COUVERTURE

Technical Report on OBC/MCDC properties

Abstract

This document gathers results established or formalized by the COUVERTURE project team about relationships between specific coverage criteria. We focus in particular on how *Object Branch Coverage* (OBC) relates to the *Modified Condition/Decision Coverage* (MC/DC) criterion.

We provide two broad categories of results: formal proofs of important properties over a model of the two criteria, and a machine-automated verification of some of these properties for concrete subsets of the model expressed in Alloy. These results constitute the grounds on which our project coverage analysis framework operates to infer source coverage results from object coverage information out of an instrumented execution environment.

1 Common definitions

1.1 Decisions, conditions

We are considering decisions that are short circuit boolean expressions, i.e. expressions consisting in elementary boolean conditions combined together using only the `and then`, `or else` and `not` operators.

In order to compare *Object Branch Coverage* to *Modified Condition/Decision Coverage*, it will be shown that the appropriate abstractions are Reduced Ordered Binary Decision Diagram. For each decision (ROBDD), we will provide an algorithm that constructs the associated ROBDD, whose nodes are conditions. (RO)BDDs have an entry point, and two or more exit edges labeled True and False.

In the remainder of this document, unless otherwise indicated, all references to BDDs denote reduced ordered BDDs. For a set S, #S will refer to this set's cardinal. For a decision D, cond(D) will be the set of its conditions.

1.2 Coverage metrics

Coverage assessment is accomplished by exercising some piece of object code in a variety of test cases, recording data along the way, and then determining whether the successive executions of the object code satisfy a given criterion of exhaustiveness. This section will define the main coverage criterion that we will consider.

The minimum number of distinct executions required to achieve a specific criterion is an important aspect in the evaluation of any coverage assessment methodology. Here we establish some properties that give a hard limit on test set size for various coverage metrics.

1.2.1 MC/DC

Formal definition: The formal definition of MC/DC that it used here is the one given in [Chi01], using a graph-coloring algorithm to determine whether two given evaluations show the independent influence of a condition. Amongst the various formal definitions that you may find in the litterature, this one has the advantage to apply also to short-circuit operators, which is mandatory in the context of COUVERTURE.

The following will not re-define what has already been detailed in [Chi01]; we will just give an example on how it works and will invite the interested reader to consult the

original document. A formal definition may also be found in the Alloy model, in file `decision_coverage.als`.

This formal definition of MC/DC uses a simple graph-coloring algorithm to determine whether two given evaluations show the independent influence of a condition. It first colors each node of the decision syntax tree with both evaluations; for example, for a decision (A and then B) or else C, the syntax tree can be seen on figure 1(a).

(a)
(A and then B) or else C
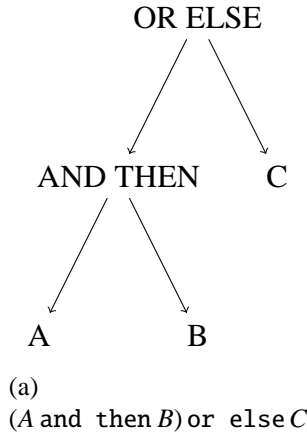
For each evaluation, each node of the tree (atomic condition or root of sub-decision) is colored with the value that it takes in this evaluation: True (T), False (F), Not_Evaluated (X). For the evaluation $e1 = (A = True, B = True, C = Not\_Evaluated)$, we have the coloring given on figure 1(b); and for $e2 = (A = False, B = Not\_Evaluated, C = False)$, we have the coloring given on figure 1(c).
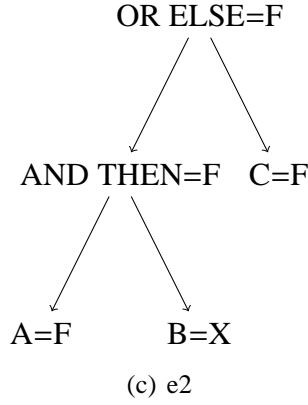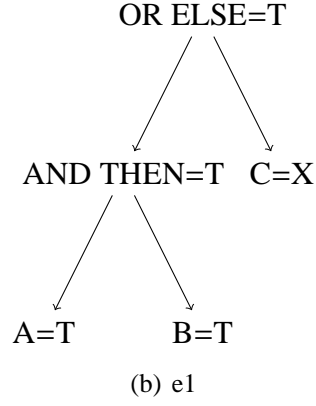
Both graphs are then combined into an influence tree by xor'ing each node. An extension of xor to three-value boolean algebra is used: anything xor Not_Evaluated gives False. In our case, the result is showed on figure 1(d).

The influence set of these two evaluations for the given decision is the set of conditions that have a path of "True nodes" to the root. So here, it is the singleton {A}.

The two classical variants of MC/DC are then defined as follow:

DEFINITION 1 *Given a decision D, a pair of truth vectors satisfies Masking MC/DC for a condition c in cond(D) if and only if its influence set is the singleton {c}.*

A test set satisfies Masking MC/DC for a decision D if, and only if, for each condition in cond(D), there exists a pair of tests in the test set that satisfies Masking MC/DC.

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```
                    OR ELSE=T
                   /          \
                  /            \
          AND THEN=T        C=X
            /      \
           /        \
        A=T          B=T
```

(b) e1

