Constrained Detecting Arrays: Mathematical Structures for Fault Identification in Combinatorial Interaction Testing

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\section*{ABSTRACT}

\textbf{Context:} Detecting arrays are mathematical structures aimed at fault identification in combinatorial interaction testing. However, they cannot be directly applied to systems that have constraints among test parameters. Such constraints are prevalent in real-world systems.

\textbf{Objectives:} This paper proposes Constrained Detecting Arrays (CDAs), an extension of detecting arrays, which can be used for systems with constraints.

\textbf{Methods:} The paper examines the properties and capabilities of CDAs with rigorous arguments. The paper also proposes two algorithms for constructing CDAs: One is aimed at generating minimum CDAs and the other is a heuristic algorithm aimed at fast generation of CDAs. The algorithms are evaluated through experiments using a benchmark dataset.

\textbf{Results:} Experimental results show that the first algorithm can generate minimum CDAs if a sufficiently long generation time is allowed, and the second algorithm can generate minimum or near-minimum CDAs in a reasonable time.

\textbf{Conclusion:} CDAs enhance detecting arrays to be applied to systems with constraints. The two proposed algorithms have different advantages with respect to the array size and generation time.

\section{1. Introduction}

Combinatorial Interaction Testing (CIT) is a testing approach that aims to exercise interactions among test parameters. The basic strategy of CIT is to test all interactions among a specified number (usually a small integer such as 2 or 3) of parameters. Empirical results suggest that it is sufficient to only test those interactions involving a small number of parameters to reveal most of the latent faults \cite{1, 2}. Using CIT can cut off testing cost significantly when compared to exhaustive testing. The test suites used in CIT are usually modeled as arrays where each row represents a test case and each column corresponds to each test parameter. The most typical class of arrays used for CIT is \textit{t-way Covering Arrays} (\textit{t}-CAs). In a \textit{t}-CA every interaction involving \textit{t} parameters appears in at least one test case; thus the use of a \textit{t}-CA ensures exercising all \textit{t}-way interactions.

There are many directions to expand the capability of CIT. One of the directions is to add fault localization capability to test suites. \textit{(d, t)}-\textit{Locating Arrays} (LAs) and \textit{(d, t)}-\textit{Detecting Arrays} (DAs) proposed in \cite{3} represent test suites that can not only detect but also identify faulty interactions. The integers \textit{d} and \textit{t} are predefined parameters: \textit{d} represents the number of faulty interactions that can be identified and \textit{t} represents the number of parameters involved in the faulty interactions. LAs and DAs add this capability to CAs at the cost of an increased number of test cases.

Another direction of expanding CIT is to incorporate constraints. Real-world systems usually have constraints on the input space. These constraints are originated from, for example, user-defined requirements or running environment restrictions. In order to test systems with constraints correctly, proper handling of the constraints is necessary. For example, all test cases must satisfy the constraints. In addition, constraints may make some interactions no longer testable. These \textit{invalid} interactions require additional handling. \textit{Constrained Covering Arrays} (CCAs) are an extension of CAs in which such constraints are incorporated into the definition. Many previous studies on CIT have tackled the problem of generating CCAs of small sizes \cite{4, 5, 6}.

In \cite{7} we proposed the concept of \textit{Constrained Locating Arrays} (CLAs) which extends LAs by incorporating constraints. CLAs inherit basic properties of LAs and at the same time can be used as test suites for systems that have constraints on the input space. In this paper, we further develop this line of research: We propose a new mathematical structure called \textit{Constrained Detecting Arrays} (CDAs). As the name suggests, CDAs extend DAs by incorporating
The valid/invalid distinction also applies to interactions: Interactions that no valid test cases can cover are invalid; the other interactions, i.e., those that are covered by at least one valid test case are valid. We let \( I_v \) and \( V I \) denote the set of all t-way interactions and the set of all valid t-way interactions, respectively. Similarly we let \( \bar{I}_v \) and \( \bar{V}I \) be the set of all interactions of strength at most \( t \) and the set of all valid interactions of strength at most \( t \).

A valid interaction is either faulty or non-faulty. The outcome of execution of a valid test case is either PASS or FAIL. The outcome is FAIL iff the test case covers at least one faulty interaction; the outcome is PASS otherwise. The
test outcome of an array is the collection of the outcomes of all rows.

Table 1 shows an SUT model which represents an online shopping mobile application. This example, which is a modification of one in [8], serves as a running example throughout the paper. This model is formally represented as $M = \{T, S, \phi\}$ where $T = \{T_1, T_2, F_1, F_2\}$, $S = \{S_1, S_2, S_3, S_4\}$, $S_1 = S_3 = \{0, 1, 2\}$, $S_2 = \{0, 1\}$, $S_4 = \{0, 1, 2, 3\}$ and $\phi = \phi_1 \land \phi_2 = (F_1 = 1 \Rightarrow F_3 \neq 0) \land (F_4 = 3 \Rightarrow (F_2 = 0 \land F_3 = 0))$. An example of a valid test case is $(0, 0, 0, 0)$, representing $(50, Domestic, Same-Day Delivery, Visa)$. On the other hand, $(0, 1, 0, 0)$ (i.e., $(50, International, Same-Day Delivery, Visa))$ is an invalid test case. Invalid interactions include, for example, \{$(F_2, 1), (F_3, 0)$\}, \{$(F_2, 1), (F_4, 3)$\}, etc.

2.2. Covering arrays, locating arrays, and detecting arrays

Covering arrays (CAs), locating arrays (LAs), and detecting arrays (DAs) are mathematical structures that can be implemented as test suites. They are usually used to detect or locate faulty interactions for unconstrained SUTs (i.e., $\phi = true$).

A $t$-way covering array, $t$-CA for short, is defined as follows:

$t$-CA \quad $\forall T \in I_t : \rho_A(T) \neq \emptyset$

The condition requires that all interactions $T$ in $I_t$ be covered by at least one row in the array. In other words, when the test cases in $A$ are executed, all $t$-way interactions are examined or executed at least once. This condition is sufficient to reveal the existence of a $t$-way faulty interaction; but it is generally not possible to identify the faulty interaction from the test outcome. Figure 1a shows a 2-CA for the running example.

On the other hand, LAs and DAs can be used to not only detect the existence of faulty interactions but also locate them. LAs and DAs were first proposed by Colbourn and McClary in [3]. They introduced a total of six types for both LAs and DAs according to fault locating capability. Two types of them exist only in extreme cases. The rest four types, namely, $(d, t)$-, $(\overline{d}, t)$-, $(d, \overline{t})$-, $(\overline{d}, \overline{t})$-LA (and DA), are as follows $(d \geq 0, 0 \leq t \leq k)$.

$(d, t)$-LA \quad $\forall T_1, T_2 \subseteq I_t$ such that $|T_1| = |T_2| = d :$

$\rho_A(T_1) = \rho_A(T_2) \iff T_1 = T_2$

$(\overline{d}, t)$-LA \quad $\forall T_1, T_2 \subseteq I_t$ such that $0 \leq |T_1| \leq d, 0 \leq |T_2| \leq d :$

$\rho_A(T_1) = \rho_A(T_2) \iff T_1 = T_2$

$(d, \overline{t})$-LA \quad $\forall T_1, T_2 \subseteq \overline{I_t}$ such that $|T_1| = |T_2| = d$ and $T_1, T_2$ are independent :

$\rho_A(T_1) = \rho_A(T_2) \iff T_1 = T_2$

$(\overline{d}, \overline{t})$-LA \quad $\forall T_1, T_2 \subseteq \overline{I_t}$ such that $0 \leq |T_1| \leq d, 0 \leq |T_2| \leq d$ and $T_1, T_2$ are independent :

$\rho_A(T_1) = \rho_A(T_2) \iff T_1 = T_2$

$(d, t)$-DA \quad $\forall T \subseteq I_t$ such that $|T| = d, \forall T \in I_t :$

$\rho_A(T) \subseteq \rho_A(T) \iff T \in T$

$(\overline{d}, t)$-DA \quad $\forall T \subseteq I_t$ such that $0 \leq |T| \leq d, \forall T \in I_t :$

$\rho_A(T) \subseteq \rho_A(T) \iff T \in \overline{T}$

$(d, \overline{t})$-DA \quad $\forall T \subseteq \overline{I_t}$ such that $|T| = d, \forall T \in \overline{I_t}$ and $T \cup \{T\}$ is independent :

$\rho_A(T) \subseteq \rho_A(T) \iff T \in \overline{T}$

$(\overline{d}, \overline{t})$-DA \quad $\forall T \subseteq \overline{I_t}$ such that $0 \leq |T| \leq d, \forall T \in \overline{I_t}$ and $T \cup \{T\}$ is independent :

$\rho_A(T) \subseteq \rho_A(T) \iff T \in \overline{T}$

The parameter $d$ of these arrays stands for the number of faulty interactions that the array can correctly locate, while $t$ represents the strength of the target interactions. Writing $\overline{d}$ or $\overline{t}$ in place of $d$ or $t$ means that the array permits
at most of \( d \) faulty interactions or strength at most \( t \). For example, a \((1,2)\)-LA (or DA) can locate one 2-way faulty interaction, while a \((2,2)\)-LA (or DA) can locate at most two faulty interactions that have strength not greater than 2.

The reason why when dealing with \( T \) it is required that \( T_1, T_2 \) or \( T \cup \{T\} \) be independent is that if there are \( T_1 \subseteq T_2 \), whether \( T_2 \) is faulty or not cannot be determined when \( T_1 \) is faulty. Figure 1b and Figure 1c show a \((1,2)\)-LA and a \((1,2)\)-DA for the running SUT.

In [3] it is proved that a \((d, t)\)-DA is a \((d, t)\)-LA and a \((\bar{d}, \bar{t})\)-LA and that a \((\bar{d}, \bar{t})\)-LA is a \((d - 1, t)\)-DA (Lemma 7.1). It is also proved that a \((d, t)\)-DA is equivalent to a \((\bar{d}, \bar{t})\)-DA and that a \((d, \bar{t})\)-DA is equivalent to a \((\bar{d}, t)\)-DA (Lemma 7.2). We will later provide theorems for CDAs, namely Theorems 2 and 5, that are parallel to these lemmas.

How to identify faulty interactions using these arrays is the same as their constrained versions, namely, CLAs and CDAs.

### 2.3. Constrained versions of covering arrays and locating arrays

CAs, LAs, and DAs do not take constraints into account. However real-world systems usually have complicated constraints that must be satisfied by all test cases.

#### 2.3.1. Constrained covering arrays

Constrained Covering Arrays (CCAs) are the constrained version of CAs. CCAs are the most common form of test suites used in CIT: Most of test generation tools for CIT are in effect generators of CCAs.

