

# Error Locating Driven Array\*

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## ABSTRACT

Combinatorial testing(CT) seeks to handle potential faults caused by various interactions of factors that can influence the Software systems. When applying CT, it is a common practice to first generate a bunch of test cases to cover each possible interaction and then to locate the failure-inducing interaction if any failure is detected. Although this conventional procedure is simple and straightforward, we conjecture that it is not the ideal choice in practice. This is because 1) testers desires to isolate the root cause of failures before all the needed test cases are generated and executed 2) the early located failure-inducing interactions can guide the remaining test cases generation, such that many unnecessary and invalidate test cases can be avoided. For this, we propose a novel CT framework that allows for both generation and localization process to better share each other's information, as a result, both this two testing stages will be more effectively and efficiently when handling the testing tasks. We conducted a series of empirical studies on several open-source software, of which the result shows that our framework can locate the failure-inducing interactions more quickly than traditional approaches, while just needing less test cases.

## Categories and Subject Descriptors

D.2.5 [Software Engineering]: Testing and debugging—*Debugging aids, testing tools*

## General Terms

Reliability, Verification

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## Keywords

Software Testing, Combinatorial Testing, Covering Array, Failure-inducing combinations

## 1. INTRODUCTION

Modern software is designed modular, platform-cross, language-cross and configurable. Although the sophisticated design can make the system more flexible and more powerful, it also makes the testing task a challenging work. This is because the candidate factors that can influence the system's behaviour, e.g., possible configuration options, system inputs, message events and the like, increases rapidly, and what's worse, the interactions between these factors can also crash the system, e.g., the compatibility problems. In consideration of the scale of the possible configurations, inputs, events, and their possible interactions (we call them the interactions space) of the real industrial software, to test all the possible combination of all the possible factors is not feasible, and even it is possible, it is also wasting as numerous interactions do not provide any useful information.

Many empirical studies shows that in real software systems, the effective interaction space, i.e., targeting fault defects, makes up only a small proportion of the overall interaction space. What's more, the number of factors involved in these effective interactions is relatively small, of which 4-6 is usually the upper bonds. With this observation, applying CT in practice is appealing, as it is proven to be effective to handle the interaction faults between the system.

A typical CT life-circle is listed as Figure 1, in which there are four main testing stages. At the very beginning of the testing, engineers should extract the specific model of the software under test. In detail, they should identify the factors, such as user inputs, configure options, environment states and the like that could affect the system's behavior. Next, which values each factor can take should also be determined to maximize the chances to expose the potential faults in the system. More efforts will be needed to describe the constraints and dependencies between each factor and corresponding values to make the testing work valid. After the modeling stage, a bunch of test cases should be generated and executed to test the system and try to detect some . In CT, one test case is a set of assignment of all the factors that we modeled before. Thus, when such a test case is executed, the is deemed to be validated. The main target of this stage is to using a relatively small size of test cases to get some coverage, for CT the most common coverage is to cover all the possible interactions with the number of factors no more than a fixed number :  $t$ . The third testing stage

in this circle is the fault localization, which is responsible for diagnosing the root cause of the failure we detect in the previous stage. To characterize such root cause, i.e., failure-inducing interactions of corresponding factors and values is important for future bug fixing, as it will reduce the suspicious code scope that needed to inspect. The last testing stage of CT is evaluation. It aims at assessing the quality of this testing,

Although this conventional CT framework is simple and straightforward, however in terms of the test cases generation and fault localization testing stages, we conjecture that the first-generation-then-localization is not the proper choice for most test engineers. This is because, first, it is not realistic for testers wait for all the needed test cases are generated before they can diagnosis and fix some failures during testing, second, and the most important, utilizing the early determined failure-inducing interactions can guide the following test cases generations, such that many unnecessary and invalid test cases will be avoided. For this we get to the most key idea of this paper:

*Generation and Localization process should be integrated more tightly.*

Based on the idea, we propose a new CT framework, which modifies. To this aim, we remodel the generation and localization modular to make them better adapt to this newly framework. In specific, our generation adopts the one-test-one-time strategy, i.e., generate and execute one test case in one iteration. Rather than generating the overall needed test cases, this strategy is more flexible so that it allows for terminating at any time during generation, no matter whether the coverage is met or not. With this strategy, we can let the generation stops at the point we detect some failures, and then after localization we can utilize the information got by localization to change some coverage criteria, e.g., the interactions related to the failure-inducing interactions do not need to cover any more. Then the generation goes on the process repeats. For the fault localization, we will also take use of the test cases generation before, this will shows in the extra test case. Fault localization will need more.

We conducted a series empirical studies on several open-source software to evaluate our newly framework. In short, we take two main comparison, one is for , and another is for. The results shows that our approach can and significantly reduced the needed test cases to detect and locate the failure-inducing interactions in the system under test, and our approach can than the traditional ones.