```
                    OR ELSE=F
                   /          \
                  /            \
          AND THEN=F        C=F
            /      \
           /        \
        A=F          B=X
```

(c) e2

DEFINITION 2 *Given a decision D, a pair of truth vectors satisfies Unique Cause MC/DC for a condition c in cond(D) if and only if its influence set is the singleton {c} and if c is the only condition that is colored with True in the influence tree.*

A test set satisfies Unique Cause MC/DC for a decision D if, and only if, for each condition in cond(D), there exists a pair of tests in the test set that satisfies Unique Cause MC/DC.

In our example, our pair of truth vectors satisfies both Unique Cause and Masking MC/DC for condition A.

Properties: For Unique Cause, the following property holds:

THEOREM 1 *Unique MC/DC on a decision with n independent conditions is achieved with a test set of exactly $n+1$ tests, and cannot be achieved in fewer tests.*

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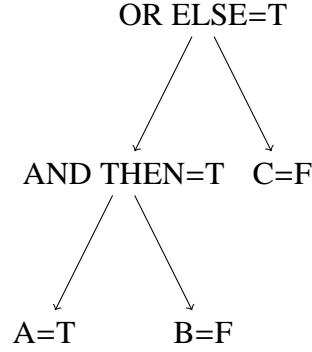
(d) Influence tree

The existence of the test set is proved by induction. For one condition, MC/DC is achieved with two tests, one setting it True and the other False.

Now assume that the property holds for all $n \leqslant N$, and consider a decision with $N + 1$ conditions. It is of the form $D = D_l \star D_r$ where $\star$ is either and then or or else. There is also possibly a negation, which is omitted here since it has no impact on MC/DC. For the remainder of this proof we assume the operator is and then; the same reasoning applies similarly for the case of or else.

D_l and D_r are decisions with respectively n_l and n_r conditions (both at most N), and $n_l + n_r = N + 1$. From the induction hypothesis we have two test vector sets $T_l = \{v_l(0)..v_l(n_l)\}$ and $T_r = \{v_r(0)..v_r(n_r)\}$ that satisfy MC/DC for D_l and D_r respectively, and we can arbitrarily choose the indices so that D_l is True for $v_l(0)$ and D_r is True for $v_r(0)$.

We can now create a combined test set for the complete decision as follows.

$$(\forall j \in [0, n_l]) \, v(j) = (v_l(j) \cdot v_r(0))$$

$$(\forall j \in [1, n_r]) \, v(n_l + j) = (v_l(0) \cdot v_r(j))$$

We have thus created a set $T = \{v(0)..v(n_l + n_r)\}$ of $N + 2$ tests. It is immediate that the elements with indices 0 to n_l give for D the same outcome as the corresponding elements of T_l for D_l, and they all have identical values for the conditions coming from D_r, so they show independent influence of all conditions coming from D_l. Similarly the vector set $\{v(0), v(n_l + 1)..v(n_l + n_r)\}$ shows independent influence of those conditions coming from D_r, so the new test set T satisfies MC/DC for D and thus the induction property holds at $N + 1$ as well, since T has $n_l + n_r + 1 = N + 2$ elements.

The proof that this is the minimal test set size is given in [Chi01]. As for Masking MC/DC, we have the following property:

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Theorem 2 *Masking MC/DC on a decision with n independent conditions requires a minimum of $RUTW(2 * SQRT(n))$ tests, where RUTW stands for round up to whole.*

The proof of this property is given in [Chi01] as well.

The formal definitions of Masking MC/DC also allows to identify interesting properties when only short circuit operators are used. The following theorem tells us that, with short circuit operators, determining Masking MC/DC is as simple as for Unique Cause, and that no reasoning on the logical structure of the decision is needed:

Theorem 3 *If a decision Dec contains only short-circuit operators, Masking MC/DC is reached for a condition C if and only if there exists a pair of evaluation (e1, e2) such that Dec[e1] = not Dec[e2], and C is the rightmost condition that is evaluated to True and False in e1 and e2.*

The order here is the order of conditions in the decision expression; e.g. in (A `and then` B) `or else` C, the order is A, B, C.

This theorem can be re-formulated in terms of influence trees, as in the following lemma:

Lemma 1.2.1 *If the root node of an influence tree is colored with True, and if it contains only short-circuit operators, then its influence set contains only the rightmost True-colored condition.*

This lemma can be proven by structural induction:

- if $D ::= C$ (simple condition decision):
  the root of D's influence tree is C, so if it is colored with True then it follows that the influence set is { C }.

- if $D ::= $ `not` $D1$, supposing that the property holds for D1:
  The root of D's influence tree is colored with True if and only if the root of D1's influence tree is colored with True as well: The former is evaluated to two different values (say: True, then False) if and only if the latter is evaluated to two different values (False, then True). Therefore, D1's influence set contains only the rightmost True-colored condition in D1's influence tree, which also is the rightmost True-colored condition in D's influence tree. This condition is the only one to have a path of True-colored nodes to root in D1's influence tree, by definition of the influence set; as a consequence, it is also the only condition to have such path to root in D's influence tree. So the rightmost True-colored condition in D's influence tree is the only condition in its influence set.

- if $D ::= DL \star DR$, $\star$ being a short-circuit operator, and supposing that the property holds for DL and DR:

  Let us define the boolean constant SC as follow:
  If $\star$ is `and then`, let SC be False;
  if $\star$ is `or else`, let SC be True.

  Then there are three possible evaluations for DL, DR, D:

Eval name	DL	DR	D
e1	not SC	not SC	not SC
e2	not SC	SC	SC
e3	SC	x	SC

  If the root of D's influence tree is colored with True, then there are only two possible pairs of evaluations:

  - (e1, e2): in this case, DR is colored with True and DL is colored with False; using the induction hypothesis, we know that DR's influence tree will have only one condition with a True-colored path to DR's root, and consequently to D's root in D's influence tree : that will be the rightmost True-colored condition in DR, and obviously in D. No conditions in DL have a True-colored path to D's root, since DR's root node is colored with False ; so the influence set contains only the rightmost True-colored condition.
  - (e1, e3): in e3, DR is not evaluated, which means that none of its nodes are evaluated. This means that all of them are colored with False in DR's influence tree, and that no conditions of DR are colored with True in D's influence tree. Thus, the rightmost True-colored condition of D is in DL. By induction hypothesis, it is the only element of DL's influence set; it is therefore the only condition to have a True-colored path to DL's root node, and to D's root node. So this is the only element in D's influence set.

The theorem has now been demonstrated in all cases.

1.2.2 OBC

Applicants for DO-178 certification have sometimes proposed the use of object code coverage instead of source code coverage as a metric to satisfy the objectives of DO-178. The proposed approach involves measuring either instruction coverage or branch

coverage. Object instruction coverage (OIC) requires assessing whether all object instructions are executed at least once; object branch coverage (OBC) in addition requires that all conditional branches be exercised for both directions (branch and fall through). The use of object code coverage has also been proposed as a way to cope with untraceable object code. Section 6.4.4.2 of DO-178 indeed states: *The structural coverage analysis may be performed on the Source Code, unless the software level is A and the compiler generates object code that is not directly traceable to Source Code statements. Then, additional verification should be performed on the object code to establish the correctness of such generated code sequences.*

Untraceable object code is compiler generated machine code that impacts the execution control flow in a way not directly visible from source code. An example is the Ada `mod` operator, which requires an implicit conditional branch to distinguish between positive and negative moduli. Achieving a certain source coverage level is not indicative of coverage of untraceable code: even a comprehensive testing campaign (from a source coverage point of view) may not assure all object code is executed. The untraceable object code is nevertheless present in the application and may lead to unintended behaviour not verified during the testing process.

Measuring object code coverage may be a way to ensure even untraceable code is executed during the requirement-driven testing campaign. However, CAST paper 12 [CAS02] suggests the use of a *traceability study* to satisfy the additional verification activities on untraceable object code for level A software: a traceability study provides evidence that, in a given context (coding standard, compiler, compilation switches), the compiler either generates traceable code or the untraceable code is correct — i.e. it correctly implements the requirements expressed by the specifications of the chosen programming language.

The issue raised by section 6.4.4.2 of DO-178, and in general the equivalence of source and object coverage, is considered in both FAQ 42 of DO-248B [RTC01] (*Can structural coverage be demonstrated by analyzing the object code instead of the source code?*) and CAST paper 17 (*Structural Coverage of Object Code*) issued by the Federal Aviation Administration [CAS03].