**Definition 1 (CCA).** An array \( A \) that consists of valid test cases is a \( t \)-CCA iff the following condition holds.

\[
(\text{t-CCA}) \quad \forall T \in \mathcal{V}I_t : \rho_A(T) \neq \emptyset
\]

The definition of CCAs requires that all valid \( t \)-way interactions be covered by at least one test case in the test suite. This condition implies that every valid interaction of strength \( < t \) is covered by at least one test case. That is, a \( t \)-CCA

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**Figure 1:** CA, LA, and DA for the running example (constraint ignored)

| (a) 2-CA | (b) (1,2)-LA | (c) (1,2)-DA |
|----------|--------------|--------------|
| \( F_1 \) | \( F_2 \) | \( F_3 \) | \( F_4 \) | \( F_1 \) | \( F_2 \) | \( F_3 \) | \( F_4 \) | \( F_1 \) | \( F_2 \) | \( F_3 \) | \( F_4 \) |
| \( \sigma_1 \) | 0 | 0 | 0 | 0 | \( \sigma_1 \) | 0 | 0 | 0 | 0 | \( \sigma_1 \) | 0 | 0 | 0 | 0 |
| \( \sigma_2 \) | 0 | 0 | 2 | 2 | \( \sigma_2 \) | 0 | 0 | 1 | 1 | \( \sigma_2 \) | 0 | 0 | 0 | 2 |
| \( \sigma_3 \) | 0 | 1 | 0 | 3 | \( \sigma_3 \) | 0 | 0 | 1 | 2 | \( \sigma_3 \) | 0 | 0 | 1 | 1 |
| \( \sigma_4 \) | 0 | 1 | 1 | 1 | \( \sigma_4 \) | 0 | 1 | 0 | 3 | \( \sigma_4 \) | 0 | 0 | 2 | 3 |
| \( \sigma_5 \) | 1 | 0 | 0 | 1 | \( \sigma_5 \) | 0 | 1 | 2 | 0 | \( \sigma_5 \) | 0 | 1 | 0 | 1 |
| \( \sigma_6 \) | 1 | 0 | 2 | 3 | \( \sigma_6 \) | 0 | 1 | 2 | 2 | \( \sigma_6 \) | 0 | 1 | 1 | 3 |
| \( \sigma_7 \) | 1 | 1 | 0 | 2 | \( \sigma_7 \) | 1 | 0 | 0 | 1 | \( \sigma_7 \) | 0 | 1 | 2 | 0 |
| \( \sigma_8 \) | 1 | 1 | 1 | 0 | \( \sigma_8 \) | 1 | 0 | 1 | 0 | \( \sigma_8 \) | 0 | 1 | 2 | 2 |
| \( \sigma_9 \) | 2 | 0 | 0 | 3 | \( \sigma_9 \) | 1 | 0 | 2 | 1 | \( \sigma_9 \) | 1 | 0 | 0 | 1 |
| \( \sigma_{10} \) | 2 | 0 | 1 | 2 | \( \sigma_{10} \) | 1 | 1 | 0 | 0 | \( \sigma_{10} \) | 1 | 0 | 1 | 3 |
| \( \sigma_{11} \) | 2 | 0 | 2 | 0 | \( \sigma_{11} \) | 1 | 1 | 0 | 2 | \( \sigma_{11} \) | 1 | 0 | 2 | 0 |
| \( \sigma_{12} \) | 2 | 1 | 1 | 3 | \( \sigma_{12} \) | 1 | 1 | 1 | 1 | \( \sigma_{12} \) | 1 | 0 | 2 | 2 |
| \( \sigma_{13} \) | 2 | 1 | 2 | 1 | \( \sigma_{13} \) | 1 | 1 | 2 | 3 | \( \sigma_{13} \) | 1 | 1 | 0 | 3 |
| \( \sigma_{14} \) | 2 | 0 | 0 | 2 | \( \sigma_{14} \) | 1 | 1 | 1 | 0 | \( \sigma_{14} \) | 1 | 1 | 1 | 0 |
| \( \sigma_{15} \) | 2 | 0 | 0 | 3 | \( \sigma_{15} \) | 1 | 1 | 1 | 2 | \( \sigma_{15} \) | 1 | 1 | 2 | 1 |
| \( \sigma_{16} \) | 2 | 0 | 1 | 1 | \( \sigma_{16} \) | 1 | 1 | 2 | 1 | \( \sigma_{16} \) | 1 | 1 | 2 | 1 |
| \( \sigma_{17} \) | 2 | 0 | 2 | 3 | \( \sigma_{17} \) | 2 | 0 | 0 | 3 | \( \sigma_{17} \) | 2 | 0 | 0 | 3 |
| \( \sigma_{18} \) | 2 | 1 | 0 | 0 | \( \sigma_{18} \) | 2 | 0 | 1 | 0 | \( \sigma_{18} \) | 2 | 0 | 1 | 0 |
| \( \sigma_{19} \) | 2 | 1 | 1 | 3 | \( \sigma_{19} \) | 2 | 0 | 1 | 2 | \( \sigma_{19} \) | 2 | 0 | 1 | 2 |
| \( \sigma_{20} \) | 2 | 0 | 0 | 2 | \( \sigma_{20} \) | 2 | 0 | 2 | 1 | \( \sigma_{20} \) | 2 | 0 | 2 | 1 |
| \( \sigma_{21} \) | 2 | 1 | 0 | 0 | \( \sigma_{21} \) | 2 | 1 | 0 | 0 | \( \sigma_{21} \) | 2 | 1 | 0 | 0 |
| \( \sigma_{22} \) | 2 | 1 | 0 | 2 | \( \sigma_{22} \) | 2 | 1 | 0 | 2 | \( \sigma_{22} \) | 2 | 1 | 0 | 2 |
| \( \sigma_{23} \) | 2 | 1 | 1 | 1 | \( \sigma_{23} \) | 2 | 1 | 1 | 1 | \( \sigma_{23} \) | 2 | 1 | 1 | 1 |
| \( \sigma_{24} \) | 2 | 1 | 2 | 3 | \( \sigma_{24} \) | 2 | 1 | 2 | 3 | \( \sigma_{24} \) | 2 | 1 | 2 | 3 |
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Figure 2: CCA and CLA for the running example

(a) 2-CCA

| $\sigma_i$ | $F_1$ | $F_2$ | $F_3$ | $F_4$ |
|-----------|-------|-------|-------|-------|
| $\sigma_1$ | 0     | 0     | 0     | 0     |
| $\sigma_2$ | 0     | 0     | 0     | 3     |
| $\sigma_3$ | 0     | 1     | 1     | 1     |
| $\sigma_4$ | 0     | 1     | 2     | 2     |
| $\sigma_5$ | 1     | 0     | 0     | 3     |
| $\sigma_6$ | 1     | 0     | 0     | 3     |
| $\sigma_7$ | 1     | 0     | 1     | 1     |
| $\sigma_8$ | 1     | 1     | 1     | 0     |
| $\sigma_9$ | 2     | 0     | 0     | 1     |
| $\sigma_{10}$ | 2     | 0     | 0     | 3     |
| $\sigma_{11}$ | 2     | 0     | 1     | 2     |
| $\sigma_{12}$ | 2     | 1     | 2     | 0     |

(b) (1,2)-CLA

| $\sigma_i$ | $F_1$ | $F_2$ | $F_3$ | $F_4$ |
|-----------|-------|-------|-------|-------|
| $\sigma_1$ | 0     | 0     | 0     | 0     |
| $\sigma_2$ | 0     | 0     | 0     | 3     |
| $\sigma_3$ | 0     | 0     | 1     | 1     |
| $\sigma_4$ | 0     | 1     | 1     | 2     |
| $\sigma_5$ | 0     | 1     | 2     | 0     |
| $\sigma_6$ | 1     | 0     | 0     | 2     |
| $\sigma_7$ | 1     | 0     | 0     | 3     |
| $\sigma_8$ | 1     | 0     | 1     | 2     |
| $\sigma_9$ | 1     | 1     | 1     | 1     |
| $\sigma_{10}$ | 1     | 1     | 2     | 0     |
| $\sigma_{11}$ | 1     | 1     | 2     | 2     |
| $\sigma_{12}$ | 2     | 0     | 0     | 1     |
| $\sigma_{13}$ | 2     | 0     | 0     | 3     |
| $\sigma_{14}$ | 2     | 0     | 2     | 0     |
| $\sigma_{15}$ | 2     | 1     | 1     | 0     |
| $\sigma_{16}$ | 2     | 1     | 1     | 2     |
| $\sigma_{17}$ | 2     | 1     | 2     | 1     |

is also a $(t-1)$-CCA when $t > 0$. Thus, a $t$-CCA can also be defined as follows.

$t$-CCA $\forall T \in \overline{I} \setminus F_A(T) \neq \emptyset$

Figure 2a shows a 2-CCA for the running example. All four invalid 2-way interactions which violate the constraints are listed as follows.

$\{ (F_2, 1), (F_3, 0) \} \{ (F_2, 1), (F_4, 3) \}$
$\{ (F_3, 2), (F_4, 3) \} \{ (F_3, 1), (F_4, 3) \}$

It is easy to observe that none of the invalid interactions appears in any rows in Figure 2a.

2.3.2. Constrained locating arrays

LAs allow us to identify the set of faulty interactions using the test outcome. This becomes possible because for any LA $A$, $\rho_A(\cdot)$ injectively maps an interaction set to a test outcome (i.e., $\rho_A(T_1) = \rho_A(T_2) \Rightarrow T_1 = T_2$). Since a test outcome corresponds to at most one interaction set, the set of faulty interactions can be uniquely inferred from the test outcome.

When incorporating constraints into LAs, it is necessary to handle the situation where constraints may prevent some sets of faulty interactions from being identified. For the running example, when either one of the two interaction sets shown below is faulty, it is impossible to determine which is indeed faulty.

$T_1 = \{ (F_2, 0), (F_4, 3) \}$  $T_2 = \{ (F_3, 0), (F_4, 3) \}$

This is because the two interactions always appear simultaneously in any valid test case and thus no valid test case exists that yields different outcomes for the two faulty interaction sets. We say that two sets of valid interactions, $T$ and $T'$, are not **distinguishable** or **indistinguishable** if $\rho_A(T) = \rho_A(T')$ for any array $A$ that consists of valid tests.

The definition of CLAs adapts LAs to the presence of indistinguishable interaction sets by exempting them from fault identification.

**Definition 2 (CLA).** Let $d \geq 0$ and $0 \leq t \leq k$. An array $A$ that consists of valid tests is a $(d, t)$-, $(\overline{d}, t)$-, $(d, \overline{t})$- or
(\bar{d}, \bar{t})\text{-CLA} iff the corresponding condition shown below holds.

