v_1), (p_2, v_2), (p_3, v_3), (p_4, v_4)\}$ , and a healthy schema  $\{(p_1, v_1), (p_3, v_3)\}$ . Then the CMNS set is  $\{\{(p_2, v_2)\}, \{(p_4, v_4)\}\}$ . With respect to the CMNS set of a single healthy schema, we can get the following proposition:

PROPOSITION 4.7. Given a healthy schema  $c_1$ , a failing test case t, where  $c_1 < t$ , we have  $\{c \mid c < t \&\& c \nleq c_1\} = \{c \mid c < t \&\& \exists c_1' \in CMNS(c_1, t), c_1' \leq c\}$ .

This proposition means that the pending schema must be the super-schema or equal to the schemas in CMNS, otherwise, it is a healthy schema.

In the equation of Proposition 4.7, the schemas of the left side, i.e.,  $\{c \mid c \leq t \&\& c \nleq c_1\}$ , are the sub-schemas of test case t, but not the sub-schemas of healthy schema  $c_1$  nor equal to  $c_1$ . It is obvious that the pending schemas can also only appear in this set, because they cannot be the sub-schema of any healthy schema nor equal to them. The right side set in this equation, i.e., $\{c \mid c \leq t \&\& \exists c'_1 \in CMNS(c_1,t), c'_1 \leq c\}$ , are sub-schemas of test case t, and also are the super-schemas of or equal to at least one schema in  $CMXS(c_1,t)$ . Proposition 4.7 indicates that these two schema sets are equivalent. As an example, considering a failing test case  $t - \{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}$ , and a healthy schema  $c_h - \{(p_1, 1), (p_2, 1), (p_3, 1)\}$ . Table 4 shows the schema set  $\{c \mid c \leq t \&\& c_h \nleq c\}$ ,  $CMNS(c_h,t)$  and  $c \mid c \leq t \&\& \exists c'_1 \in CMXS(c_h,t), c'_1 \leq c\}$ .

Similarly, for two healthy schemas  $c_1, c_2$ , and a failing test case t ( $c_1 \le t, c_2 \le t$ ), let  $CMNS(c_1, t) \lor CMNS(c_2, t) = \{c \mid c = c'_1 \cup c'_2, where c'_1 \in CMXS(c_1, t), and c'_2 \in CMXS(c_2, t)\}.$ 

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Test case t	$\left\{c\mid c\prec t\;\&\&\;c\nleq c_{h}\right\}$	$CMNS(c_h, t)$	$ \{c \mid c < t \&\& \exists c'_1 \in CMNS(c_h, t), c'_1 \le c \} $
$\{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}$	$\{(p_4, 1)\}$	$\{(p_4, 1)\}$	$\{((p_4, 1)\}$
Heathy schema c <sub>h</sub>	$\{(p_1, 1), (p_4, 1)\}$		$\{(p_1, 1), (p_4, 1)\}$
$\{(p_1, 1), (p_2, 1), (p_3, 1)\}$	$\{(p_2, 1), (p_4, 1)\}$		$\{(p_2, 1), (p_4, 1)\}$
	$\{p_3, 1\}, (p_4, 1)\}$		$\{p_3, 1\}, (p_4, 1)\}$
	$\{(p_1, 1), (p_2, 1), (p_4, 1)\}$		$\{(p_1, 1), (p_2, 1), (p_4, 1)\}$
	$\{(p_1, 1), (p_3, 1), (p_4, 1)\}$		$\{(p_1, 1), (p_3, 1), (p_4, 1)\}$
	$\{(p_2, 1), (p_3, 1), (p_4, 1)\}$		$\{(p_2, 1), (p_3, 1), (p_4, 1)\}$

Table 4. An example of Proposition 4.7

For example, let  $t = \{(p_1, v_1), (p_2, v_2), (p_3, v_3)\}$ ,  $c_1 = \{(p_1, v_1), (p_2, v_2)\}$ ,  $c_2 = \{(p_2, v_2), (p_3, v_3)\}$ . Then we have  $CMNS(c_1, t) = \{\{(p_3, v_3)\}\}$ ,  $CMNS(c_2, t) = \{\{(p_1, v_1)\}\}$ , and  $CMNS(c_1, t) \lor CMNS(c_2, t) = \{\{(p_1, v_1), (p_3, v_3)\}\}$ . Based on this, we denote CMNS(HSS, t) for a set of faulty schemas.