Both documents assert that object code coverage can substitute for source code coverage *as long as analysis can be provided which demonstrates that the coverage analysis conducted at the object code will be equivalent to the same coverage analysis at the source code level.* In the following section, we provide a precise analysis of the conditions under which the OBC and MC/DC properties imply each other, for a given set of test cases. It should be noted that such an equivalence applies only in the context of a given code generator and coding guidelines: different code generation algorithms may produce object code whose OBC status is different for the same set of inputs achieving a given source coverage objective.

1.2.3 BDDBC

As stated earlier, Binary Decision Diagrams are good abstractions to compare the two criteria that we are considering; this section will justify this choice. It will also introduce a new structural coverage criterion, named *Binary Decision Diagram Branch Coverage*, that relates naturally to OBC.

A Binary Decision Diagrams is a standard data structure that can be used to represent a decision; a formal definition, along with a set of associated algorithms and properties, can be found in [Bry86]. In a nutshell, BDDs can be described as a directed acyclic graph; each node is mapped to one condition, and leaf nodes are outcome True or False; two edges labeled True and False link each condition to its children nodes ; those children nodes are either a decision outcome or the next condition to evaluate, and are executed if the father condition takes the value labeled on the corresponding edge. Let us illustrate that on our decision (*A* `and then` *B*) `or else` *C*; one of its BDD is shown on figure 1(e).

One key property of BDDs is that they have a reduced form (no isomorphic subgraph, no trivial node), and that this reduced form is canonical (unique) for a particular decision. For instance, figure 1(e) is a reduced ordered BDD.

Now we can define BDDBC as follow:

DEFINITION 3 *Given a decision D, a set of truth vectors satisfies BDDBC if the corresponding set of paths through its reduced ordered BDD covers all edges of the BDD.*

One can now understand how this relates to OBC. For each pair of edges that exits from a condition node, the compiler will generate a branch that may or may not be taken, depending on the value of the corresponding condition; both possibilities (taking it, not taking it) maps to one edge of the pair. If OBC is reached, then both edges should be taken, and consequently all edges of the BDD. In other words: OBC implies BDDBC.

This assumes that the compiler will generate one branch per condition for each decision. In the context of COUVERTURE, an compilation option -fpreserve-control-flow is provided that enforces this property; this allows a straight-forward comparizon between OBC and BDDBC. In the following, we specifically assume when discussing object branch coverage of the object code, that we can instead reason, unless indicated explicitly, on BDD branch coverage of the corresponding (RO)BDD.

2 Characterization of cases of BDDBC — MC/DC equivalence

This section discusses the distinction between expressions for which BDDBC of the associated ROBDD implies MC/DC, and expressions for which no such implication holds.

2.1 Some cases of non equivalence between OBC and MC/DC

First let us have a look at how MC/DC and object coverage relate to each other on some simple cases. Cases of non-equivalence for decisions with up to 5 conditions have been studied in [FAA07]: non-equivalence cases have been shown to occur in decisions with three or more conditions, and an illustration is provided with (A and then B) or else C, where A, B and C are three independent conditions. A representation of this decision's BDD is depicted on figure 1(e).
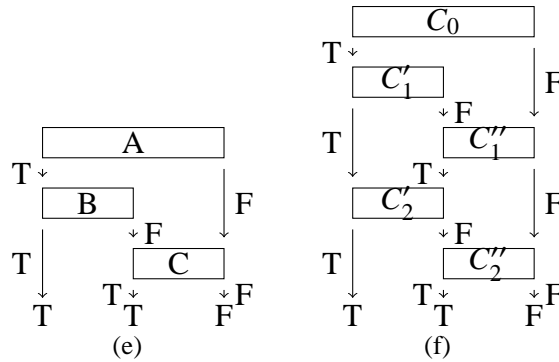

Figure 1: Example decision BDDs

From this representation, we can see that a set of three evaluations can achieve branch coverage of the whole BDD, corresponding to the three vertical paths in figure 1(e). These evaluations are:

A	B	C	(A and then B) or else C
T	T	x	T
T	F	T	T
F	x	F	F

where "x" means not evaluated and thus can be indifferently True or False. Now, indeed, even though all the BDD edges are covered, MC/DC is not met. In particular, the independent effect of conditions B on the decision is not shown in the case of Masking MC/DC, and the independent effect of both B and C are not shown in the case of Unique Cause. Since $n+1$ tests are needed to cover a decision with n conditions with respect to Unique Cause MC/DC, 3 evaluations cannot cover a three-condition decision. Masking MC/DC requires a minimum of 4 evaluations as well in this particular case.

It turns out that this particular case can be generalized in a quite spectacular counterexample: there exists classes of decisions with an arbitrary high number of conditions that can be branch covered by just three evaluations; the previous example was one element of this class with 3 conditions.

Consider the following set $\{D_n\}_{n \in \mathbb{N}}$ of decisions:

- let D_0 be a simple condition decision; by convention, we will call C_0 its condition;

- let us define D_n, for any $n > 0$, as follows:
  $D_n = (D_{n-1}$ and then $C'_n)$ or else C''_n
  C'_n and C''_n being independent from each other and from any condition in D_{n-1}.

In other words:

- $D_0 = C_0$

- $D_1 = (C_0$ and then $C'_1)$ or else C''_1

- $D_2 = (((C_0$ and then $C'_1)$ or else $C''_1)$
  and then $C'_2)$ or else C''_2

- $D_3 = (((((C_0$ and then $C'_1)$ or else $C''_1)$
  and then $C'_2)$ or else $C''_2)$
  and then $C'_3)$ or else C''_3

- ...

Figure 1(f) shows the BDD for D_2, where it is visible that all the edges can be covered by three evaluation paths which only demonstrate the independent effect of C_0:

C_0	C_1'	C_1''	C_2'	C_2''	D_2
T	T	x	T	x	T
T	F	T	F	T	T
F	x	F	x	F	F

We can thus build a decision D_n with an arbitrary number of conditions, that can be BDD branch covered by just three evaluation paths. As Unique Cause MC/DC can only be achieved with a minimal number of $n + 1$ evaluations, and Masking MC/DC with a minimal number of $RUTW(2 * SQRT(N))$ tests, this is a striking case where BDD branch coverage (and consequently OBC) is far from being equivalent to MC/DC.

We can now adopt a more general perspective and characterize more precisely the difference between these two criteria.

2.2   Construction of the ROBDD

The BDD we associate with a decision is constructed using the following recursive procedure:

Build_BDD.Condition The BDD for a decision consisting in a single condition C has the node "test C" as its entry point, the label True is assigned to the branch corresponding to "C is True", and the label False is assigned to the branch corresponding to "C is False".

Build_BDD.NOT The BDD for `not(D)` is the BDD for D where the labels of the exit edges have been swapped.

Build_BDD.Short_Circuit_Operator This rule defines how the BDD for $(DL) \star (DR)$ is constructed for any short-circuit operator $\star$.

If $\star$ is `and then`, let SC be False;
if $\star$ is `or else`, let SC be True.

Let BL be the BDD for DL, and BR the BDD for DR.

Then B, the BDD for D is obtained by combining BL and BR as follows:

- the entry point is that of BL
- the exit edge labeled SC of BL is an exit edge labeled SC of B
- the other exit edge of BL connects to the entry point of BR

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- the exit edges of BR are exit edges of B with the same labels

The following invariants of ROBDDs follow from the construction process:

- There is exactly one BDD node for each condition.

- All condition nodes are reachable (i.e. there is a path from the entry point to any node in the BDD).

- There are no cycles in the BDD.

- Both outcomes are reachable (i.e. there is a path from the entry point to an exit edge labeled True and to an exit edge labeled False).

This algorithm is formally described in Alloy, in `build_bdd.als`. Given a decision D, we will now call BDD(D) its (RO)BDD as built by this recursive procedure.

2.3   Node ordering

For a decision D and a condition C in D, let us call index(D,C) the positive number built by the following recursive procedure:

Build_BDD_Order.Condition The sub-decision is a single condition decision, it is of the form:
$D ::= C$
then index(D,C) = 1

Build_BDD_Order.NOT The sub-decision is of the form:
$D ::= \texttt{not } D1$
then for each condition C in D, index(D,C) = index(D1,C)

Build_BDD_Order.Short_Circuit_Operator The sub-decision is of the form:
$D ::= DL \star DR$
then, for each condition C in D:

- if C is in DL, $index(D,C) = index(DL,C)$
- if C is in DR, $index(D,C) = index(DR,C) + \#cond(DL)$

This builds a total order over the conditions of a decision; it is a general result for such a reduced order BDD that the reflexive transitive closure defines a total order over its nodes; this order is the same as the one we just defined. It can also be demonstrated easily, by structural induction, that this order is the order of conditions in the decision's expression.

For a decision D, we will call $root(D)$ and $root(BDD(D))$ the unique node with index 1. This node is indeed the root node of the BDD.

2.4 Evaluation of a decision

Given the transformation of a decision into its ROBDD, evaluating the decision consists in computing its value using the following BDD traversal procedure:

Eval.Condition The value of a decision that consists in a lone condition is the value of the condition.

Eval.Not To evaluate $\texttt{not}(D)$, evaluate D and take the opposite value

Eval.Short_Circuit_Operator To evaluate $(D1) \star (D2)$, evaluate D1. If $D1 = SC$ then the value is SC, else evaluate D2, and the value is that of D2.

This algorithm is modeled in `decision_evaluations.als` The following property holds:

Evals_Are_Paths Evaluating a decision is equivalent to traversing the BDD, evaluating each condition as BDD nodes are traversed, and using the label of the exit edge as the value of the decision.