\( (d, t)\text{-CLA} \quad \forall \mathcal{T}_1, \mathcal{T}_2 \subseteq \mathcal{V} \mathcal{I}_s \text{ such that } |\mathcal{T}_1| = |\mathcal{T}_2| = d \text{ and } \mathcal{T}_1, \mathcal{T}_2 \text{ are distinguishable :} \\
\rho_A(\mathcal{T}_1) \neq \rho_A(\mathcal{T}_2) \)

\( (\bar{d}, t)\text{-CLA} \quad \forall \mathcal{T}_1, \mathcal{T}_2 \subseteq \mathcal{V} \mathcal{I}_s \text{ such that } 0 \leq |\mathcal{T}_1| \leq d, 0 \leq |\mathcal{T}_2| \leq d \text{ and } \mathcal{T}_1, \mathcal{T}_2 \text{ are distinguishable :} \\
\rho_A(\mathcal{T}_1) \neq \rho_A(\mathcal{T}_2) \)

\( (d, \bar{t})\text{-CLA} \quad \forall \mathcal{T}_1, \mathcal{T}_2 \subseteq \mathcal{V} \mathcal{I}_s \text{ such that } |\mathcal{T}_1| = |\mathcal{T}_2| = d \text{ and } \mathcal{T}_1, \mathcal{T}_2 \text{ are independent and distinguishable :} \\
\rho_A(\mathcal{T}_1) \neq \rho_A(\mathcal{T}_2) \)

\( (\bar{d}, \bar{t})\text{-CLA} \quad \forall \mathcal{T}_1, \mathcal{T}_2 \subseteq \mathcal{V} \mathcal{I}_s \text{ such that } 0 \leq |\mathcal{T}_1| \leq d, 0 \leq |\mathcal{T}_2| \leq d \text{ and } \mathcal{T}_1, \mathcal{T}_2 \text{ are independent and distinguishable :} \\
\rho_A(\mathcal{T}_1) \neq \rho_A(\mathcal{T}_2) \)

A (1,2)-CLA for the running example is shown in Figure 2b. From the array, one can see that none of the invalid interactions shown above appears in any rows of the CLA. In addition, for any pair of 2-way interactions except the above indistinguishable pair, the rows that cover one of them are different from those that cover the other. The exception is the rows where the indistinguishable pair of interaction sets appear, namely \( \sigma_2, \sigma_7 \text{ and } \sigma_{13} \). This occurs because of the second constraint \( \varphi_2 \) in the SUT. \( \varphi_2 \) enforces all test cases that have \( (F_3, 3) \text{ to contain } (F_2, 0) \text{ and } (F_3, 0) \text{ at the same time. Thus, the interaction sets } \mathcal{T}_1 = \{ ((F_2, 0), (F_4, 3)) \} \text{ and } \mathcal{T}_2 = \{ ((F_3, 0), (F_4, 3)) \} \text{ appear in the same test cases as long as the test cases are valid.} \)

3. Constrained Detecting Arrays

In this section, we propose Constrained Detecting Arrays (CDAs). First we present the definition of CDAs; then we show how one can identify faulty interactions with CDAs. In addition, we provide some theorems that relate CDAs to CCAs and CLAs.

3.1. Definition

For an array \( A \) to be a DA, \( A \) must satisfy \( \rho_A(T) \subseteq \rho_A(T) \iff T \in \mathcal{T} \) for any pair of an interaction \( T \) and an interaction set \( \mathcal{T} \). However, this may be impossible if \( A \) consists only of valid test cases. Here we introduce the concept of masking to capture such a situation.

**Definition 3 (Masking).** A set \( \mathcal{T} \) of valid interactions masks a valid interaction \( T \) iff \( T \notin \mathcal{T} \) and

\[
\forall \sigma \in R : T \subseteq \sigma \Rightarrow (\exists T' \in \mathcal{T} : T' \subseteq \sigma).
\]

If \( \mathcal{T} \) masks \( T \), we write \( \mathcal{T} \succ T \); otherwise we write \( \mathcal{T} \npreceq T \). By definition, \( \mathcal{T} \npreceq T \) iff \( T \in \mathcal{T} \) or

\[
\exists \sigma \in R : T \subseteq \sigma \Rightarrow (\forall T' \in \mathcal{T} : T' \subseteq \sigma).
\]

In words, when \( \mathcal{T} \) masks \( T \), \( T \) always appears together with some interaction \( T' \) in \( \mathcal{T} \) in any valid test case \( \sigma \). In this case, \( T \notin \mathcal{T} \) but \( \rho_A(T) \subseteq \rho_A(T) \) always holds for any \( A \) that meets the constraints. In the running example, such \( T\mathcal{T} \) pairs include:

\[
\begin{align*}
\mathcal{T}_1 &= \{ ((F_1, 0), (F_2, 0)) \} \succ T_0 = \{ (F_1, 0), (F_3, 0) \} & \mathcal{T}_1 &= \{ ((F_1, 0), (F_2, 0)) \} \succ T_b = \{ (F_1, 0), (F_4, 3) \} \\
\mathcal{T}_2 &= \{ ((F_1, 1), (F_2, 0)) \} \succ T_c = \{ (F_1, 1), (F_3, 0) \} & \mathcal{T}_3 &= \{ ((F_2, 0), (F_3, 0)) \} \succ T_b = \{ (F_1, 0), (F_4, 3) \} \\
\mathcal{T}_4 &= \{ ((F_3, 0), (F_4, 3)) \} \succ T_d = \{ (F_1, 2), (F_4, 3) \} & \mathcal{T}_4 &= \{ ((F_3, 0), (F_4, 3)) \} \succ T_e = \{ (F_2, 0), (F_4, 3) \} \\
& \cdots \cdots (31 \text{ pairs in total})
\end{align*}
\]

When \( \mathcal{T} \) masks \( T \), the failure caused by \( T \) cannot be inherently distinguished from that caused by \( \mathcal{T} \). The idea at the core of CDAs is to relax the condition of DAs by exempting \( T\mathcal{T} \) pairs such that \( \mathcal{T} \succ T \) from fault identification.
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Figure 3: CDAs for the running example

| (a) (1,1)-CDA | (b) (2,1)-CDA | (c) (1,2)-CDA |
|--------------|---------------|---------------|
| $F_1$ | $F_2$ | $F_3$ | $F_4$ | $F_1$ | $F_2$ | $F_3$ | $F_4$ | $F_1$ | $F_2$ | $F_3$ | $F_4$ |
| $\sigma_1$ | 0 | 0 | 0 | 0 | $\sigma_1$ | 0 | 0 | 0 | 0 | $\sigma_1$ | 0 | 0 | 0 | 0 |
| $\sigma_2$ | 0 | 1 | 2 | 1 | $\sigma_2$ | 0 | 0 | 0 | 3 | $\sigma_2$ | 0 | 0 | 0 | 1 |
| $\sigma_3$ | 1 | 0 | 0 | 3 | $\sigma_3$ | 0 | 0 | 2 | 1 | $\sigma_3$ | 0 | 0 | 0 | 3 |
| $\sigma_4$ | 1 | 0 | 1 | 1 | $\sigma_4$ | 0 | 1 | 1 | 2 | $\sigma_4$ | 0 | 0 | 1 | 1 |
| $\sigma_5$ | 1 | 1 | 1 | 2 | $\sigma_5$ | 0 | 1 | 2 | 0 | $\sigma_5$ | 0 | 0 | 2 | 2 |
| $\sigma_6$ | 2 | 0 | 0 | 3 | $\sigma_6$ | 1 | 0 | 1 | 1 | $\sigma_6$ | 0 | 1 | 1 | 2 |
| $\sigma_7$ | 2 | 0 | 2 | 2 | $\sigma_7$ | 1 | 0 | 0 | 1 | $\sigma_7$ | 0 | 1 | 2 | 0 |
| $\sigma_8$ | 2 | 1 | 1 | 0 | $\sigma_8$ | 1 | 0 | 0 | 2 | $\sigma_8$ | 0 | 1 | 2 | 1 |
| $\sigma_9$ | 1 | 0 | 0 | 3 | $\sigma_9$ | 1 | 0 | 0 | 2 | $\sigma_9$ | 1 | 0 | 0 | 3 |
| $\sigma_{10}$ | 1 | 1 | 1 | 0 | $\sigma_{10}$ | 1 | 0 | 0 | 3 | $\sigma_{10}$ | 1 | 0 | 0 | 3 |
| $\sigma_{11}$ | 1 | 1 | 2 | 1 | $\sigma_{11}$ | 1 | 0 | 1 | 0 | $\sigma_{11}$ | 1 | 0 | 1 | 0 |
| $\sigma_{12}$ | 2 | 0 | 0 | 3 | $\sigma_{12}$ | 1 | 0 | 2 | 1 | $\sigma_{12}$ | 1 | 0 | 2 | 1 |
| $\sigma_{13}$ | 2 | 0 | 1 | 2 | $\sigma_{13}$ | 1 | 1 | 1 | 1 | $\sigma_{13}$ | 1 | 1 | 1 | 1 |
| $\sigma_{14}$ | 2 | 0 | 2 | 0 | $\sigma_{14}$ | 1 | 1 | 2 | 0 | $\sigma_{14}$ | 1 | 1 | 2 | 0 |
| $\sigma_{15}$ | 2 | 1 | 1 | 1 | $\sigma_{15}$ | 1 | 1 | 2 | 2 | $\sigma_{15}$ | 1 | 1 | 2 | 2 |
| $\sigma_{16}$ | 2 | 1 | 2 | 2 | $\sigma_{16}$ | 2 | 0 | 0 | 0 | $\sigma_{16}$ | 2 | 0 | 0 | 0 |
| $\sigma_{17}$ | 2 | 0 | 0 | 1 | $\sigma_{17}$ | 2 | 0 | 0 | 1 | $\sigma_{17}$ | 2 | 0 | 0 | 1 |
| $\sigma_{18}$ | 2 | 0 | 0 | 2 | $\sigma_{18}$ | 2 | 0 | 0 | 2 | $\sigma_{18}$ | 2 | 0 | 0 | 2 |
| $\sigma_{19}$ | 2 | 0 | 0 | 3 | $\sigma_{19}$ | 2 | 0 | 0 | 3 | $\sigma_{19}$ | 2 | 0 | 0 | 3 |
| $\sigma_{20}$ | 2 | 0 | 1 | 2 | $\sigma_{20}$ | 2 | 0 | 1 | 2 | $\sigma_{20}$ | 2 | 0 | 1 | 2 |
| $\sigma_{21}$ | 2 | 0 | 2 | 0 | $\sigma_{21}$ | 2 | 0 | 2 | 0 | $\sigma_{21}$ | 2 | 0 | 2 | 0 |
| $\sigma_{22}$ | 2 | 1 | 1 | 0 | $\sigma_{22}$ | 2 | 1 | 1 | 0 | $\sigma_{22}$ | 2 | 1 | 1 | 0 |
| $\sigma_{23}$ | 2 | 1 | 1 | 1 | $\sigma_{23}$ | 2 | 1 | 1 | 1 | $\sigma_{23}$ | 2 | 1 | 1 | 1 |
| $\sigma_{24}$ | 2 | 1 | 2 | 2 | $\sigma_{24}$ | 2 | 1 | 2 | 2 | $\sigma_{24}$ | 2 | 1 | 2 | 2 |

Definition 4 (CDA). Let $d \geq 0$ and $0 \leq t \leq k$. An array $A$ that consists of valid test cases or no rows is a $(d,t)$-, $(\bar{d},t)$-, $(d,\bar{t})$- or $(\bar{d},\bar{t})$-CDA iff the corresponding condition shown below holds.

$$(d,t)$-CDA \quad \forall T \subseteq \mathcal{V}I, \text{ such that } |T| = d, \forall T \subseteq \mathcal{V}I, : \quad T \nsubseteq T \Rightarrow (T \in T \Leftrightarrow \rho_A(T) \subseteq \rho_A(T))$$