Definition 4.8. Given a failing test case  $t = \{(p_1, v_1), (p_2, v_2), ..., (p_n, v_n)\}$ , and a set of healthy schemas HSS =  $\{c_1, c_2, ...c_i, ...\}$ , where  $c_i \le t$ , we denote the candidate minimal pending schema of this set as CMNS(HSS,t)=  $\bigvee_{c_i \in HSS} CMNS(c_i, t)$ .

Similar to CMXS(FSS,t), the complexity to obtain CMNS(HSS,t) is  $O(\tau^{|HSS|})$ , where |HSS| is the number of healthy schemas in the schema set, and  $\tau$  is the degree of the schema. With respect to CMNS(HSS,t), we have:

PROPOSITION 4.9. Given a failing test case  $t = \{(p_1, v_1), (p_2, v_2), ..., (p_n, v_n)\}$ , and a set of healthy schemas  $HSS = \{c_1, c_2, ...c_i, ...\}$ , where  $c_i \le t$ , we have  $\{c \mid c < t \&\& \forall c_i \in HSS, c \nleq c_i\} = \{c \mid c < t \&\& \exists c_i' \in CMNS(HSS, t), c_i' \le c\}$ .

Similar to Proposition 4.3 and 4.5, Proposition 4.9 extends Proposition 4.7 from a single healthy schema to a set of healthy schemas.

As an example, considering a failing test case  $t - \{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}$ , and a set of schema schemas  $HSS - \{\{(p_1, 1), (p_2, 1), (p_3, 1)\}, \{(p_3, 1), (p_4, 1)\}\}$ . Table 5 shows the schema set  $\{c \mid c < t \&\& \ \forall c_i \in HSS, c \nleq c_i\}$ , CMNS(HSS, t) and  $\{c \mid c < t \&\& \ \exists c_1' \in CMNS(HSS, t), c_1' \le c\}$ .

Test case t	$\{c \mid c \prec t \&\& \ \forall c_i \in HSS, c \nleq c_i\}$	CMNS(HSS, t)	$\{c \mid c < t \&\& \exists c'_1 \in CMNS(HSS, t), c'_1 \leq c\}$
$\{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}$	$\{(p_1, 1), (p_4, 1)\}$	$\{(p_1, 1), (p_4, 1)\}$	$\{(p_1, 1), (p_4, 1)\}$
Heathy schema set HSS	$\{(p_2, 1), (p_4, 1)\}$	$\{(p_2, 1), (p_4, 1)\}$	$\{(p_2, 1), (p_4, 1)\}$
$\{(p_1, 1), (p_2, 1), (p_3, 1)\}$	$\{(p_1, 1), (p_2, 1), (p_4, 1)\}$		$\{(p_1, 1), p_2, 1), (p_4, 1)\}$
$\{(p_3, 1), (p_4, 1)\}$	$\{(p_1, 1), (p_3, 1), (p_4, 1)\}$		$\{(p_1, 1), (p_3, 1), (p_4, 1)\}$
	$\{(p_2, 1), (p_3, 1), (p_4, 1)\}$		$\{(p_2, 1), (p_3, 1), (p_4, 1)\}$

Table 5. An example of Proposition 4.9

Based on Proposition 4.5 and 4.9, we can easily learn that  $\{c \mid c < t \&\& \forall c_i \in FSS, c_i \nleq c\}$   $\cap \{c \mid c < t \&\& \forall c_i \in HSS, c \nleq c_i\} = \{c \mid \exists c_1' \in CMXS(FSS, t), c \leq c_1'\} \cap \{c \mid c < t \&\& \exists c_1' \in CMNS(HSS, t), c_1' \leq c\}$ . Considering  $\forall c \in \{c \mid \exists c_1' \in CMXS(FSS, t), c \leq c_1'\}, c < t$ , we can transform the aforementioned formula into the following equation.