This property, and most properties that we'll discuss here, is proved by induction on the structure of the decision. For each case of the BDD_Build procedure (Build_BDD.Condition, Build_BDD.Not, Build_BDD.Short_Circuit_Operator), we will assume that the the property holds for the parameters and prove that the build step preserves the property. A formal model for this proof can be found in `bdd_dec_evaluations.als`.

Note that the practical implementation of coverage analysis systems based on control flow traces relies on the assumption that the code generator used to produce executable code from expressions actually implements this evaluation strategy.

2.5 Equivalence case

We now consider the case of an expression whose BDD has no diamond paths, i.e. for each BDD node there is exactly one path from the entry point to that node. The following property holds:

BDDBC_No_Diamond_Indep_Implies_MCDC For a BDD with no diamonds, if conditions are independent, then BDD branch coverage implies Unique Cause MC/DC and Masking MC/DC.

Let's consider a condition C. Since we have BDD branch coverage, all possible paths starting at C have been taken (by recurrence on path length, taking advantage of no cycles and no diamonds).

From the independent outcome reachability property, we have two paths starting at C, beginning each with one edge from C, and ending on the two outcomes of the decision. Let's call them PCT and PCF.

These paths are disjoint: any condition appearing in one is not evaluated in the other (because of no-diamond).

These two paths are parts of paths PT and PF from the BDD entry point to either outcome, and they cannot differ on the part of the path from the entry point to C.

So, PT and PF differ in C, in no other condition before C, and in no other non-masked condition after C, so they prove independent influence of C over the decision.

This holds for each condition in the decision, so Unique Cause MC/DC is proved; and since Unique Cause is stronger than Masking MC/DC, Masking MC/DC is proved as well.

This property and the non-equivalence case that follows is modeled in `bdd_coverage.als`.

2.6 Non-equivalence case

The general idea of this proof is to show that, for any BDD with a diamond, we can build a set of evaluations that covers the BDD branches in such a way that there is at least one condition for which MC/DC is not met.

If we want to have a proof that will work for any definition of MC/DC, we cannot rely on a failure of the independence criteria; *to independently affect* means something

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different in Unique Cause MC/DC or in Masking MC/DC. So we should rather rely, if it is possible, on the set of properties that these criteria have in common; a sort of *greatest common divisor* of the two criteria. Let us define this Weak MC/DC criteria as follow:

- every possible outcome of the decision has been tried;

- each condition in the decision has taken on every possible outcome;

- each condition in the decision is shown to affect the outcome of the decision.

The only difference with MC/DC here is that Weak MC/DC does not care if the condition *independently* affects the outcome of the decision; whatever *independently* means. Weak MC/DC is weaker than any other MC/DC criteria. So, if a set of evaluations does not satisfy Weak MC/DC, it won't satisfy Unique Cause MC/DC or Masking MC/DC.

The formal definition would be:

DEFINITION 4 *Given a decision D, a pair of truth vectors satisfies Weak MC/DC for a condition c if and only if, in their influence tree, the node c and the root node are both colored in True.*

A test set satisfies Weak MC/DC for a decision D if, and only if, for each condition in cond(D), there exists a pair of tests in the test set that satisfies Weak MC/DC.

It turns out that we can build a set of evaluations such that BDDBC is reached, but not MC/DC, when the BDD has a diamond. Let's take the canonical example:
(*A* and then *B*)or else *C*
and these evaluations:

A	B	C	(A and then B) or else C
T	T	x	T
T	F	T	T
F	x	F	F

This set covers this decision for BDDBC. However, this does not meet Weak MC/DC; whenever B is evaluated, the decision outcome is True. So this cannot show that B affects the outcome of the decision.

Let us give a general idea of how we would prove that in the general case. First, remember that we are dealing with reduced ordered BDDs; so each node can be ordered. This property allows us to have a proper concept of "last diamond node" and its "last parent". In our example, the last diamond node would be C, its last parent B.

Now, we would show that the last parent of the last diamond node has an interesting property: one of its exit edge is always connected directly to an outcome (for B, it is outcome True). Its other exit edge being connected to the last diamond node (obviously), it is possible to cover this edge in such a way that the outcome of the corresponding evaluation is the same as the "direct outcome"; e.g. when evaluating B to False, we would evaluate C to True and exit on True. We know that this is possible using the property that, from each node of the BDD (and, in this case, the diamond C), there exists at least one evaluation that reaches outcome True and at least one that reach outcome False.

This means that we can cover all exit and incoming edges of the last parent of the last diamond in such a way that the evaluations *always* exit on the same outcome; and, from the property of its exit edges, we can see that this set of evaluations can be completed to reach BDDBC *without evaluating this node anymore*. This builds a BDD coverage that does not satisfy Weak MC/DC for this node.

The following sections will detail this proof.

2.6.1 Last diamond

We have previously built a function index(D,C) that defines a full order over the nodes of a (RO)BDD. Based on this function, we can define the following entities:

- in a BDD, let the last node be the node with the greatest index;

- let a diamond node be a node with more than one parent;

- let the last diamond node of a decision be the diamond with the greatest index;

- let the last parent of a node be the parent with the greatest index.

Some examples from our canonical case $D ::= (A$ and then $B)$or else C:

- index(D,A) = 1, index(D,B) = 2, index(D,C) = 3;

- the last diamond node is C; it is also the last node;

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- the last parent of C is B.

We have the following properties:

Lemma 2.6.1 *The two exit edges of the last node are connected to both outcomes.*

Lemma 2.6.2 *If a BDD contains diamonds, then the last parent of the last diamond has an exit edge that is directly connected to an outcome.*

This outcome will be called the direct outcome (of the last of parent of the last diamond). The edge connected to the the direct outcome will be called the direct exit edge.

This can be seen in our example:

- the last node being C, its exit edges are connected directly to True and False;

- the last parent of the last diamond being B; when it is True, the outcome True is reached.

The two properties can be proved by structural induction on the BDD. The first is the easiest one and we will keep it as an exercise for the reader. Here is a proof of the second one:

- if $D ::= C$ (simple condition decision):
  the BDD does not contain any diamonds, so the property is trivially true in this case.

- if $D ::= \text{not } D1$:
  if D1 does not contain diamond, D does not either, the propery is trivial as well. If it does contain diamond, well, only outcome labels are different between BDD(D) and BDD(D1), so the property holds for D if it holds for D1.

- if $D ::= DL \star DR$, supposing that the property holds for DL and DR; four cases then:

  - If no diamonds were in BDD(DL) and BDD(DR), and none were introduced by building BDD(D) from them: then BDD(D) contains no diamonds, the property is trivial.

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- If DR contains a diamond: then the last diamond node of D is the last diamond node in DR. By construction, BDD(DR) is a sub-tree of BDD(D), the exit edges of the last diamond node are the same in BDD(DR) and in BDD(D). So if the property holds for DR, it holds for D.

- If DR contains no diamonds, and if building BDD(D) from BDD(DR) and BDD(DL) introduces a new diamond: this new diamond node has to be root(DR), as only this node has gained new incoming edges: in Build_BDD, the exit edges of BDD(DR) labeled "not SC" are connected to BDD(DL)'s root. Therefore, its parent nodes are all in cond(DL). And the last node of BDD(DL) is one of these parent nodes; by property 2.6.1, it has an exit edge to "not SC" in BDD(DL). Its other exit edge will remain unchanged in BDD(D) and will be connected to an outcome ("SC"). This last node of BDD(DL) is also the last parent of this diamond node, as no other nodes in cond(DL) have a greater index. And (finally), this introduced diamond node is D's last diamond node, as DR contains no diamonds (initial hypothesis). So we have identified the last parent of the last diamond and showed that one of its exit edges is connected to an outcome in BDD(D); that's the property that we were trying to prove.

- If DR contains no diamonds, and if no diamonds were introduced when building BDD(D), but if DL contains a diamond: the last diamond node in D is the same node as the last diamond node in DL. The last parent of the last diamond in DL has an exit edge that connects to SC; otherwise, it would be connected to BDD(DR)'s root when building BDD(D), and as the last node would also be connected to this root node (it also has an exit edge to SC in DR, as a consequence of property 2.6.1), BDD(DR)'s root would have at least two fathers in D, which contradicts the hypothesis that no diamonds were introduced. So this exit edges of the last parent of the last diamond in DL (and D) will still be connected to an outcome after building D. So the property is true in this case as well.