$(\bar{d},t)$-CDA \quad \forall T \subseteq \mathcal{V}I, \text{ such that } 0 \leq |T| \leq d, \forall T \subseteq \mathcal{V}I, : \quad T \nsubseteq T \Rightarrow (T \in T \Leftrightarrow \rho_A(T) \subseteq \rho_A(T))$

$(d,\bar{t})$-CDA \quad \forall T \subseteq \mathcal{V}I, \text{ such that } |T| = d, \forall T \subseteq \mathcal{V}I, \text{ and } T \cup \{T\} \text{ is independent : } \quad T \nsubseteq T \Rightarrow (T \in T \Leftrightarrow \rho_A(T) \subseteq \rho_A(T))$

$(\bar{d},\bar{t})$-CDA \quad \forall T \subseteq \mathcal{V}I, \text{ such that } 0 \leq |T| \leq d, \forall T \subseteq \mathcal{V}I, \text{ and } T \cup \{T\} \text{ is independent : } \quad T \nsubseteq T \Rightarrow (T \in T \Leftrightarrow \rho_A(T) \subseteq \rho_A(T))$

Figure 3a. Figure 3b, and Figure 3c respectively show a (1,1)-CDA, a (2,1)-CDA, and a (1,2)-CDA for the running example. Now let us take the (1,2)-CDA in Figure 3c and the pair of $T_3 = \{ ((F_2,0), (F_3,0)) \}$ and $T_b = \{ (F_1,0), (F_4,3) \}$ as examples. Let $A$ denote the (1,2)-CDA for now; then $\rho_A(T_3) = \{ \sigma_1, \sigma_2, \sigma_3, \sigma_9, \sigma_{10}, \sigma_{16}, \sigma_{17}, \sigma_{18}, \sigma_{19} \}$, whereas $\rho_A(T_b) = \{ \sigma_3 \}$. Hence, $T_b \in T_3 \Leftrightarrow \rho_A(T_b) \subseteq \rho_A(T_3)$ does not hold. This is prohibited by the definition of DAs but is allowed by CDAs, because $T_3 > T_b$, which means that no array can satisfy it unless the constraints are violated.

By definition, it is straightforward to see that the following observations hold.

Observation 1. A $(d,\bar{t})$-CDA is a $(\bar{d},\bar{t})$-CDA and a $(d,\bar{t})$-CDA. A $(\bar{d},t)$-CDA and a $(d,\bar{t})$-CDA are both a $(d,t)$-CDA. When $d > 0$, a $(\bar{d},\bar{t})$-CDA and a $(d,\bar{t})$-CDA are a $(\bar{d} - 1,\bar{t})$-CDA and a $(d - 1,\bar{t})$-CDA, respectively. When $t > 0$, a $(\bar{d},\bar{t})$-CDA and a $(d,\bar{t})$-CDA are a $(\bar{d},t - 1)$-CDA and a $(d,t - 1)$-CDA, respectively.
Observation 2. Suppose that the SUT has no constraints, i.e., \( \phi(\sigma) = true \) for all \( \sigma \in R = V_1 \times V_2 \times \cdots \times V_k \) and that a \((d, t)\)-DA \( A \) exists. Then \( A \) is a \((d, t)\)-CDA. This also applies when \( d \) or \( t \) is replaced with \( \bar{d} \) or \( \bar{t} \), respectively.

According to the above definition, if \( |T| \) is very large, then \( \rho_A(T) = A \) for any array \( A \), in which case all interactions are masked by \( T \). In order to avoid such cases of no practical interest, here we introduce an upper bound, denoted \( \tau_i \), on \( d \). We let \( \tau_i \geq 0 \) be the greatest integer that satisfies the condition as follows:

\[
\forall T \subseteq V_{I_i} \text{ such that } |T| \leq \tau_i : R - \rho_R(T) \neq \emptyset
\]

In words, given \( \tau_i \) interactions of strength \( t \), there is always a test case in \( R \) in which none of the given interactions appears. Note that \( \tau_0 = 0 \).

Theorem 1. For \( d \leq \tau_i \), a \((d, t)\)-CDA is equivalent to a \((\bar{d}, t)\)-CDA.

**Proof.** Trivially a \((0, t)\)-CDA is a \((\bar{0}, t)\)-CDA. Let \( A \) be a \((d, t)\)-CDA such that \( 1 \leq d \leq \tau_i \) and \( t > 0 \). We will show that \( A \) is a \((d - 1, t)\)-CDA. Let \( T \) and \( T' \) be a set of \( d - 1 \) valid interactions of strength \( t \) and a \( t \)-way valid interaction, respectively. If \( T' \in T \), then \( T' \notin T \Rightarrow (T \in T \Leftrightarrow \rho_A(T) \subseteq \rho_A(T)) \) trivially holds. The rest of the proof considers the case where \( T \notin T \). In this case, there is some \( \sigma \in R \) such that \( T \subseteq \sigma \) and \( \sigma \notin \rho_R(T) \). Since \( |T \cup \{T\}| \leq \tau_i \), \( R - \rho_R(T \cup \{T\}) \) is not empty. Let \( T' \) be any \( t \)-way interaction that appears in a test case in \( R - \rho_R(T \cup \{T\}) \) and has exactly the same \( t \) parameters as \( T \). Note that \( T \) and \( T' \) cannot appear in any test case simultaneously. Let \( T' = T \cup \{T'\} \). \( T' \notin T \) since \( T \subseteq \sigma \), \( \sigma \notin \rho_R(T) \), and \( \sigma \notin \rho_R(T') \). Because \( A \) is a \((d, t)\)-CDA and \( T' \notin T \), \( \rho_A(T') \notin \rho_A(T) \). Hence \( \rho_A(T) \notin \rho_A(T') \). By induction, \( A \) is a \((d', t)\)-CDA for any \( 0 \leq d' \leq d \) and thus is a \((\bar{d}, t)\)-CDA.

A similar argument applies to \((d, \bar{t})\)-CDAs and \((\bar{d}, \bar{t})\)-CDAs.

Theorem 2. For \( d = t = 0 \) or \( d \leq \tau_i \) and \( t > 0 \), a \((d, \bar{t})\)-CDA is equivalent to a \((\bar{d}, \bar{t})\)-CDA.

**Proof.** Trivially \((0, \bar{t})\)-CDA is a \((\bar{0}, \bar{t})\)-CDA. Let \( A \) be a \((d, \bar{t})\)-CDA such that \( 1 \leq d \leq \tau_i \) and \( t > 0 \). Below we will show that \( A \) is a \((d - 1, \bar{t})\)-CDA. Let \( T \subseteq \overline{V_{I_i}} \) such that \( T \) is independent and \( |T| = d - 1 \). Let \( T \) be a valid interaction of strength at most \( t \). If \( T' \in T \) or \( T \notin T \), then \( T' \notin T \Rightarrow (T \in T \Leftrightarrow \rho_A(T) \subseteq \rho_A(T)) \) trivially holds.

Parallel sets lead to the relationships illustrated in Figure 4. Because of these results, we henceforth focus on \((\bar{d}, d)\)-CDAs and \((\bar{d}, \bar{d})\)-CDAs.
Figure 5: 2-CCA, (\(\overline{T}, 2\))-CLA, and test outcomes in Case 1 and Case 2.

|       | \(F_1\) | \(F_2\) | \(F_3\) | \(F_4\) | Case1 | Case2 |
|-------|---------|---------|---------|---------|-------|-------|
| \(\sigma_1\) | 0       | 0       | 0       | 0       | Fail  | Fail  |
| \(\sigma_2\) | 0       | 0       | 0       | 3       | Fail  | Fail  |
| \(\sigma_3\) | 0       | 1       | 1       | 1       | Pass  | Fail  |
| \(\sigma_4\) | 0       | 1       | 1       | 2       | Pass  | Pass  |
| \(\sigma_5\) | 1       | 0       | 0       | 2       | Pass  | Pass  |
| \(\sigma_6\) | 1       | 0       | 0       | 3       | Pass  | Pass  |
| \(\sigma_7\) | 1       | 0       | 2       | 1       | Pass  | Pass  |
| \(\sigma_8\) | 1       | 1       | 1       | 1       | Pass  | Pass  |
| \(\sigma_9\) | 2       | 0       | 0       | 1       | Pass  | Pass  |
| \(\sigma_{10}\) | 2     | 0       | 0       | 3       | Pass  | Pass  |
| \(\sigma_{11}\) | 2     | 0       | 1       | 2       | Pass  | Pass  |
| \(\sigma_{12}\) | 2     | 1       | 2       | 0       | Pass  | Pass  |

|       | \(F_1\) | \(F_2\) | \(F_3\) | \(F_4\) | Case1 | Case2 |
|-------|---------|---------|---------|---------|-------|-------|
| \(\sigma_1\) | 0       | 0       | 0       | 0       | Fail  | Fail  |
| \(\sigma_2\) | 0       | 0       | 0       | 3       | Fail  | Fail  |
| \(\sigma_3\) | 0       | 0       | 1       | 1       | Fail  | Fail  |
| \(\sigma_4\) | 0       | 1       | 1       | 2       | Pass  | Pass  |
| \(\sigma_5\) | 0       | 1       | 2       | 0       | Pass  | Pass  |
| \(\sigma_6\) | 1       | 0       | 0       | 2       | Pass  | Pass  |
| \(\sigma_7\) | 1       | 0       | 0       | 3       | Pass  | Pass  |
| \(\sigma_8\) | 1       | 0       | 1       | 2       | Pass  | Pass  |
| \(\sigma_9\) | 1       | 1       | 1       | 1       | Pass  | Pass  |
| \(\sigma_{10}\) | 1     | 1       | 2       | 0       | Pass  | Pass  |
| \(\sigma_{11}\) | 1     | 1       | 2       | 2       | Pass  | Pass  |
| \(\sigma_{12}\) | 2     | 0       | 0       | 1       | Pass  | Pass  |
| \(\sigma_{13}\) | 2     | 0       | 0       | 3       | Pass  | Pass  |
| \(\sigma_{14}\) | 2     | 0       | 2       | 0       | Pass  | Pass  |
| \(\sigma_{15}\) | 2     | 1       | 1       | 0       | Pass  | Pass  |
| \(\sigma_{16}\) | 2     | 1       | 1       | 2       | Pass  | Pass  |
| \(\sigma_{17}\) | 2     | 1       | 2       | 1       | Pass  | Pass  |

Theorem 3. A \((\overline{d}, t)\)-CDA is a \(t\)-CCA. A \((\overline{d}, \overline{t})\)-CDA is a \(t\)-CCA.

Proof. Let \(T \subseteq \mathcal{V}I_t\). Let \(A\) be a \((\overline{d}, t)\)-CDA or a \((\overline{d}, \overline{t})\)-CDA. Then \(T \not\supset T \Rightarrow (T \in T \iff \rho_A(T) \subseteq \rho_A(T))\) for any \(T \subseteq \mathcal{V}I_t\) such that \(|T| \leq d\). If \(|T| = 0\), then \(T = \emptyset\), in which case \(T \not\supset T, T \not\in T\), and \(\rho_A(T) = \emptyset\). Hence \(\rho_A(T) \neq \emptyset\).