The main contributions of this paper:

1. we propose a newly CT framework.
2. we propose two modification to the generation and localization to make them adapt to the newly framework.
3. A series empirical studies is conducted.

The reset of the paper is organised as follows: Section 2 presents the preliminary background of some definitions of the CT. Section 3 gives Section 4

## 2. MOTIVATING EXAMPLE

Combinatorial testing can effectively detect the failures caused by the interactions between various options or inputs of the SUT. Covering arrays, the test suite generated by this technique can cover each combination of the options at

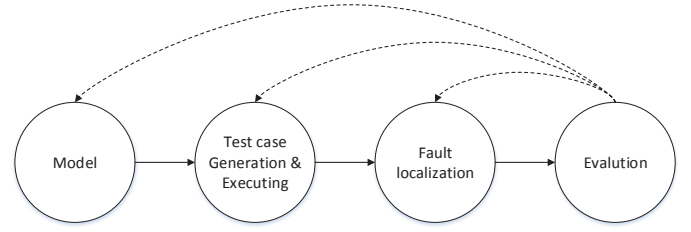


Figure 1: The life circle of CT

least once. We conjecture, however, although covering array can effectively, in practice, covering array was too much for detecting and locating the error in particular software.

As an motivating example, we looked through the following scenarios for detecting and locating the errors in the SUT.

Too much redundant fault test cases:

Too much redundant right test cases:

## 3. BACKGROUND

This section presents some definitions and propositions to give a formal model for the FCI problem.

### 3.1 Failure-inducing combinations in CT

Assume that the SUT is influenced by  $n$  parameters, and each parameter  $p_i$  has  $a_i$  discrete values from the finite set  $V_i$ , i.e.,  $a_i = |V_i|$  ( $i = 1, 2, \dots, n$ ). Some of the definitions below are originally defined in .

*Definition 1.* A test case of the SUT is an array of  $n$  values, one for each parameter of the SUT, which is denoted as a  $n$ -tuple  $(v_1, v_2, \dots, v_n)$ , where  $v_1 \in V_1, v_2 \in V_2 \dots v_n \in V_n$ .

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 software.

We consider the fact that the abnormally executing test cases as a *fault*. It can be a thrown exception, compilation error, assertion failure or constraint violation. When faults are triggered by some test cases, what is desired is to figure out the cause of these faults, and hence some subsets of this test case should be analysed.

*Definition 2.* For the SUT, the  $n$ -tuple  $(-, v_{n_1}, \dots, v_{n_k}, \dots)$  is called a  $k$ -value combination ( $0 < k \leq n$ ) when some  $k$  parameters have fixed values and the others can take on their respective allowable values, represented as “-”.

In effect a test case itself is a  $k$ -value combination, when  $k = n$ . Furthermore, if a test case contain a combination, i.e., every fixed value in the combination is in this test case, we say this test case *hits* the combination.

*Definition 3.* let  $c_l$  be a  $l$ -value combination,  $c_m$  be an  $m$ -value combination in SUT and  $l < m$ . If all the fixed parameter values in  $c_l$  are also in  $c_m$ , then  $c_m$  subsumes  $c_l$ . In this case we can also say that  $c_l$  is a sub-combination of  $c_m$  and  $c_m$  is a parent-combination of  $c_l$ , which can be denoted as  $c_l \prec c_m$ .

For example, in the motivation example section, the 2-value combination  $(-, 4, 4, -)$  is a sub-combination of the 3-value combination  $(-, 4, 4, 5)$ , that is,  $(-, 4, 4, -) \prec (-, 4, 4, 5)$ .

*Definition 4.* If all test cases contain a combination, say  $c$ , trigger a particular fault, say  $F$ , then we call this combination  $c$  the *faulty combination* for  $F$ . Additionally, if none sub-combination of  $c$  is the *faulty combination* for  $F$ , we then call the combination  $c$  the *minimal faulty combination* for  $F$  (It is also called Minimal failure-causing schema(MFS) in ).

In fact, MFS and *minimal faulty combinations* are identical to the failure-inducing combinations we discussed previously. Figuring it out can eliminate all details that are irrelevant for causing the failure and hence facilitate the debugging efforts.

## **4. ALGORITHMS**

### **4.1 Description**

### **4.2 A case study**

## **5. EMPIRICAL STUDIES**

### **5.1 The existence of**

#### *5.1.1 Study setup*

#### *5.1.2 Result and discussion*

### **5.2 Performance of the traditional algorithms**

#### *5.2.1 Study setup*

#### *5.2.2 Result and discussion*

### **5.3 Performance of our approach**

#### *5.3.1 Study setup*

#### *5.3.2 Result and discussion*

### **5.4 Threats to validity**

## **6. RELATED WORKS**

## **7. CONCLUSIONS**