The property 2.6.2 has now been established in all possible cases. This proof will actually be quite useful as its structure will be used to build a coverage that satisfies BDDBC but not Weak MC/DC.

2.6.2 Path though BDDs - Conventions

Some conventions first to help us manipulate paths through BDD:

- For two paths pl, pr into two different BDDs, concat(pl, pr) is the concatenation of these two paths obtained by replacing the outcome of pl by the first node in pr.

- For two set of paths PL, PR through two different BDDs, merge(PL, PR) is the set built by doing an arbitrary mapping of each element of PL to each element of PR, and concatenating each pair; as these two sets may not have the same number of elements, all the remaining elements of the greatest set will be mapped to one arbitrary element of the smallest set.

- For any set of paths PS through a BDD (each going from the root node to an outcome), let us call path_to(True, PS) the subset of paths reaching outcome True, and path_to(False, PS) the subset reaching False. We have the obvious property:

$$PS = path\_to(True, PS) + path\_to(False, PS)$$

2.6.3 Building a BDD branch coverage that does not verify Weak MC/DC

For a given decision D, our coverage will be ensured by the union of three sets of paths Short_Circuit_LPLD(D), Long_Circuit_LPLD(D) and LPLD_Not_Evaluated(D), which verifies the following set-specific invariants:

Short_Circuit_LPLD_Invariant If BDD(D) is diamondless, Short_Circuit_LPLD(D) is empty; otherwise, it contains a non-empty set of paths, each reaching the last parent of the last diamond node and then exiting on its direct outcome.

Long_Circuit_LPLD_Invariant If BDD(D) is diamondless, Long_Circuit_LPLD(D) is empty; otherwise, it contains a non-empty set of paths, each reaching the last parent of the last diamond node, then going to the last diamond node, and then reaching the same outcome as Short_Circuit_LPLD(D).

LPLD_Not_Evaluated_Invariant No path in LPLD_Not_Evaluated(D) evaluates the last parent of the last diamond node.

...plus one other "global" invariants:

BDDBC_Coverage_Invariant The union of these three sets covers each edges of the considered BDD.

As a consequence of this last invariant, we will now call BDDBC_Coverage the union of these three sets:

$$BDDBC\_Coverage(D) = Short\_Circuit\_LPLD(D)$$
$$+ Long\_Circuit\_LPLD(D)$$
$$+ LPLD\_Not\_Evaluated(D)$$

Note also that an other property falls naturally from LPLD_Not_Evaluated_Invariant and BDDBC_Coverage_Invariant:

Incoming_Edges The union of Long_Circuit_LPLD and Short_Circuit_LPLD covers all incoming edges of the last parent of the last diamond node.

Anyway, here is the definition, case by case, of a recursive build procedure that builds such sets from a decision D. As always when doing a structural induction, it assumes that all its sub-decisions have such sets verifying these invariants (not considering D as a sub-decision of D, obviously; "strict" sub-decisions):

Build_BDD_Cov.Condition The sub-decision is a single condition decision, it is of the form:
$D ::= C$
In this case:

$$Short\_Circuit\_LPLD(D) = Long\_Circuit\_LPLD(D) = \{\}$$
$$LPLD\_Not\_Evaluated(D) = \{C- > True, C- > False\}$$

Build_BDD_Cov.NOT The sub-decision is of the form:
$D ::= \texttt{not } D1$

Then Short_Circuit_LPLD(D) is built by switching the outcome of each path contained in Short_Circuit_LPLD(D1). Same operations for Long_Circuit_LPLD(D) and No_LPLD(D).

Build_BDD_Cov.Short_Circuit_Operator The sub-decision is of the form:
$D ::= DL \star DR$

Four sub-cases:

- No diamonds:
  No diamonds in BDD(DL) and BDD(DR), and none were introduced by

building BDD(D) from them. Then:

$$Short\_Circuit\_LPLD(D) = \{\}$$
$$Long\_Circuit\_LPLD(D) = \{\}$$
$$LPLD\_Not\_Evaluated(D) = path\_to(SC, BDDBC\_Coverage(DL))$$
$$+ merge(path\_to(notSC, BDDBC\_Coverage(DL)),$$
$$BDDBC\_Coverage(DR))$$

- Diamond in DR:
  i.e. Short_Circuit_LPLD(DR) and Long_Circuit_LPLD(DR) contain at least one element each. Let p_dl be an arbitrary path to "not SC" in BDD(DL), taken from BDDBC_Coverage(DL). In BDD(D), it corresponds to a path to root(DR). Then:

$$Short\_Circuit\_LPLD(D) = merge(\{p\_dl\}, Short\_Circuit\_LPLD(DR))$$
$$Long\_Circuit\_LPLD(D) = merge(\{p\_dl\}, Long\_Circuit\_LPLD(DR))$$
$$LPLD\_Not\_Evaluated(D) = path\_to(SC, BDDBC\_Coverage(DL))$$
$$+ merge(path\_to(notSC, BDDBC\_Coverage(DL)),$$
$$LPLD\_Not\_Evaluated(DR))$$

- Last diamond introduced:
  i.e. no diamonds in BDD(DR) and a diamond has been created when building BDD(D). In this case:
  - $Short\_Circuit\_LPLD(DR) = Long\_Circuit\_LPLD(DR) = \{\}$
  - the last diamond node in D is root(DR).

Let SC_DL be the subset of paths in path_to(SC, BDDBC_Coverage(DL)) that evaluates the last node in BDD(DL). Similarly, let LC_DL be the subset of paths in path_to(not SC, BDDBC_Coverage(DL) that evaluates the last node in BDD(DL); in BDD(D), this one corresponds to a path to root(DR). Let $REST\_DL = BDDBC\_Coverage(DL) - (SC\_DL + LC\_DL)$ be the subset of paths in BDDBC_Coverage(DL) that do not evaluates this last node. Let sc_dr be an arbitrary path in DR exiting on SC.
Then the coverage for BDD(D) is built as follows:

$$Short\_Circuit\_LPLD(D) = SC\_DL$$
$$Long\_Circuit\_LPLD(D) = merge(LC\_DL, \{sc\_dr\})$$
$$LPLD\_Not\_Evaluated(D) = path\_to(SC, REST\_DL)$$
$$+ merge(path\_to(notSC), REST\_DL),$$
$$BDDBC\_Coverage(DR))$$

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(Note that we are allowed to merge path_to(not SC, REST_DL) because it is non-empty; there is at least one other node that has an exit edge directly connected to "not SC" in BDD(DL), otherwise no diamond would have been created; so it must have at least one path that does not evaluate the last node in BDD(DL); that property gives us the right to use the merge operation. Same thing for LC_DL and SC_DL that are both non-empty thanks to property 2.6.1).

- Diamond in DL only:
  i.e. no diamonds in BDD(DR), no new diamonds introduced when building BDD(D), but at least one diamond in BDD(DL). The last parent of the last diamond in DL has its "direct" exit edge that connects to SC; otherwise, it would have been connected to BDD(DR)'s root when building BDD(D), and as the last node would also be connected to this root node (it also has an exit edge to SC in DR, as a consequence of property 2.6.1), BDD(DR)'s root would have at least two fathers in D, which contradicts the hypothesis that no diamonds were introduced. Let sc_dr be an arbitrary path in DR exiting on SC. Then the coverage for BDD(D) is built as follows:

$$Short\_Circuit\_LPLD(D) = Short\_Circuit\_LPLD(DL)$$
$$Long\_Circuit\_LPLD(D) = merge(Long\_Circuit\_LPLD(DL), \{sc\_dr\})$$
$$LPLD\_Not\_Evaluated(D) = path\_to(SC, LPLD\_Not\_Evaluated(DL))$$
$$+ merge(path\_to(notSC, LPLD\_Not\_Evaluated(DL),$$
$$BDDBC\_Coverage(DR))$$

Our sets are now defined in all possible cases. It is quite easy to check that Short_Circuit_LPLD_Invariant, Long_Circuit_LPLD_Invariant, LPLD_Not_Evaluated_Invariant and BDDBC_Coverage_Invariant are enforced by this build procedure; each one falls so obviously from the construction (in every case) that it will be quite painful to elaborate.

So we can build a BDD coverage in such a way that, if there is a diamond in the BDD, then the decision will have the same outcome for any evaluation where the last parent of the last diamond is evaluated. This means that this BDD coverage does not imply Weak MC/DC; that which was to be demonstrated.

2.7 Conclusion

We have therefore proved the following property:

THEOREM 4 *Given a decision D, branch coverage of the BDD implies MC/DC if, and only if, there is no diamond in the BDD (i.e. no node of the BDD is reachable through more that one path from the root).*

Intuitively, BDDBC is a local property of BDD traversals (i.e. of evaluations of the decision): it is evaluated individually for each BDD node, without respect to how the BDD node was reached. In contrast, MC/DC is a non-local property, since it involves the complete path through the BDD. To establish independent influence of condition C, it is necessary to study what happens for two values of C leading to different outcomes, *with all other conditions fixed or masked*.

3   A second characterization of the equivalence case

In this section we provide an alternative (equivalent) property that characterizes cases where BDD branch coverage implies MC/DC:

THEOREM 5 *Given a decision D, BDD branch coverage implies MC/DC if, and only if, when considering the negation normal form D' of D, for every sub-decision E of D', all binary operators in the left-hand-side operand of E, if any, are of the same kind as E's operator.*

The negative normal form is obtained by rewriting the expression using De Morgan's laws so that negations apply only to atomic conditions (and not to more complex subexpressions).

We prove this theorem by showing that this alternative characterization is equivalent to having no diamonds in the associated BDD. Theorem 5 follows by application of Theorem 4.

Let's consider a decision D and its BDD.

There is a diamond in a BDD if and only if there is at least one node such that the number of paths to reach it from root is strictly greater than 1; we will use this caracterization to prove our property.

Proof of the simplified case:

Consider first the case where decision D does not contain any negation. Each node in the BDD corresponds to a sub-decision D' in D (i.e. the root of each sub-decision is a node in D's BDD).

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```
        DL              DL
      F/  \T        F/    \T
      DR    T      F      DR
    F/  \T              F/  \T
    F    T              F    T

    (a)             (b)
  DL or else DR   DL and then DR
```
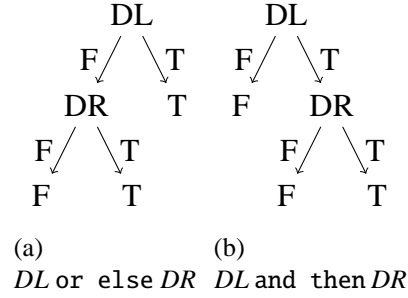
Figure 2: case 2 and 3