 $\{c \mid c < t \&\& \forall c_i \in FSS, c_i \nleq c\&\& \forall c_i \in HSS, c \nleq c_i\} = \{c \mid \exists c_1' \in CMXS(FSS, t), c \leq c_1' \&\& \exists c_1' \in CMNS(HSS, t), c_1' \leq c\} = \{c \mid \exists c_1' \in CMXS(FSS, t), c_2' \in CMXS(FSS, t), c_2' \leq c \leq c_1' \}.$ 

Note that the leftmost side of this equation is identical to Formula 1, hence, we can learn that

$$PSS = \{c \mid \exists c_1' \in CMXS(FSS, t), c_2' \in CMNS(HSS, t), c_2' \le c \le c_1'\}. \tag{2}$$

According to Formula 2, the complexity of obtaining one pending schema is  $O(\tau^{|FSS|} \times \tau^{|HSS|})$ . This is because to obtain one pending schema, we only need to search the schemas in CMXS(FSS,t) and CMNS(HSS,t), of which the complexity are  $O(\tau^{|FSS|})$  and  $O(\tau^{|HSS|})$ , respectively. Then we need to check each pair of schemas in these two sets, to find whether exists  $c_1 \in CMXS(FSS,t)$ ,  $c_2 \in CMNS(HSS,t)$ , such that  $c_2 \leq c_1$ . If so, then both  $c_2$  and  $c_1$  satisfy Formula 2. Furthermore,  $\forall c_3, c_2 \leq c_3 \leq c_1, c_3$  also satisfy Formula 2. Hence, the complexity of obtaining one pending schema is  $O(\tau^{|FSS|} \times \tau^{|HSS|})$ .

In fact, we can further reduce the complexity of obtaining pending schemas. When given a set of schemas S, let the minimal schemas as  $S^{\perp} = \{c | c \in S, \nexists c' \in S, c' < c\}$  and the maximal schemas as  $S^{\top} = \{c | c \in S, \nexists c' \in S, c < c'\}$ . Based on this, we can have the following proposition:

PROPOSITION 4.10. Given a failing test case t, a set of faulty schemas FSS, and a set of healthy schemas HSS, we have  $\{c \mid \exists c_1' \in CMXS(FSS^{\perp}, t), c_2' \in CMNS(HSS^{\top}, t), c_2' \leq c \leq c_1' \} = \{c \mid \exists c_1' \in CMXS(FSS, t), c_1' \in CMNS(HSS, t), c_2' \leq c \leq c_1' \}.$ 

As an example, consider a failing test case  $t = \{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}$ , the faulty schema set  $FSS = \{\{(p_1, 1), (p_2, 1), (p_3, 1)\}, \{(p_1, 1), (p_2, 1)\}\}$ , and the healthy schema set  $HSS = \{\{(p_2, 1), (p_3, 1), (p_4, 1)\}, \{(p_2, 1), (p_3, 1)\}\}$ . It is easy to learn that the minimal faulty schema set  $FSS^{\perp} = \{\{(p_1, 1), (p_2, 1)\}\}$ , and the maximal healthy schema set  $HSS^{\top} = \{\{(p_2, 1), (p_3, 1), (p_4, 1)\}\}$ .

Fig 7 lists all the faulty schemas, healthy schemas, and pending schemas of test case t. At the second part, it lists the CMXS(FSS,t), CMNS(HSS,t), and the schema set  $\{c \mid \exists c_1' \in CMXS(FSS,t), c_1' \in CMNS(HSS,t), c_2' \leq c \leq c_1' \}$ . At last it shows  $CMXS(FSS^{\perp},t)$ ,  $CMNS(HSS^{\top},t)$  and  $\{c \mid \exists c_1' \in CMXS(FSS^{\perp},t), c_2' \in CMNS(HSS^{\top},t), c_2' \leq c \leq c_1' \}$ .

We can learn that these two schema set mentioned in Proposition 4.10 are identical in Fig 7, but the schema set based on  $CMXS(FSS^{\perp},t)$ ,  $CMNS(HSS^{\top},t)$  used much fewer schemas (3 in total) to obtain the result than the schema set based on CMXS(FSS,t), CMNS(HSS,t) (7 in total). Another observation from this figure is that these two schema set are both identical to the pending schemas. It suggests that using CMXS and CMNS can effectively and efficiently to get the pending schemas, when compared to list all the schemas in a test case and find the pending schema one by one according to Formula 1.