### 3.2. Identification of faulty interactions

Using a CDA as a test suite, faulty interactions are identified as follows. The execution of a test suite yields a test outcome which is a set of failed test cases and a set of passing test cases. Every interaction of the target strength is determined to be faulty or not faulty, except when an exhaustive test suite is used. When the assumption on the number of faulty interactions is false, it is not possible to identify all faulty interactions. This is because in that case some faulty interaction may not appear in any test case, unless an exhaustive test suite is used.

Next consider the case where the assumption on the number of faulty interactions is false. That is, there are more than \(d\) faulty interactions. In this case, interactions \(T\) can be falsely determined to be faulty even if \(T \not\supset T\). However, all faulty interactions are correctly identified as faulty, because all valid \(t\)-way interactions appear in \(A\) (Theorem 3). When the assumption on the strength is false, it is not possible to identify all faulty interactions. This is because in that case some faulty interaction may not appear in any test case, unless an exhaustive test suite is used.

The situation is similar when \(A\) is a \((\overline{d}, \overline{t})\)-CDA. In this case, the assumptions are: \(|T| \leq d\) and the strength of the faulty interactions is at most \(t\).

When these assumptions are true, \(T(\subseteq \overline{\mathcal{V}I_t})\) is accurately determined to be faulty or not faulty unless \(T > T\) or \(T \cup \{T\}\) is not independent, since \(T \in T \iff \rho_A(T) \subseteq \rho_A(T)\).

But even when \(T \cup \{T\}\) is not independent, accurate identification is still possible if \(T\) contains neither proper subsets nor proper supersets of \(T\) and \(T \not\supset T\). In that case, if we let \(T_{\text{min}} = \{T' \in T : T' \not\supset T' \text{ for all } T'' \in T\}\) (i.e., \(T_{\text{min}} \subseteq T\)) is the set of minimal interactions in \(T\), then \(T_{\text{min}} \not\supset T\) and \(T_{\text{min}} \cup \{T\}\) becomes independent, and thus...
Figure 6: $(\bar{1}, 2)$-CDA and test outcomes in Case 1 and Case 2.

|σ| $F_1$ | $F_2$ | $F_3$ | $F_4$ | Case 1 | Case 2 |
|---|---|---|---|---|---|---|
| $\sigma_1$ | 0 | 0 | 0 | 0 | Fail | Fail |
| $\sigma_2$ | 0 | 0 | 0 | 1 | Fail | Fail |
| $\sigma_3$ | 0 | 0 | 0 | 3 | Fail | Fail |
| $\sigma_4$ | 0 | 0 | 1 | 1 | Fail | Fail |
| $\sigma_5$ | 0 | 1 | 1 | 2 | Pass | Pass |
| $\sigma_6$ | 0 | 1 | 2 | 0 | Pass | Pass |
| $\sigma_7$ | 0 | 1 | 2 | 1 | Pass | Fail |
| $\sigma_8$ | 1 | 0 | 0 | 2 | Pass | Pass |
| $\sigma_9$ | 1 | 0 | 0 | 3 | Pass | Pass |
| $\sigma_{10}$ | 1 | 0 | 1 | 0 | Pass | Pass |
| $\sigma_{11}$ | 1 | 0 | 2 | 2 | Pass | Pass |
| $\sigma_{12}$ | 1 | 1 | 1 | 1 | Pass | Pass |
| $\sigma_{13}$ | 1 | 1 | 1 | 1 | Pass | Pass |
| $\sigma_{14}$ | 1 | 1 | 2 | 0 | Pass | Pass |
| $\sigma_{15}$ | 1 | 1 | 2 | 2 | Pass | Pass |
| $\sigma_{16}$ | 2 | 0 | 0 | 0 | Pass | Pass |
| $\sigma_{17}$ | 2 | 0 | 0 | 1 | Pass | Pass |
| $\sigma_{18}$ | 2 | 0 | 0 | 2 | Pass | Pass |
| $\sigma_{19}$ | 2 | 0 | 0 | 3 | Pass | Pass |
| $\sigma_{20}$ | 2 | 0 | 1 | 2 | Pass | Pass |
| $\sigma_{21}$ | 2 | 0 | 2 | 0 | Pass | Pass |
| $\sigma_{22}$ | 2 | 1 | 1 | 0 | Pass | Pass |
| $\sigma_{23}$ | 2 | 1 | 1 | 1 | Pass | Pass |
| $\sigma_{24}$ | 2 | 1 | 2 | 2 | Pass | Pass |

$T \in T_{\min} \Leftrightarrow \rho_A(T) \in \rho_A(T_{\min})$. Also $T \in \mathcal{T} \Leftrightarrow T \in T_{\min}$ and $\rho_A(T) = \rho_A(T_{\min})$. Consequently $T \in \mathcal{T} \Leftrightarrow \rho_A(T) \subseteq \rho_A(\mathcal{T})$.

In sum, faulty interactions are all identified as faulty and a non-faulty interaction may be identified as faulty only if the set of the faulty interactions masks the non-faulty interaction or contains its proper subsets or supersets. When $|\mathcal{T}| > d$, non-faulty interactions might be falsely identified as faulty; but all faulty interactions are correctly identified as faulty. The faulty interactions can be correctly identified because they only appear in the failed test cases. When the assumption on the strength is not false, it is not possible to identify all faulty interactions.

### 3.2.1. Examples

Here using the running example, we illustrate how CCA, CLA and CDA arrays are used to detect and locate faulty interactions. We consider the cases $d = 1$ and $t = 2$. Suppose that the 2-way CCA, the $(1, 2)$-CLA, and the $(1, 2)$-CDA shown in Figures 2a, 2b, and Figure 3c are used as test suites. In fact the CLA and the CDA are a $(\bar{1}, 2)$-CLA and a $(1, 2)$-CDA. Figure 5 and Figure 6 summarize the results of test cases when executed in the two cases below.

**Case 1** The only faulty interaction is $T_a = \{(F_1, 0), (F_2, 0)\}$.

**Case 2** There are two faulty interactions $T_b = \{(F_1, 0), (F_3, 0)\}$ and $T_c = \{(F_1, 0), (F_4, 1)\}$.

**CCA** In Case 1, within the test cases in the 2-CCA (Figure 2a), only the test cases $\sigma_1$ and $\sigma_3$ fail. The two-way interactions that appear only in those failed test cases are as follows (the faulty interaction is indicated by underline.)

$$
\{(F_1, 0), (F_2, 0)\}, \{(F_1, 0), (F_3, 0)\}, \{(F_1, 0), (F_4, 0)\} \\
\{(F_2, 0), (F_4, 0)\}, \{(F_3, 0), (F_4, 0)\}, \{(F_1, 0), (F_4, 3)\}
$$
In Case 2, the failed test cases are $\sigma_1$, $\sigma_2$, and $\sigma_3$; thus the candidates for faulty interactions are:

$$\{(F_1, 0), (F_2, 0)\} \quad \{(F_1, 0), (F_3, 0)\} \quad \{(F_1, 0), (F_4, 0)\}$$

$$\{(F_2, 0), (F_4, 0)\} \quad \{(F_3, 0), (F_4, 0)\} \quad \{(F_1, 0), (F_4, 3)\}$$

$$\{(F_1, 0), (F_1, 1)\} \quad \{(F_1, 0), (F_4, 1)\} \quad \{(F_2, 1), (F_4, 1)\}$$

$$\{(F_1, 1), (F_4, 1)\}$$

For both cases it is impossible to further reduce the candidates of faulty interactions.

**CLA** Suppose that the (1, 2)-CLA $(\overline{1}, 2\text{-CLA})$ shown in Figure 2b is used. In Case 1, the test cases $\sigma_1$, $\sigma_2$ and $\sigma_3$ fail and all the other test cases pass. The interactions that appear only in the failed test cases are as follows.

$$\{(F_1, 0), (F_2, 0)\} \quad \{(F_1, 0), (F_3, 0)\} \quad \{(F_3, 0), (F_4, 0)\}$$

$$\{(F_1, 0), (F_4, 3)\}$$

The core idea of CLAs is that it allows a test outcome to be uniquely associated with a set of faulty interactions, which is mathematically represented as $T_1 = T_2 \iff \rho_A(T_1) = \rho_A(T_2)$. In this case, $\rho_A(T) = \rho_A(\overline{T_a}) = \{\sigma_1, \sigma_2, \sigma_3\}$ holds only for $T = \{(F_1, 0), (F_2, 0)\}$, provided that $S \subseteq \mathcal{I}_2$ and $|S| \leq 1$. Thus, we correctly locate the faulty interaction.

Now consider Case 2. The failed test cases are the same as in Case 1, i.e., $\sigma_1$, $\sigma_2$ and $\sigma_3$. Hence the conclusion that $\overline{T_a}$ is the only faulty interaction is also the same. This incorrect result is caused by the fact that the number of faulty interactions does not coincide with the assumption (namely $d = 1$). In general, if faulty interactions exceed the number assumed, CLAs may identify non-faulty interactions as faulty but also identify faulty interactions as non-faulty.

**CDA** Suppose that the (1, 2)-CDA $(\overline{1}, 2\text{-CDA})$ shown in Figure 3c is used to locate faulty interactions. For Case 1, the failed test cases are $\sigma_1$, $\sigma_2$, $\sigma_3$, $\sigma_4$ and $\sigma_5$. The interactions occurring only in the failed test cases are all identified as faulty. In this case these interactions are:

$$\{(F_1, 0), (F_2, 0)\} \quad \{(F_1, 0), (F_3, 0)\} \quad \{(F_1, 0), (F_4, 3)\}$$

$T_a$ is correctly identified as faulty, whereas $\{(F_1, 0), (F_3, 0)\}$ and $\{(F_1, 0), (F_4, 3)\}$ are incorrectly identified as faulty. Since $\overline{T_a}$ masks $\{(F_1, 0), (F_3, 0)\}$ and $\{(F_1, 0), (F_4, 3)\}$, $(\overline{T_a}) \neq \{(F_1, 0), (F_3, 0)\}$ and $\overline{T_a} \neq \{(F_1, 0), (F_4, 3)\}$, it is inherently impossible to determine that $\{(F_1, 0), (F_2, 0)\}$ and $\{(F_1, 0), (F_4, 3)\}$ are not faulty when $\overline{T_a}$ is faulty. However, it should be noted that if we relied on the assumption that the number of faulty interactions is $d = 1$, just as in the case of the CLA above, we could correctly identify only $\overline{T_a}$ as faulty. In fact, we will show later that any CDA is a CLA.

For Case 2, the failed test cases are $\sigma_1$, $\sigma_2$, $\sigma_3$, $\sigma_4$, and $\sigma_5$. The interactions that are identified as faulty are:

$$\{(F_1, 0), (F_2, 0)\} \quad \{(F_1, 0), (F_3, 0)\} \quad \{(F_1, 0), (F_4, 3)\}$$

Although the last interaction is in fact not faulty, all the faulty ones are correctly identified. In general, when using a CDA, non-faulty interactions are never wrongly identified as faulty even if the number of faulty interactions exceeds the assumed number $d$.