Proof of reverse implication (simplified case):

Let's call NF(D) the number of paths from the root of decision D to False (F) and NT(D) the number of paths from the root of decision D to True (T). We then have three cases:

case 1: D = A (atom)
$NF(D) = NT(D) = 1$

case 2: D = *DL* or else *DR*
BDD(D) is constructed from BDD(DL) and BDD(DR). It is shown in Figure 2(a). Then:
$NF(D) = NF(DL) * NF(DR)$
$NT(D) = NT(DL) + NF(DL) * NT(DR)$

case 3: D = *DL* and then *DR*
BDD(D) is constructed from BDD(DL) and BDD(DR). It is shown in Figure 2(b). Then:
$NF(D) = NF(DL) + NT(DL) * NF(DR)$
$NT(D) = NT(DL) * NT(DR)$

Lemma 3.0.1 *For every decision D, $NF(D) \geqslant 1$ and $NT(D) \geqslant 1$.*
If D is of the form or else *then $NT(D) \geqslant 2$.*
If D is of the form and then *then $NF(D) \geqslant 2$.*

Lemma 3.0.2 *NF and NT are monotonic functions, i.e. if D' is a sub-decision of D, then we have $NF(D') \leqslant NF(D)$ and $NT(D') \leqslant NT(D)$.*

Proofs of Lemmas 3.0.1 and 3.0.2 are on structural induction on the form of decisions, based on the three cases distinguished above.

Then, suppose that a decision D of the form and then contains a sub-decision D' of the form or else on the left-hand side. If DL, DR are the two sub-decisions such that $D = DL$ and then *DR*, then D' is also a sub-decision of DL.

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By Lemma 3.0.1, we know that $NT(D') \geqslant 2$. By Lemma 3.0.2, we know that $NT(DL) \geqslant NT(D')$.

Then, in BDD(D), the paths reaching the DR's root node are exactly those paths in BDD(DL) reaching True, by construction; the number of these paths is NT(DL).

Thus, DR's root node in the BDD(DL) is reachable by more than one path. As a consequence, there is a diamond in this BDD.

Similarly, if a decision D of the form `or else` contains a sub-decision D' of the form `and then` on the left-hand side, we can exhibit a node in the BDD(D) that is reachable by more than one path.

By contraposition, if there are no diamonds in the BDD associated to a decision D, then D has no `or else` sub-decision on the left of its `and then` sub-decisions, and no `and then` sub-decision to the left of its `or else` sub-decision.

Proof of implication (simplified case):

Now suppose that for every sub-decision in decision D, the left operand of an `and then` sub-decision contains only `and then` sub-decisions and the left operand of an `or else` sub-decision contains only `or else` sub-decisions.

Lemma 3.0.3 *If E contains only `or else` sub-decisions then $NF(E) = 1$. If E contains only `and then` sub-decisions then $NT(E) = 1$.*

Proof of Lemma 3.0.3 is on structural induction on the form of decisions, based on the 3 cases distinguished above.

By structural induction on the depth of the BDD, we can show that there cannot be any diamonds in the BDD associated to D, which proves that the desired implication holds.

Proof of the general case:

In the general case, decision D may contain `not` sub-decisions. Then, consider D' the negation normal form of D. Since D and D' are represented by the same BDD, Theorem 5 follows.

4   Coupled conditions

Handling of coupled conditions is a well-known problem that one has to face when considering MC/DC, and in particular Unique Cause. It may be stated as follow:

DEFINITION 5 *Two or more conditions are said to be coupled if changing one condition can cause the other condition(s) to change.*

Conditions are said to be strongly coupled if changing one always changes the others. At the contrary, they are said to be weakly coupled if changing one sometimes (but not always) changes the others.