At last, according to Proposition 4.10, we have the third formula to compute pending schemas as follow:

$$PSS = \{c \mid \exists c_1' \in CMXS(FSS^{\perp}, t), c_2' \in CMNS(HSS^{\top}, t), c_2' \le c \le c_1'\}.$$
(3)

According to Formula 3, the complexity of obtaining one pending schema is  $O(\tau^{|FSS^{\perp}|} \times \tau^{|HSS^{\top}|})$ , where  $|FSS^{\perp}|$  and  $|HSS^{\top}|$  are two relatively small numbers during MFS identification.

# 4.3 Getting back to the motivation example

When applying this analysis, we can get back to the motivation example. We do not need list all the schemas, but we can figure out which one is the pending schema (Covering array, OFOT, OFOT with single MFS, FIC, FIC with multiple )

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Failing Test Case t: \{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}
                            Faulty Schema Set (FSS): \{(p_1, 1), (p_2, 1), (p_3, 1)\}, \{(p_1, 1), (p_2, 1)\}
                            Healthy Schema Set (HSS): \{(p_2, 1), (p_3, 1), (p_4, 1)\}, \{(p_2, 1), (p_3, 1)\}
                                        Minimal Faulty Schema Set (FSS^{\perp}): {(p_1, 1), (p_2, 1)}
                                 Maximal Healthy Schema Set (HSS<sup>T</sup>): \{(p_2, 1), (p_3, 1), (p_4, 1)\}
Faulty Schemas
                                                   Healthy Schemas
                                                                                                       Pending Schemas
                                                   \{(p_2, 1), (p_3, 1), (p_4, 1)\}
                                                   \{(p_2, 1), (p_3, 1)\}
\{(p_1, 1), (p_2, 1), (p_3, 1), (p_4, 1)\}
                                                                                                       \{(p_1, 1), (p_3, 1), (p_4, 1)\}
                                                   \{(p_2, 1), (p_4, 1)\}
                                                                                                       { (p<sub>1</sub>, 1), (p<sub>4</sub>, 1) }
\{(p_1, 1), (p_2, 1), (p_3, 1)\}
                                                   \{(p_3, 1), (p_4, 1)\}
                                                                                                      { (p<sub>1</sub>, 1), (p<sub>3</sub>, 1) }
{ (p<sub>1</sub>, 1) }
\{(p_1, 1), (p_2, 1), (p_4, 1)\}
                                                   \{(p_2, 1)\}
{(p<sub>1</sub>, 1),(p<sub>2</sub>, 1)}
                                                   \{(p_3, 1)\}
                                                   { (p<sub>4</sub>, 1) }
CMXS(FSS, t)
                                                   CMNS(HSS, t)
                                                                                                       \{c \mid \exists c_1 \in CMXS(FSS,t), c_2 \in A_1\}
                                                                                                       CMNS(HSS,t), c_2 \le c \le c_1
\{(p_2, 1), (p_3, 1), (p_4, 1)\}
                                                                                                       \{(p_1, 1), (p_3, 1), (p_4, 1)\}
\{(p_1, 1), (p_3, 1), (p_4, 1)\}
                                                   \{(p_1, 1), (p_4, 1)\}
                                                                                                       \{(p_1, 1), (p_4, 1)\}
\{(p_3, 1), (p_4, 1)\}
                                                                                                      { (p<sub>1</sub>, 1), (p<sub>3</sub>, 1) } { (p<sub>1</sub>, 1) }
                                                   \{(p_1, 1)\}
\{(p_2, 1), (p_4, 1)\}
{ (p_1, 1), (p_4, 1) }
                                                                                                       \{c \mid \exists c_1 \in CMXS(FSS^{\perp},t), c_2 \in
CMXS(FSS^{\perp}, t)
                                                   CMNS(HSS^{T}, t)
                                                                                                       CMNS(HSS<sup>T</sup>,t), c_2 \le c \le c_1 }
                                                                                                       \{(p_1, 1), (p_3, 1), (p_4, 1)\}
\{(p_2, 1), (p_3, 1), (p_4, 1)\}
                                                                                                       \{(p_1, 1), (p_4, 1)\}
                                                   \{(p_1, 1)\}
\{(p_1, 1), (p_3, 1), (p_4, 1)\}
                                                                                                       { (p<sub>1</sub>, 1), (p<sub>3</sub>, 1) }
                                                                                                       \{(p_1, 1)\}
```

Fig. 7. The example of Proposition 4.10

### **5 EMPIRICAL STUDIES**

- 5.1 The pending schemas for covering arrays
- 5.2 The existence of pending schemas for different MFS identification approaches
- 5.3 The characteristics of pending schemas with various types of MFS (multiple, overlapped, single, low-high degrees)
- 5.4 The effectiveness of the approach
- 5.5 Threats to validity
- 6 DISCUSSION