### 3.3. Properties of CDAs

In the rest of the section we provide some theorems on the properties of CDAs.

**Theorem 4.** $\mathcal{R}$, the exhaustive test suite, is a $(d, t)$-, $(d, \overline{t})$-, $(\overline{d}, t)$- and $(\overline{d}, \overline{t})$-CDA for any $d$ and $t$.

**Proof.** Let $T$ be a valid interaction and $S$ be a set of valid interactions. Below we will show $T \not\supseteq T \Rightarrow (T \in T \iff \rho_R(T) \subseteq \rho_R(S))$. If $T \not\supseteq T$ and $T \not\subseteq T$, then there is some $\sigma \in R$ such that $T \not\subseteq \sigma$ and $\forall T' \in T : T' \not\subseteq \sigma$, in which case $\sigma \in \rho_R(T) - \rho_R(S)$. That is, $T \not\supseteq T \Rightarrow (T \not\subseteq T \iff \rho_R(T) \not\subseteq \rho_R(S))$. In addition $T \subseteq T \Rightarrow \rho_R(T) \subseteq \rho_R(T)$ trivially holds. As a result, the theorem follows.

**Theorem 5.** A $(d, t)$-CDA is also a $(d, t)$-CLA; a $(d, \overline{t})$-CDA is also a $(d, \overline{t})$-CLA; A $(\overline{d}, t)$-CDA is also a $(\overline{d}, t)$-CLA and a $(\overline{d}, \overline{t})$-CDA is also a $(\overline{d}, \overline{t})$-CLA.
Theorem 6. A \((t+d)\)-CCA is a \((\bar{d},\bar{t})\)-CCA.

Proof. Suppose that \(A\) is a \((t+d)\)-CCA. The theorem holds if \(\rho_A(T) \subseteq \rho_A(T')\) for any \(T \in \overline{V T_i}\) and \(T \subseteq \overline{V T_i}\) such that \(0 \leq |T| \leq d, T \not\subseteq T, T \not\supseteq T,\) and \(\{T\} \cup T\) is independent. We show this by constructing a valid interaction \(\hat{T}\) of strength \(\leq d+t\) that covers \(T\) but cannot appear with any interaction in \(T\) in the same row. If such \(\hat{T}\) exists, some row of \(A\) contains it because \(A\) is a \((t+d)\)-CCA. This row is in \(\rho_A(T)\) but not in \(\rho_A(T)\); thus \(\rho_A(T) \not\subseteq \rho_A(T)\).

Since \(\hat{T} \not\supseteq T\), there must be a valid test case \(\sigma\) that covers \(T\) but does not cover any \(T' \in \hat{T}\). Let \(\sigma = (s_1, s_2, \ldots, s_k)\).

We regard \(\sigma\) as \(k\)-way interaction \(\{(F_1, s_1), (F_2, s_2), \ldots, (F_k, s_k)\}\). \(\hat{T}\) is constructed by starting from \(\hat{T} = T\) and gradually expanding it by applying the following process for all \(T' \in \hat{T}\): Select any \((F_i, v) \in T'\) such that \(s_i \neq v\). This can be done because \(T'\) is not covered by \(\sigma\) (and thus \(\lambda \not\in \hat{T}\)). Add \((F_i, s_i)\) to \(\hat{T}\). Finally \(\hat{T}\) becomes the desired interaction.

4. Generation Algorithms

In this section, we present two algorithms for generating CDAs: the satisfiability-based algorithm and the two-step heuristic algorithm. In this section we limit ourselves to \((d,t)\)-CDAs because \((d,t)\)-CDAs are \((\bar{d},\bar{t})\)-CDAs except in extreme cases (Theorem 2). Also it is straightforward to adjust the algorithms to \((d,t)\)-CDAs and \((\bar{d},\bar{t})\)-CDAs.

4.1. The satisfiability-based algorithm

The first algorithm leverages a satisfiability solver. We reduce the problem of generating a CDA of a given size to the satisfiability problem of a logical (i.e., Boolean-valued) expression. A logical expression is satisfiable iff it evaluates to true for some valuation, i.e., assignment of values to the variables. The algorithm first estimates the upper bound on the minimum size of a CDA and uses it as the initial size of a CDA. Then it creates a logical expression that is satisfiable iff a CDA of the initial size exists. The logical expression is in turn evaluated by a satisfiability solver. We design the logical expression so that the valuation that satisfies it directly represents a CDA. Satisfiability solvers can produce such a satisfying valuation when the expression is satisfiable; hence a CDA can be obtained from the output of the solver. Repeating the process while decreasing the CDA size, the algorithm can obtain the smallest CDA.

4.1.1. The logic expression

To represent an array with a collection of variables, we adopt the naive matrix model which is used by Hnich et al. [9] in their study to find CAs. In this model, an \(N \times k\) array is represented as an \(N \times k\) matrix of integer variables as follows.

\[
A = \begin{pmatrix}
  p_1^1 & \cdots & p_1^k \\
  \vdots & \ddots & \vdots \\
  p_k^1 & \cdots & p_k^N
\end{pmatrix}
\]

The variable \(p_i^j\) represents the value on the parameter \(F_i\) in the \(n\)-th test case. The domain of \(p_i^j\) is \(S_i = \{0, 1, \ldots, |S_i| - 1\}\).

In order for the array \(A\) to become a \((d,t)\)-CDA, we impose the following conditions on \(A\) using logical expressions.

1. The rows of \(A\) represent valid test cases.
2. \(\forall T \subseteq \overline{V I_i}\) such that \(|T| = d, \forall T \in \overline{V I_i} : T \not\supseteq T \Rightarrow (T \in T \Leftrightarrow \rho_A(T) \subseteq \rho_A(T))\)
Below we present logical expressions that represent the above two conditions. By conjuncting all the expressions, we obtain a single logical expression to be checked for satisfiability.

**Condition 1** In A, the n-th row is expressed as a tuple of k variables \((p_1^n, p_2^n, ..., p_k^n)\). As defined in Section 2, a test case is valid iff it satisfies the constraints and the constraints are represented by \(\phi_i\), a Boolean-valued formula over parameters \(F_1, ..., F_k\). We let \(\phi_i|_{p_1^n,p_2^n,...,p_k^n}\) denote \(\phi\) with each \(F_i\) being replaced with \(p_i^n\). Then, the following expression enforces A to only contain valid test cases.

\[
Valid := \bigwedge_{n=1}^{N} \phi|_{p_1^n,p_2^n,...,p_k^n}
\]

**Condition 2** It is important to note that \(T \not\subset T \Rightarrow (T \in T \Leftrightarrow \rho_A(T) \subseteq \rho_A(T))\) is equivalent to:

\[
(T \not\subset T \land T \notin T) \Rightarrow \rho_A(T) \notin \rho_A(T)
\]

because \(T \in T \Rightarrow \rho_A(T) \subseteq \rho_A(T)\) trivially holds. Hence we can focus on the case where \(T \not\subset T\) and \(T \notin T\). The right part of this formula, that is, \(\rho_A(T) \notin \rho_A(T)\) holds iff there is a row in A that covers \(T\) but none of the interactions in \(T\). This condition is represented by a logical expression as follows:

\[
Locating(T, T) := \bigvee_{n=1}^{N} \left( \bigwedge_{j=1}^{t} (p_{x_{y_j}}^n = v_{x_j}) \land \neg \left( \bigvee_{L=1}^{d} \bigwedge_{l=1}^{t} (p_{y_{l,j}}^n = v_{y_{l,j}}) \right) \right)
\]

where \(T = \{(F_{x_1}, v_{x_1}), ..., (F_{x_t}, v_{x_t})\}, ..., \{(F_{x_{d_1}}, v_{x_{d_1}}), ..., (F_{x_{d_t}}, v_{x_{d_t}})\}\) and \(T = \{(F_{x_1}, v_{x_1}), ..., (F_{x_t}, v_{x_t})\}\). 

For given \(T\) and \(T\), \(\rho_A(T) \notin \rho_A(T)\) holds iff \(Locating(T, T)\) is satisfiable.

Let us define \(U^*\) as follows:

\[
U^* := \{(T, T) \mid T \subseteq \forall I_t, |T| = d, T \in \forall I_t, T \not\subset T, T \notin T\}
\]

By ANDing \(Locating(T, T)\) for all \((T, T) \in U^*\), we obtain an expression that represents the second condition.

**The whole expression** The whole expression that will be checked for satisfiability is obtained by conjuncting the expressions defined above as follows:

\[
existCDA := Valid \land \bigwedge_{(T, T) \in U^*} Locating(T, T)
\]

By checking the satisfiability of this expression, whether a \((d, t)\)-CDA of size \(N\) exists or not can be determined. If it is satisfiable, then a CDA of size \(N\) exists. In this case, the satisfying valuation for the \(N \times k\) variables \(p_i^n\) represents all the entries of one such CDA. On the other hand, if the expression is unsatisfiable, then it can be concluded that no \((d, t)\)-CDA of size \(N\) exists.

The satisfiability of the above expression can be checked using Constraint Satisfaction Problem (CSP) solvers, Satisfiability Modulo Theories (SMT) solvers, or Boolean Satisfiability (SAT) solvers with a Boolean encoding of integers.

**4.1.2. Computing \(U^*\)**

In order to construct the above logical expression \(existCDA\), we need to obtain \(U^*\) (see the subscript of the \(\land\) in the expression). Computing \(U^*\) requires \(\forall I_t\). We will show how to compute \(\forall I_t\) later. Here we describe how one can compute \(U^*\) when \(\forall I_t\) is available.

Now consider enumerating all \(T\)-\(T\) pairs such that \(T \in \forall I_t, T \subseteq \forall I_t, |T| = d, T \notin T\), and \(T \not\subset T\). The problem here is how to decide whether or not \(T \subseteq \forall I_t\) masks \(T \in \forall I_t\) when \(T\) and \(T \notin T\) are given. This too is possible by making use of satisfiability solving. We let integer variables \(p_1, p_2, ..., p_k\) to symbolically represent a test case \(\sigma\); that is,

\[
\sigma = (p_1, p_2, ..., p_k)
\]
The domain of \( p_i \) is \([0, 1, \ldots, |S_j| - 1]\). Note that \( S_j \) is the domain of parameter \( F_i \).

By the definition of masking, given such a \( T\)-\( T \) pair, \( T \) does not mask \( T \) iff the following condition holds:

\[
\exists \sigma \in \mathcal{R} : T \subseteq \sigma \land \neg(\exists T' \in \mathcal{T} : T' \subseteq \sigma)
\]

In words, the condition holds if there is a valid test case that covers the interaction \( T \) but does not cover any interactions in the interaction set \( T \). Hence, given \( T \) and \( T \neq T \) holds iff the following expression evaluates to true.

\[
\text{checkUnMasking}(T,T) := \bigwedge_{j=1}^{t} (p_{x_j} = v_{x_j}) \land \neg \left( \bigwedge_{i=1}^{d} \left( \bigwedge_{l=1}^{t} (p_{y_{l,i}} = v_{y_{l,i}}) \right) \right) \land \phi|_{p_1,\ldots,p_k}
\]

where \( T = \{(F_{y_{1,1}}, v_{y_{1,1}}), \ldots, (F_{y_{1,t}}, v_{y_{1,t}})\}, \ldots, \{(F_{y_{d,1}}, v_{y_{d,1}}), \ldots, (F_{y_{d,t}}, v_{y_{d,t}})\} \) and \( T = \{(F_{x_1}, v_{x_1}), \ldots, (F_{x_t}, v_{x_t})\} \).