In [VB02], it is proved that strongly coupled conditions are pairs of conditions that are either equal, or complimentary. In other words, if A and B are strongly coupled, then either $A = B$ or $A = $ not B

Without short-circuit operators, decisions with coupling cannot be covered for Unique Cause MC/DC; Masking MC/DC has been defined to solve this problem. When proving Masking MC/DC for a condition C, one may exhibit a pair of evaluation that have different values for other conditions, as long as it can be shown, by a structural analysis of the decision, that changing these other conditions had no influence on the result. Typically, for a decision A and $SUBDEC$ ($SUBDEC$ being a sub-decision, A being a condition), any evaluation that sets $SUBDEC$ to True ensures that only the value of A impacts the value of the decision. e.g. if $SUBDEC ::= (B$ or $C)$, the following evaluation pairs proves Masking MC/DC, but not Unique Cause MC/DC:

A	B	C	SUBDEC ::= B or C	A and SUBDEC
T	F	T	T	T
F	T	F	T	F

B and C change, so Unique Cause MC/DC is not reached; but SUBDEC is kept to True, so Masking MC/DC is reached. So if C and A were strongly coupled ($A = C$), we would still be able to cover this decision for Masking MC/DC; but we would not be able to reach Unique Cause.

When using only short-circuit operators, one may wonder if the problem still holds. It can be seen that the previous example is covered for Unique Cause MC/DC when the operators are short-circuit; in this case, the evaluations are:

A	B	C	A and then (B or else C)
T	F	T	T
F	x	x	F

So there are some cases of coupling that are not problematic in the case of short-circuit operator; short-circuits seem to offer some sort of masking by avoiding to evaluate some sub-decisions. The question would now be: can we show some cases of coupling for which Unique Cause may not be reached even with short-circuit operators? Would these cases show up in real life, or would they be dummy decisions that may be trivially simplified (e.g. *A* and then *A*)?

The following sections sched some light on these problems. First, Masking MC/DC is characterized in terms of BDD, in order to make use of the results that have been demonstrated in the previous sections. This characterization would then allow to comment on coupling in the context of short-circuit operators.

4.1 MC/DC in terms of BDD

4.1.1 Unique Cause MC/DC

THEOREM 6 *Given a decision D, a pair of evaluations of D satisfies Unique Cause MC/DC for a condition C if and only if:*

- *both reach C using the same path through BDD(D);*

- *their paths from C exits on two different outcomes and do not cross each other (C excluded).*

Let us first prove that this characterization implies Unique Cause MC/DC as defined previously.

Both evaluations reach C using the same path; this implies that all conditions that are before C are either not evaluated by any of the two evaluations, or that they have the same value. In both cases, they are colored with False in the influence tree.

After C, their paths through the BDD do not cross each other; this means that they evaluate C to both values, otherwise they would have at least one edge in common. C is evaluated to True and False, which means that it is colored with True in the influence tree.

An other consequence of this hypothesis is that all conditions after C are not evaluated in at least one of the two paths; a condition that would be evaluated on both paths would be a meeting point of the two paths, which would contradict the hypothesis. This means that any condition after C is colored with False as, by definition, anything xor Not_Evaluated gives False.

So, of all conditions, only C is colored with True. As the two paths exits on two different outcomes, the root node of the influence tree is colored with True as well. We now have these general properties of influence trees:

Lemma 4.1.1 *If the root node of an influence tree is colored with True, then its influence set is not empty.*

If the root node of an influence tree is colored with True, and if only one condition c is colored with True, then the corresponding influence set is {c}

This lemma can be proved by induction on the structure of a decision; the first property is easier to prove and comes in handy to prove the second property. We will not detail these proofs here as they offer no technical difficulty.

Applying the second property to our case, we can now prove that the influence set of our decision is the singleton {C} ; so the two evaluations satisfy Masking MC/DC. As all other conditions are colored with False in the influence tree, Unique Cause MC/DC is satisfied.

To demonstrate the other side of the equivalence, let us assume that one of the two properties on the BDD is violated and let us show that Unique Cause MC/DC cannot be achieved.

If the two paths are not identical before the considered condition: they start from the same node (the root of the BDD) and ends up on the same node (C), so it means that there exists at least one node from which those paths diverge (this is a general property of paths through graphs). This node takes therefore two different values in the two evaluations; so it is colored with True in the influence tree. It is the second condition colored with True in the influence tree (with C), so Unique Cause MC/DC cannot be achieved.

If the paths from C cross each other, then there is at least one condition after C that is evaluated in both paths. Let us call lc the last condition where the two paths cross. By definition of lc, the two sub-paths that start from lc and go to outcomes do not cross each other. This means that they exit by two different edges from lc; otherwise, the node that they would reach would be a common node after lc, which would contradict the fact that lc is the last common node. So lc takes two different values, True and False, on these two paths. As a consequence, lc is colored with True in the Influence Tree, and therefore C is not the only one colored with True, so Unique Cause MC/DC cannot be achieved.

The two sides of the equivalence in Theorem 6 are now proved.

4.1.2 Masking MC/DC

THEOREM 7 *Given a decision D, a pair of evaluations of D satisfies Masking MC/DC for a condition C if and only if:*

- *both reach C;*

- *their paths from C exit on two different outcomes and do not cross each other (C excluded).*

The difference with Unique Cause is that the paths to reach C may be different. That is to say: only the sub-bdd of BDD(D) whose root is C is taken into account when proving Masking MC/DC; anything before C does not matter (as long as C is reached). An other way to see that: Masking MC/DC is achieved for C if Unique Cause MC/DC is achieved on the sub-bdd of BDD(D) whose root is C. Or: any condition on the left of C is not considered when trying to achieve Masking MC/DC for C.

This may be justified as follow: we have seen that short-circuit operators do provide some sort of masking in our original example A and then $SUBDEC$. Now, consider $SUBDEC$ and then A; in this case, SUBDEC will always be executed before A, so the short-circuit operator and then will not mask SUBDEC. Short-circuit operators are non-commutative, and introduce this asymetricity. Say, if $SUBDEC ::= B$ or else C, then the following evaluations satisfy Masking MC/DC for C, but not Unique Cause:

A	B	C	SUBDEC ::= B or else C	SUBDEC and then A
T	F	T	T	T
F	T	x	T	F

For Masking MC/DC, we have to ignore the details of the evaluation of SUBDEC, as long as it ends up being evaluated to True. In terms of BDD, that means that we do not care about how we reached the considered condition, as long as we reached it.

A formal model of this characterisation can be found in `bdd_coverage.als`, and a proof of equivalence in a small scope (at most 4 conditions per decisions) in `bdd_dec_coverage.als`.

Here is a proof in the general case. Yet another way to see Theorem 7 is: a pair of evaluations proves Masking MC/DC for a condition C if and only if C is the last node where the two paths to (different) outcomes crosses each other and diverges. Nodes where these two paths diverges are exactly the conditions colored with True in the influence

tree; and the last node is the righmost condition. This means that Theorem 7 is directly implied by Lemma 1.2.1.

Let us now comment a bit about coupling:

- We have seen that the only difference between Unique Cause and Masking MC/DC is the constraint on the way to reach the considered condition: the same path for both evaluations for Unique Cause, any path for Masking MC/DC. In a BDD, the path to reach a node is unique in the case of no diamond; that shall encourage us to have a closer look at this special case.

- When we commented about Masking MC/DC, we saw that the asymetricity of short-circuit was the problem: when two conditions are coupled, the leftmost one can take advantage of the partial masking that short-circuit operators provide; however, the rightmost cannot. Now, can we any problematic case for this rightmost condition? This will be discussed in the following sections.

4.2 Coupled conditions without diamonds

First, it can be seen that strong coupling only gives degredated cases, where some branches of the BDD cannot possibly be covered; this is what the following theorem states:

THEOREM 8 *A BDD with no diamonds and with strongly coupled conditions cannot be covered for BDD branch coverage, Masking MC/DC, Unique Cause MC/DC.*

As Unique Cause MC/DC and Masking MC/DC are stronger than BDD branch coverage, we only need to prove this theorem for BDD branch coverage.