# 7 RELATED WORKS

Shi and Nie [15] presented an approach for failure revealing and failure diagnosis in CT , which first tests the SUT with a covering array, then reduces the value schemas contained in the failing test case by eliminating those appearing in the passing test cases. If the failure-causing schema is found in the reduced schema set, failure diagnosis is completed with the identification of the specific input values which caused the failure; otherwise, a further test suite based on SOFOT is

developed for each failing test case, and the schema set is then further reduced, until no more faults are found or the fault is located. Based on this work, Wang [17] proposed an AIFL approach which extended the SOFOT process by adaptively mutating factors in the original failing test cases in each iteration to characterize failure-inducing interactions.

Nie et al. [12] introduced the notion of Minimal Failure-causing Schema(MFS) and proposed the OFOT approach which is an extension of SOFOT that can isolate the MFS in the SUT. This approach mutates one value of that parameter at a time, hence generating a group of additional test cases each time to be executed. Compared with SOFOT, this approach strengthens the validation of the factor under analysis and can also detect the newly imported faulty interactions.

Delta debugging [20] is an adaptive divide-and-conquer approach to locate interaction failure. It is very efficient and has been applied to real software environment. Zhang et al. [22] also proposed a similar approach that can efficiently identify the failure-inducing interactions that have no overlapped part. Later, Li [8] improved the delta-debugging based approach by exploiting useful information in the executed covering array.

Colbourn and McClary [2] proposed a non-adaptive method. Their approach extends a covering array to the locating array to detect and locate interaction failures. Martínez [9, 10] proposed two adaptive algorithms. The first one requires safe value as the assumption and the second one removes this assumption when the number of values of each parameter is equal to 2. Their algorithms focus on identifying faulty tuples that have no more than 2 parameters.

Ghandehari et al. [5] defined the suspiciousness of tuple and suspiciousness of the environment of a tuple. Based on this, they ranked the possible tuples and generated the test configurations. They [4] further utilized the test cases generated from the inducing interaction to locate the fault.

Yilmaz [18] proposed a machine learning method to identify inducing interactions from a combinatorial testing set. They constructed a classification tree to analyze the covering arrays and detect potential faulty interactions. Beside this, Fouché [3] and Shakya [14] made some improvements in identifying failure-inducing interactions based on Yilmaz's work.

Our previous work [13] proposed an approach that utilizes the tuple relationship tree to isolate the failure-inducing interactions in a failing test case. One novelty of this approach is that it can identify the overlapped faulty interaction. This work also alleviates the problem of introducing new failure-inducing interactions in additional test cases.

# 8 CONCLUSION

To identify the MFS in the SUT is important because it provides additional supports to find the cause of the failure. Although many efforts have been proposed to make the MFS identification approaches more efficient and effective, the existence of pending schemas still make them incomplete, which would be hidden dangers to the software under testing. Hence, to identify these pending schemas is of great importance to reveal such potential risks.

In this paper, we studied the characteristics of the relationships among faulty schemas, healthy schemas, minimal failure-causing schemas, and maximal healthy schemas. The subsuming relationships among them form the basis of the methodology for pending schemas. Particulary, we proposed several propositions to formulate the set of pending schemas and give three equivalent formulas to obtain them. Among the three formulas, the last formula reduce the complexity of obtaining pending schemas from  $O(2^n)$  to  $O(\tau^{|FSS^\perp|+|HSS^\top|})$ , where n is the number of factors in the software, while  $|FSS^\perp|$  and  $|HSS^\top|$  are two relatively small numbers and independent on the number of n.

We conduct a series empirical studies on some real software systems with various number of parameters and values. Our results shows that the incompleteness in very common in the covering

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arrays and MFS identification approaches. We also observed that the third proposed formula is the most efficient when compared others in most cases.

As a future work, it is appealing to take the advantages of the pending schemas to assist in the MFS identification approaches. More specifically, we would like to develop a new framework for MFS identification, which take the pending schemas as a indicator. That is, any MFS identification algorithm that runs on this framework aims to empty the set of pending schemas. It is of interest to evaluate the efficiency and effectiveness of this framework with respect to various MFS identification approaches that run on it.

#### A PROOF

We will give the proofs of several important propositions.

Proposition 4.3.