\( U' \) is obtained by, for every \( T\)-\( T \) pair, checking the satisfiability of \( \text{checkUnMasking}(T,T) \) and keeping the pair in \( U' \) if the expression is satisfiable.

### 4.1.3. The algorithm

The CDA generation algorithm that uses satisfiability solving is shown as Algorithm 1. The algorithm repeatedly solves the problem of finding a \((d,t)\)-CDA while varying the array size \( N \). The array size \( N \) starts with a value large enough to ensure the existence of a CDA and is gradually decreased until no existence of a CDA of size \( N \) is proved. To obtain the initial value of \( N \), the algorithm creates a \((d+t)\)-CCA using an off-the-shelf algorithm (line 1), where the CCA generation algorithm is represented as function \( \text{generateCCA}(M,x) \) which returns an \( x\)-CCA. Our algorithm uses the size of the CCA minus one as the initial \( N \), as any \((d+t)\)-CCA is a \((d,t)\)-CDA. The \((d+t)\)-CCA is also used for computing \( \mathcal{V}_{I_j} \), since all valid \( t\)-way interactions appear in the CCA: The algorithm enumerates all \( t\)-way interactions occurring in the array, thus obtaining \( \mathcal{V}_{I_j} \).

In the algorithm, \( \text{generateCCA}(M, d, t, N, U') \) in line 8 represents a function that produces a \((d,t)\)-CDA of size \( N \) by checking the satisfiability of the expression \( \text{existCDA} \). If the expression is satisfiable, then the SMT solver returns the satisfying valuation, in which case a \((d,t)\)-CDA of size \( N \) is obtained, since the valuation represents the \((d,t)\)-CDA. The size \( N \) is decreased by one and the same process is repeated. If the result of satisfiability check is UNSAT

---

**Algorithm 1: The satisfiability-based algorithm**

| Line | Description |
|------|-------------|
| 1    | \( S \leftarrow \text{generateCCA}(M, d + t) \) get all valid \( t\)-way interactions from the \((d+t)\)-CCA |
| 2    | \( \forall I_j \leftarrow \text{getAllInteractions}(S, t) \) get all non-masking pairs \( U' \) of interaction sets and interactions |
| 3    | \( U' \leftarrow \text{getU}(\forall I_j, d, t) \) get the initial size for the CDA to be generated |
| 4    | \( N \leftarrow \text{The size of } S - 1 \) |
| 5    | \( \text{nextA} \leftarrow S \) |
| 6    | **do** |
| 7    | \( A \leftarrow \text{nextA} \) |
| 8    | \( \text{nextA} \leftarrow \text{generateCDA}(M, d, t, N, U') \) |
| 9    | \( N \leftarrow N - 1 \) |
| 10   | **while** \( \text{nextA} \neq \emptyset \) |
| 11   | **return** \( A \) |
Algorithm 2: The two-step heuristic generation algorithm

```
// Input: SUT M = (F, S, φ); integers d, t
// Output: (d, t)-CDA A
// construct a (d + t)-CCA for the input SUT
S ← generateCCA(M, d + t)
// get all t-way interactions from the (d + t)-CCA
VI, ← getAllInteractions(S, t)
// Rows[T] = ρ_S(T) for T ∈ VI,
Rows[] ← mapInteractionToRows(VI, S)
// DiffRows[T] = ρ_S(T) - ρ_T(T) for T ⊆ VI, |T| = d
DiffRows[][] ← getDiffRows(VI, S, d)
A ← S
while A ≠ ∅ do
    σ ← getRandomTestcase(S)
    S ← S - {σ}
    A ← A - {σ}
    DiffRows'[][] ← update(DiffRows[][], σ)
    if ∃T, T : DiffRows'[T] ≠ ∅ and DiffRows'[T][T] = ∅ then
        // the test case σ is unremovable
        S ← S + {σ}
    else
        // the test case σ is removable
        DiffRows[][] ← DiffRows'[][]
return A
```

(unsatisfiable), no CDA of size N exists (denoted as ⊥ in the algorithm). Then the algorithm returns the CDA of size N + 1 and stops its execution.

One might think that binary search could work better to vary N than the linear search adopted by the algorithm. In fact, this is not the case because showing unsatisfiability, that is, the nonexistence of a CDA, usually takes much longer time than showing satisfiability, that is, the existence of a CDA. The linear search delays solving an unsatisfiable expression until all possible sizes are checked, avoiding getting trapped in a long computation required for the unsatisfiable problem instance.

The size of the expression existCDA increases polynomially in k when t, d, |S|, and N are fixed. The expression can be expressed as a Boolean formula with a polynomial size increase, as |S| is fixed. The Boolean satisfiability problem (SAT) is NP-complete in general and there is no reason that SAT can be solved in polynomial-time for this particular case. Hence the time complexity of the algorithm is likely to be exponential.

4.2. The two-step heuristic algorithm

In this subsection, we propose a two-step heuristic algorithm for the generation of (d, t)-CDAs which aims to generate (d, t)-CDAs that are not optimal but fairly small in reasonable time.

Theorem 6 shows that a (d + t)-CCA is already a (d, t)-CDA. Based on the theorem, we devise a two-step heuristic algorithm (Algorithm 2). The algorithm generates a (d + t)-CCA first. Then it repeatedly chooses a test case in it at random and checks whether it is removable. Here we say that a test case is removable from an array if a new array with the test case being removed would still be a (d, t)-CDA. If the test case is removable, then it is removed from the current array. Otherwise, a new test case is chosen and the check is performed again. This process is repeated until no test case is removable anymore.

The algorithm works in detail as follows. In line 1 the algorithm generates a (d + t)-CCA S. At this point, S is already a (d, t)-CDA but contains many redundant test cases. Then the algorithm collects all valid t-way interactions and maps each interaction T to its covering test cases ρ_S(T) in S (line 2). The map obtained here, denoted by Rows[], is used to compute another map, DiffRows[][], that associates each pair of an interaction set T and a valid interaction.
Table 2
Benchmark Information

| ID | SUT                  | |F| |φ| |VI| |I\|I| |T > T| |
|---|---------------------|---|---|---|---|---|---|---|---|---|
| 1 | car                 | 9 | 15 | 102 | 42 | 1,487 |
| 2 | graph_product_line  | 20 | 45 | 499 | 261 | 37,212 |
| 3 | real_fm             | 14 | 23 | 275 | 89 | 5,368 |
| 4 | aircraft_fm         | 13 | 19 | 239 | 73 | 2,647 |
| 5 | connector_fm        | 20 | 37 | 537 | 223 | 49,038 |
| 6 | movies_app_fm       | 13 | 23 | 211 | 101 | 4,968 |
| 7 | stack_fm            | 17 | 28 | 465 | 79 | 6,399 |
| 8 | banking1            | 5 | 112 | 102 | 0 | 0 |
| 9 | banking2            | 15 | 3 | 473 | 3 | 208 |
| 10 | comm_protocol       | 11 | 128 | 285 | 35 | 2,177 |
| 11 | concurrency         | 5 | 7 | 36 | 4 | 130 |
| 12 | healthcare1         | 10 | 21 | 361 | 8 | 512 |
| 13 | healthcare2         | 12 | 25 | 466 | 1 | 124 |
| 14 | healthcare3         | 29 | 31 | 3,092 | 59 | 8,700 |
| 15 | healthcare4         | 35 | 22 | 5,707 | 38 | 3,359 |
| 16 | insurance           | 14 | 0 | 4,573 | 0 | 0 |
| 17 | network_mgmt        | 9 | 20 | 1,228 | 20 | 189 |
| 18 | processor_comm1     | 15 | 13 | 1,058 | 13 | 1,510 |
| 19 | processor_comm2     | 25 | 125 | 2,525 | 854 | 35,156 |
| 20 | services            | 13 | 388 | 1,819 | 16 | 1,088 |
| 21 | storage1            | 4 | 95 | 53 | 18 | 112 |
| 22 | storage2            | 5 | 0 | 126 | 0 | 0 |
| 23 | storage3            | 15 | 48 | 1,020 | 120 | 3,400 |
| 24 | storage4            | 20 | 24 | 3,491 | 24 | 0 |
| 25 | storage5            | 23 | 151 | 5,342 | 246 | 10,095 |
| 26 | system_mgmt         | 10 | 17 | 310 | 14 | 825 |
| 27 | telecom             | 10 | 21 | 440 | 11 | 151 |

$T$ with $\rho_S(T) - \rho_S(T) = \emptyset$ iff $\rho_S(T) \subseteq \rho_S(T)$. Since $S$ is a CDA, $DiffRows[T][T] = \emptyset$ if $T > T$; $DiffRows[T][T] \neq \emptyset$ otherwise.

Then the algorithm repeatedly chooses a test case at random and checks whether it is removable or not. To perform the check, the algorithm constructs a new interaction-to-row map $DiffRows[[]]$ that would hold after the test case was removed (line 9). This can be done by simply removing $\sigma$ from all $DiffRows[T][T]$. Subsequently, the algorithm compares the two maps (line 10). If $DiffRows[T][T] \neq \emptyset$ but $DiffRows[\sigma[T][T] = \emptyset$, then $\rho_S(T) \subseteq \rho_S(T)$ and thus $S$ is no longer a CDA. In this case, the algorithm reserves the test case (line 12). Otherwise, it deletes the test case and accordingly updates $DiffRows[\sigma[T][T]$ (line 14). When all test cases in the CCA are checked, the algorithm will terminate, yielding the resulting $S$.

Let $s = \max_{1 \leq i \leq k} |S_i|$. Outside the while loop, line 4 has the highest time complexity. It is $O((s't')^d s't'n)$, since $|VI_i| \leq s't'$, $|\rho_S|$ $\leq n$. Inside the while loop, line 10 and line 11 has the highest complexity $O((s't')^d s't'n)$ for the same reason. And let $n$ be the size of the initial CCA. As a result, the algorithm’s time complexity is $O((s't')^d s't'n^2)$. When $s, t$, and $d$ are fixed, the complexity is polynomial in $k$ and $n$.