```
PROOF. Let A = \{c \mid c < t \&\& c_1 \not \leq c\}, and B = \{c \mid \exists c_1' \in CMXS(c_1, t), c \leq c_1'\}.
 Let c_1 = \{(p_{x_1}, v_{x_1}), (p_{x_2}, v_{x_2}), ..., (p_{x_k}, v_{x_k})\}, t = \{(p_1, v_1), (p_2, v_2), ..., (p_n, v_n)\}, and CMXS(c_1, t) = \{t \setminus (p_{x_i}, v_{x_i}) \mid (p_{x_i}, v_{x_i}) \in c_1\}.
```

First we will show  $A \subseteq B$ .

With respect to set A,  $\forall c' \in A$ , it has c' < t and  $c_1 \nleq c'$ . That is,  $\forall e \in c', e \in t$ , and  $\exists e' \in c_1, e' \notin c'$ . As  $c_1 \leq t, e' \in t$ . Hence, we have  $\forall e \in c', e \in t \setminus e'$ , i.e.,  $c' \leq t \setminus e'$ .

Since  $t \setminus e' \in CMXS(c_1, t), c' \in \{c \mid \exists c_1' \in CMXS(c_1, t), c \le c_1'\} = B$ . Hence,  $A \subseteq B$ .

Second we will show  $B \subseteq A$ .

With respect to set B,  $\forall c' \in B$ , it has  $\exists c'_1 \in CMXS(c_1, t), c' \leq c'_1$ . Since  $c'_1 \in CMXS(c_1, t)$ ,  $\exists e' \in c_1, c'_1 = t \setminus e'$ . Consequently,  $c' \leq t \setminus e'$ . Hence,  $c_1 \nleq c'$ . Also,  $c' \leq t \setminus e' < t$ . Consequently,  $c \in \{c \mid c < t \&\& c_1 \nleq c\} = A$ , which indicates that  $B \subseteq A$ .

As we have shown  $B \subseteq A$ , and  $A \subseteq B$ , so A = B.

Proposition 4.5.

PROOF. We just need to prove that for two faulty schemas  $c_1$ ,  $c_2$ , and a failing test case t ( $c_1 \le t$ ,  $c_2 \le t$ ), we have  $\{c \mid c < t \&\& \forall c_i \in \{c_1, c_2\}, c_i \not \le c\} = \{c \mid \exists c_1' \in CMXS(c_1, t) \land CMXS(c_2, t), c \le c_1'\}$ . Let  $A = \{c \mid c < t \&\& \forall c_i \in \{c_1, c_2\}, c_i \not \le c\}$ ,  $A_1 = \{c \mid c < t \&\& c_1 \not \le c\}$ ,  $A_2 = \{c \mid c < t \&\& c_2 \not \le c\}$ . It is easily to get  $A = A_1 \cap A_2$ .

Let  $B = \{c \mid \exists c_1' \in CMXS(c_1, t) \land CMXS(c_2, t), c \leq c_1'\}$ . Here,  $CMXS(c_1, t) \land CMXS(c_2, t) = \{c \mid c = c_1' \cap c_2', \text{ where } c_1' \in CMXS(c_1, t), \text{ and } c_2' \in CMXS(c_2, t)\}$ .

Let  $B_1 = \{c \mid \exists c_1' \in CMXS(c_1, t), c \leq c_1'\}$ , and  $B_2 = \{c \mid \exists c_2' \in CMXS(c_2, t), c \leq c_2'\}$ .  $B_1 \cap B_2 = \{c \mid \exists c_1' \in CMXS(c_1, t), c \leq c_1' \&\& \exists c_2' \in CMXS(c_2, t), c \leq c_2'\}$ . Note that,  $c \leq c_1' \&\& c \leq c_2' \equiv c \leq c_1' \cap c_2'$ . Hence,  $B_1 \cap B_2 = \{c \mid \exists c_1', c_2', c_1' \in CMXS(c_1, t), \text{ and } c_2' \in CMXS(c_2, t), c \leq c_1' \cap c_2'\} = B$ . Based on Proposition 4.3,  $A_1 = B_1$ ,  $A_2 = B_2$ . Consequently,  $A = A_1 \cap A_2 = B_1 \cap B_2 = B$ .