5. Experiments

In this section we show the results of experiments to evaluate the two proposed algorithms presented in the previous section. We focus on generation of $(1, 2)$-CDAs ($d = 1, t = 2$) for the following reasons. First, by nature of CDAs no interactions can be erroneously identified as non-faulty even when more than $d$ interactions are faulty; thus it is natural to set a small value to $d$ in practice. Second, the most common form of CIT targets two-way interactions (this form of CIT is called pair-wise testing.)
### Table 3
Experimental results. Numbers with * indicate that the algorithm did not terminate within the time limit, in which case the CDAs obtained are not necessarily minimum.

| ID | Time (second) | Size |
|----|--------------|------|
|    | SMT          |      | Two-step       |      |
|    | avg. | max. | min. | avg. | avg. | max. | min. | avg. | avg. |
| 1  | 2.58 | 0.12 | 0.07 | 0.08 |      |      |
| 2  | 0.18 | 0.13 | 0.15 |      |      |      |
| 3  | 0.16 | 0.11 | 0.13 |      |      |      |
| 4  | 0.10 | 0.09 | 0.10 |      |      |      |
| 5  | 0.17 | 0.12 | 0.14 |      |      |      |
| 6  | 0.12 | 0.09 | 0.10 |      |      |      |
| 7  | 0.18 | 0.13 | 0.15 |      |      |      |
| 8  | 0.15 | 0.10 | 0.13 |      |      |      |
| 9  | 0.10 | 0.07 | 0.08 |      |      |      |
| 10 | 0.17 | 0.12 | 0.14 |      |      |      |
| 11 | 0.20 | 0.16 | 0.18 |      |      |      |
| 12 | 3.14 | 2.79 | 2.95 |      |      |      |
| 13 | 19.87 | 18.73 | 19.23 |      |      |      |
| 14 | 145.14 | 130.48 | 134.31 |      |      |      |
| 15 | 1.56 | 1.48 | 1.52 |      |      |      |
| 16 | 0.67 | 0.59 | 0.61 |      |      |      |
| 17 | 3.16 | 2.94 | 3.03 |      |      |      |
| 18 | 4.82 | 4.76 | 4.79 |      |      |      |
| 19 | 2.53 | 0.12 | 0.10 | 0.10 |      |      |
| 20 | 0.08 | 0.07 | 0.07 |      |      |      |
| 21 | 0.64 | 0.58 | 0.61 |      |      |      |
| 22 | 16.28 | 15.61 | 15.88 |      |      |      |
| 23 | 130.25 | 120.02 | 124.86 |      |      |      |
| 24 | 0.12 | 0.09 | 0.11 |      |      |      |
| 25 | 0.19 | 0.14 | 0.16 |      |      |      |

### 5.1. Experiment settings
We wrote C++ programs that implement the two algorithms. Our implementation [10] of the IPOG algorithm [6] was used as a CCA generator for both algorithms, while the Z3 solver (version 4.8.1) was used in the satisfiability-based algorithms. We performed experiments with a total of 27 benchmark instances, numbered from 1 to 27. Benchmarks No. 1 to 7 are taken from [11] which are provided as part of the CitLab tool. Benchmarks No. 8 to 27 can be found in [12]. The detailed information of these benchmark instances is shown in Table 2. In Table 2, the columns labeled with $|\mathcal{F}|$ and $|\phi|$ show the number of parameters and the number of constraints ($\phi$ is the conjunction of the constraints). Columns $|\mathcal{V}_{I_2}|$ and $|I_2 \setminus \mathcal{V}_{I_2}|$ show respectively the number of valid interactions and the number of invalid interactions. The last column labeled $|\mathcal{F} \supset \mathcal{T}|$ shows the number of pairs of an interaction set $\mathcal{F}$ and an interaction $\mathcal{T}$ such that $\mathcal{F}$ masks $\mathcal{T}$. For instance, the first line in the table shows that the benchmark car has 9 parameters with 15 constraints. In the test space there are 102 valid interactions and 42 invalid interactions. Among the valid interactions, there are 1,487 pairs of an interaction set and an interaction such that the interaction set masks the interaction. All experiments were conducted on a machine with Intel Core i7-8700 CPU, 64 GB memory and Ubuntu 18.04 LTS OS. For each benchmark instance, the two generation algorithms were executed five times. The timeout period for each run was set to 1800 seconds.

### 5.2. Experimental results
The results of the experiments are summarized in Table 3. The leftmost column shows the benchmark IDs. The rest of the table is divided into two parts representing the results of generation time and the results of sizes of the generated CDAs. Both parts have two sections describing the experiment results of the two proposed algorithms respectively.
For each problem instance, the average value is reported for the satisfiability-based algorithm as it is deterministic, while the maximum, minimum, and average values are reported for the two-step heuristic generation algorithm.

The numbers with asterisk (*) in the satisfiability-based algorithm’s columns show that the generation did not terminate within the time limit. Because the algorithm repeatedly generates CDAs with sizes varying until the minimum one is found, CDAs that are not optimal are constructed during the course of execution. The values with asterisk (*) correspond to the smallest (not necessarily optimal) CDAs that were obtained within the time limit. For example, for benchmark No. 3, the algorithm took 806.21 seconds to generate a CDA of size 28. However, when it was trying to generate a CDA of size 27, the run of the algorithm exceeded the 1800 second time limit. There are also some benchmark instances that the algorithm did not find even one CDA within the time limit. We use the symbol “…” to indicate such a case. To compare the average consumed time of the two algorithms, the better results (i.e., the shorter time) are denoted in bold font. The smaller average sizes of generated CDAs are also denoted in bold font.

The satisfiability-based algorithm completed the generation process for three instances, namely, No. 1, 6, and 11. The CDAs obtained for these instances are all optimal. The algorithm was able to find small CDAs for some remaining instances (though it timed out), whereas it failed to find even a single CDA for others. In contrast, the two-step algorithms successfully generated CDAs for all benchmark instances. In addition, the execution time of the satisfiability-based algorithm was always much longer than the other algorithm, sometimes three orders of magnitude longer. There are two main reasons why the satisfiability-based algorithm is so slow. One reason is that the algorithm generates multiple CDAs in a single run. As stated in Section 4, it generates \((d, t)\)-CDAs with sizes varying from the size of a \((d + t)\)-CCA. The CCA’s size simply serves as the upper bound on the minimum CDA size: As it is not tight bound in general, to obtain an optimal \((d, t)\)-CDA, the satisfiability solver is executed multiple times. The other reason, which is more obvious, is that satisfiability check may be time-consuming. The time required for the check becomes very long especially when the algorithm tries to find a CDA of minimum size minus 1, in which case the answer of the check is UNSAT (unsatisfiable). In the field of satisfiability, it is well known that UNSAT instances are usually more difficult than SAT instances.

The satisfiability-based algorithm is deterministic. As stated above, the CCA size affects the algorithm’s execution time and, if timeout occurs, the resulting CDA size. In contrast, the two-step heuristic algorithm is inherently nondeterministic: it generates different CDAs for different runs. The algorithm decreases the array size by repeatedly removing from the current array a test case selected at random (line 7, Algorithm 2). A test case can be removed only if the array remains to be a CDA after its removal; thus which test case is removed depends strongly on earlier selections. Hence, different orders in which test cases are deleted lead to different CDAs.

Another observation is that the two-step heuristic algorithm generated smaller CDAs than the satisfiability-based algorithm for No. 7. For the case, the satisfiability-based algorithm ran out of time before searching for minimum or near minimum CDAs. In view of these, we conclude that the two-step heuristic algorithm has balanced capabilities with respect to running time and CDA sizes it generates.

6. Threats to Validity

The experimental results about the two CDA generation algorithms showed that the scalability of the satisfiability-based algorithm is substantially limited, especially when comparing to the heuristic algorithm. This conclusion heavily relies on the performance of the satisfiability solver used in the implementation. Although Z3, which we adopted in our implementation, is one of the best known and fastest SMT solvers, solvers of SMT or similar problems, such as SAT, have seen constant progress. Hence the difference between the two algorithms might narrow in near future.

The problem instances used in the experiments are well-known and have been used in many other studies; However, they do not necessarily capture the characteristics of all real-world problems. Although we believe that the algorithms’ qualitative properties observed in the experiments are likely to hold in general, there can be new problems for which they do not hold.

7. Related Work

CIT has been widely used for many years. In the practice of CIT, constraint handling has always been a vital issue. Surveys about constraint handling for CIT include [13, 14, 15].

DAs, as well as LAs, were first introduced by Colbourn and McClary in [3]. They analyzed the mathematical properties of these arrays. As [3], most of the studies on DAs and LAs focus on their mathematical aspects [16,
Constrained Detecting Arrays: Mathematical Structures for Fault Identification in Combinatorial Interaction Testing

17, 18, 19, 20, 21]. The application to screening experiments for TCP throughput in a mobile wireless network was reported in [22, 23]. Other types of arrays that are intended for fault location include Error Locating Arrays [24] and Consecutive Detecting Arrays [25].

The concept of CLAs was first introduced in [26]. Later a computational construction algorithm was proposed in [27]. In [7] the results of applying CLAs to fault localization for real-world programs were reported.

The present paper extends our earlier works: [28] and [29]. In [28] we introduced $(d, t)$-CDAs for the first time, together with a construction algorithm using an SMT-solver. The present paper introduces the other variants of CDAs, clarifies their relations, and refines the algorithm. The two-step heuristic algorithm was first proposed in [29]. In the current paper, we improved the implementation of the algorithm and conducted a new set of experiments to compare the two different algorithms with the new implementations.

There are many other approaches to faulty interaction localization without using CDAs or other related arrays. One of such approaches is the use of adaptive testing [30, 31, 32, 33, 34, 35, 36, 37]. In adaptive testing, when a failure is encountered, new test cases are adaptively generated and executed to narrow down possible causes. On the other hand, testing using CDAs is nonadaptive in the sense that test outcomes do not alter future test plans. A clear benefit of using nonadaptive testing is the execution of test suites, which is often the most time-consuming part of the whole testing process, can be parallelized.

CDAs and other arrays of similar kinds are intended to provide sufficient test outcomes to uniquely identify faulty interactions. On the other hand, some studies attempt to infer faulty interactions from insufficient information with, for example, machine learning. The studies in this line include [38, 39, 40]. For other approaches to identification of faulty interactions, readers are referred to a recent survey [41].

8. Conclusion

In this paper, we introduced the notion of Constrained Detecting Arrays (CDAs), which incorporates constraints among test parameters into Detecting Arrays (DAs). CDAs generalize DAs so that localization of faulty interactions can be performed for systems with constraints. We proved several properties of CDAs as well as those that relate CDAs with other array structures, such as Constrained Covering Arrays (CCAs) and Constrained Locating Arrays (CLAs). We then proposed two generation algorithms. The first algorithm generates optimal CDAs using an off-the-shelf satisfiability solver. The second algorithm is heuristic and generates near-optimal CDAs in a reasonable time. The experimental results of both algorithms indicated that the heuristic algorithm can scale to problems of practical sizes.

There are several possible directions for future work. One direction is to apply CDAs to the testing of real-world programs to identify faulty interactions. The development of new algorithms for CDA construction also deserves further study. We believe that both meta-heuristic search and greedy heuristics may be promising because they have proved to be useful for the construction of CCAs. A recent study attempts to provide a systematic framework to compare CCA generators [42]. Applying such a framework to compare different CDA construction algorithms is also of interest.

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H. Jin, C. Shi and T. Tsuchiya: Preprint submitted to Elsevier