Proposition 4.7.

PROOF. Let  $A = \{c \mid c < t \&\& c \nleq c_1\}$ .  $B = \{c \mid c < t \&\& \exists c_1' \in CMNS(c_1, t), c_1' \leq c\}$ .  $CMNS(c_1, t) = \{(p_{x_i}, v_{x_i}) \mid (p_{x_i}, v_{x_i}) \in t \setminus c_1\}$ .

First we will show  $A \subseteq B$ .

With respect to set A,  $\forall c' \in A$ , it has c' < t and  $c' \nleq c_1$ . That is,  $\forall e \in c', e \in t$ , and  $\exists e' \in c', e' \notin c_1$ . Hence,  $\{e'\} \leq c', e' \in t \setminus c_1$ , which indicates that  $c' \in \{c \mid c < t \&\& \exists c'_1 \in CMNS(c_1, t), c'_1 \leq c\} = B$ , so  $A \subseteq B$ .

Second we will show  $B \subseteq A$ .

With respect to set B,  $\forall c' \in B$ , it has c' < t and  $\exists c'_1 \in CMNS(c_1, t), c'_1 \le c'$ . As  $c'_1 \in CMNS(c_1, t)$ ,  $\exists e' \in t \setminus c_1, c'_1 = \{e'\}$ . Hence,  $\{e'\} \le c'$ . Hence,  $\{c'\} \le c'$ . Consequently,  $\{c'\} \in \{c \mid c < t \& \& c \nleq c_1\} = A$ , which indicates that  $\{b\} \subseteq A$ .

Proposition 4.9.

PROOF. We just need to prove that for two healthy schemas  $c_1$ ,  $c_2$ , and a failing test case t ( $c_1 < t$ ,  $c_2 < t$ ), we have  $\{c \mid c < t \&\& \forall c_i \in \{c_1, c_2\}, c \nleq c_i\} = \{c \mid c < t \&\& \exists c_1' \in CHFS(c_1, t) \lor CHFS(c_2, t), c_1' \leq c\}.$ 

Let  $A = \{c \mid c < t \&\& \forall c_i \in \{c_1, c_2\}, c \nleq c_i\}, A_1 = \{c \mid c < t \&\& c \nleq c_1\}, A_2 = \{c \mid c < t \&\& c \nleq c_2\}.$  It is easily to get  $A = A_1 \cap A_2$ .

Let  $B = \{c \mid c < t \&\& \exists c_1' \in CHFS(c_1, t) \lor CHFS(c_2, t), c \le c_1'\}$ . Here,  $CMXS(c_1, t) \lor CMXS(c_2, t) = \{c \mid c = c_1' \cup c_2', where c_1' \in CMXS(c_1, t), and c_2' \in CMXS(c_2, t)\}$ .

Let  $B_1 = \{c \mid c < t \&\& \exists c_1' \in CHFS(c_1,t), c_1' \leq c\}$ , and  $B_2 = \{c \mid c < t \&\& \exists c_2' \in CHFS(c_2,t), c_2' \leq c\}$ .  $B_1 \cap B_2 = \{c \mid c < t \&\& \exists c_1' \in CHFS(c_1,t), c_1' \leq c \&\& \exists c_2' \in CHFS(c_2,t), c_2' \leq c\}$ . Note that,  $c_1' \leq c \&\& c_2' \leq c \equiv c_1' \cup c_2' \leq c$ . Hence,  $B_1 \cap B_2 = \{c \mid c \leq t \&\& \exists c_1', c_2', c_1' \in CHFS(c_1,t), \text{ and } c_2' \in CHFS(c_2,t), c_1' \cup c_2' \leq c\} = B$ .

Based on Proposition 4.7,  $A_1 = B_1$ ,  $A_2 = B_2$ . Consequently,  $A = A_1 \cap A_2 = B_1 \cap B_2 = B$ .

Proposition 4.10.

PROOF. Let A = { $c \mid \exists c_1' \in CMXS(FSS^{\perp}, t), c_2' \in CMNS(HSS^{\top}, t), c_2' \leq c \leq c_1'$ }. Let B = { $c \mid \exists c_1' \in CMXS(FSS, t), c_2' \in CMNS(HSS, t), c_2' \leq c \leq c_1'$ }.

Based on Proposition 4.5 and 4.9:

 $A = \{c \mid c < t \&\& \forall c_i \in FSS^{\perp}, c_i \nleq c \&\& \forall c_i \in HSS^{\top}, c \nleq c_i\}.$ 

 $B = \{c \mid c \prec t \&\& \forall c_i \in FSS, c_i \nleq c \&\& \forall c_i \in HSS, c \nleq c_i\}.$ 

First we will prove  $B \subseteq A$ .

As  $FSS^{\perp} \subseteq FSS$ ,  $HSS^{\top} \subseteq HSS$ ,  $\{c \mid \forall c_i \in FSS, c_i \nleq c\} \subseteq \{c \mid \forall c_i \in FSS^{\perp}, c_i \nleq c\}$  and  $\{c \mid \forall c_i \in HSS, c \nleq c_i\} \subseteq \{c \mid \forall c_i \in HSS^{\top}, c \nleq c_i\}$ . Hence,  $\{c \mid c \prec t \&\& \forall c_i \in FSS, c_i \nleq c \&\& \forall c_i \in HSS, c \nleq c_i\}$   $\subseteq \{c \mid c \prec t \&\& \forall c_i \in FSS^{\perp}, c_i \nleq c \&\& \forall c_i \in HSS^{\top}, c \nleq c_i\}$ . That is  $B \subseteq A$ .

Next we will prove  $A \subseteq B$ .

Note that  $\forall c < t$ , if  $\forall c_i \in FSS^{\perp}$ ,  $c_i \nleq c$ , it must have  $\forall c_i' \in FSS$ ,  $c_i' \nleq c$ . Because if not so, then  $\exists c_i' \in FSS$ ,  $c_i' \leq c$ . As  $\exists c_i \in FSS^{\perp}$ ,  $c_i' \in FSS$ ,  $c_i \leq c_i'$ , hence,  $c_i \leq c_i' \leq c$ , which is contradiction.

Similarly,  $\forall c < t$ , if  $\forall c_i \in HSS^\top$ ,  $c \nleq c_i$ , it must have  $\forall c_i' \in HSS$ ,  $c \nleq c_i'$ . Because if not so, then  $\exists c_i' \in HSS$ ,  $c \leq c_i'$ . As  $\exists c_i \in HSS^\top$ ,  $c_i' \in HSS$ ,  $c_i' \leq c_i$ , hence,  $c \leq c_i' \leq c_i$ , which is contradiction.

Combining them, we can get  $\forall c' \in \{c \mid c < t \&\& \forall c_i \in FSS^{\perp}, c_i \nleq c \&\& \forall c_i \in HSS^{\top}, c \nleq c_i\}, c' \in \{c \mid c < t \&\& \forall c_i \in FSS, c_i \nleq c \&\& \forall c_i \in HSS, c \nleq c_i\}.$  That is,  $A \subseteq B$ .

As we have shown  $B \subseteq A$ , and  $A \subseteq B$ , so A = B.

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