In a BDD with no diamonds, there is only one possible path to each node, traversing all the previous conditions in the BDD ordering. Take two strongly coupled conditions; let us call fc the first one in the BDD ordering, lc the second one. When lc is executed, fc has been evaluated to a given value, and this value is the same for any evaluation that reaches lc. Let us call this value V. Now, as lc and fc are strongly coupled, there are only two possibilities: either lc is always equal to fc, and evaluates to V, in which case the exit edge for `not` V is never executed; or it is always the complementary of c, and evaluates to `not` V, in which case the other exit edge is never executed. In any case, BDD branch coverage cannot be achieved, as at least one edge is never taken.

This means that there are no meaningful cases of strong coupling without diamonds. They could typically be optimizied out by the compiler.

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We can however exhibit some cases of weak coupling that makes sense. For instance:

Dec ::= $t > T$ `or else` $(t = T$ `and then` $A)$

Or an even more common pattern, that can often be found in C:

```
name != NULL && strcmp (name, "something")
```

In these two examples, BDD branch coverage can be reached. For the first one, it can be checked that these evaluations cover the decision for BDD branch coverage, Masking MC/DC and Unique Cause MC/DC:

t > T	t = T	A	Dec
T	x	x	T
F	F	x	F
F	T	T	T
F	T	F	F

It can be seen that Unique Cause MC/DC and Masking MC/DC are equivalent in the case of no diamonds, as there is only one possible path to each node; this is a consequence of Theorem 7 and Theorem 6. So, with no diamonds, Masking MC/DC does not gain us anything, comparing to Unique Cause MC/DC (or BDD branch coverage):

- in the cases of strong coupling, none of these coverage criteria can be reached;

- in all other cases, all of them are equivalent anyway.

If Masking MC/DC is of some help to handle coupled conditions, it has to be when the decision's BDD contains diamonds. That is the second case that we will investigate.

4.3   Coupled conditions with diamonds

The Theorem 8 cannot be extended to the case of diamonds. Take for instance this decision:

Dec ::= $(A$ `and then` $B)$ `or else` $($`not` A `and then` $C)$

This decision is a common pattern, logically equivalent to `if` A `then` B `else` C. It cannot be covered for Unique Cause MC/DC, because the first and the third condition are strongly coupled; but it can be covered for Masking MC/DC, with the following evaluations:

A	B	not A	C	Dec
F	x	T	T	T
F	x	T	F	F
T	T	x	x	T
T	F	F	x	F

More precisely: the third condition cannot be covered for Unique Cause; but the first and the last evaluation cover it for Masking MC/DC.

Another example with weak coupling:

$Dec ::= (t < T$ and then $A)$ or else $(t = T$ and then $B)$ or else $(t > T$ and then $C)$

...which is another case where the logical space is partioned (into three cases here: $t < T$, $t = T$, $t > T$) using conditions that are such that exactly one is True for each evaluation.

From the truth table, the only condition for which we cannot reach Unique Cause is $t > T$:

- when it is True, both $t < T$ and $t = T$ have been evaluated to False;

- when it is False, either $t < T$ or $t = T$ has been evaluated to True;

so we cannot reach Unique Cause for this one. We can cover it for Masking MC/DC though, with any of the pairs (eT, eF1) and (eT, eF2), these three evaluations being:

- eT =
  $(t < T$ = False, A = Not_Evaluated,
  $t = T$ = False, B = Not_Evaluated,
  $t > T$ = True, C = True)

- eF1 =
  $(t < T$ = True, A = False,
  $t = T$ = False, B = Not_Evaluated,
  $t > T$ = False, C = Not_Evaluated)

- eF2 =
  $(t < T$ = False, A = Not_Evaluated,
  $t = T$ = True, B = False,
  $t > T$ = False, C = Not_Evaluated)

One could object that this decision may be rewritten as follow:

(A and then $t < T$) or else (B and then $t = T$) or else (C and then $t > T$)

...in which case Unique Cause MC/DC could be achieved. However, such a rewriting may not be possible; the execution of condition may have the precondition that $t < T$, in which case the rewriting would not be valid. So re-ordering the decision to allow Unique Cause MC/DC to be reached is not an appropriate general solution to the coupling problem.

In other words, Masking MC/DC is still useful in the case of short-circuit operators; but only when there are diamonds in the decision's BDD.

5 Computing Masking MC/DC vectors

We have given earlier a simple way to determine if a pair of evaluations proves Masking MC/DC for a condition if only short-circuit operators are used. It was Theorem 3: if a decision Dec contains only short-circuit operators, Masking MC/DC is reached for a condition C if and only if there exists a pair of evaluation (e1, e2) such that Dec[e1] = not Dec[e2], and C is the rightmost condition that is evaluated to two different values in e1 and e2.

In order to have an algorithm to generate a Masking MC/DC coverage of a decision, one may actually simplify even more the identification of a MC/DC pair. It is indeed possible to determine which condition may be proven to have an independent effect from a given evaluation, and identify a polarity of this evaluation; an evaluation pair for a condition C is then constituted by two evaluated marked for C with a different polarity.

The marker of this polarity for a condition C will there be noted t(C) and f(C), depending on the value of C in the evaluation. For instance, for (A and then B) or else C:

	A	B	C	Dec	Markers
(1)	T	T	X	T	t(A), t(B)
(2)	F	X	T	T	t(C)
(3)	F	X	F	F	f(A), f(C)
(4)	T	F	T	T	t(C)
(5)	T	F	F	F	f(B), f(C)

One property of Masking MC/DC that is worth mentioning is that any evaluation may be used to prove MC/DC on some condition of the decision. From Theorem 3, if you have a evaluation e1, you may take any evaluation e2 such that Dec[e1] = not Dec[e2] (there is always such an e2 if the decision is not a simple boolean constant), and then the pair e1, e2 proves MC/DC for some condition in the decision. In the previous example, this means that all lines of truth tables have a least one marker.

To build the influence pairs, one would then have to build (f(x),t(x)) pairs for each x in A, B, C. And one would want to choose the minimal number of evaluations that covers all t(x) and f(x). This is exactly the optimization version of the set covering problem, which is a well known (NP-hard) problem.

Can we do better than that from the structure of the decision? e.g. get a minimal set without having to build the whole truth table? Not really. The thing is that we cannot know, from the structure of the decision, which evaluations should be prefered. That is really something that depends on the context of execution. In particular, coupling cannot be deduced from the structure of the decision; we have to know more about the context of the execution to deduce which evaluations are actually to be removed from the truth table. e.g. if the decision that we considered was actually:

$(N1 = 0$ and then $N2 = 0)$ or else $N1 > 0$

...(that is to say: A ::= $N1 = 0$, B ::= $N2 = 0$, C ::= $N1 > 0$), we would not be able to deduce, from the structure of the decision, that eval (4) is impossible. That is really something that only an analysis of the context of execution (e.g. by abstract interpretation) can deduce.

So, to compute the set of evaluations to reach Masking MC/DC on a decision, one could typically:

- build the annotated truth table from the structure of the decision;

- let abstract interpretation remove entries that are impossible (because of coupling);

- use a well-known algorithm to solve the Covering Set problem.

The following sections will detail the first part.

5.1 Marking an evaluation

This section will first define formally the markers f(C)/t(C) and will prove its relationship with Masking MC/DC: that a pair of evaluations (eT,eF) is marked with (t(C), f(C))

if and only if it proves Masking MC/DC for C.

Let e be a valid evaluation vector for a decision; e(C) being the value of condition C in this decision.

For such an evaluation, we annotate the syntax tree with both the evaluation of each sub-decision and the MCDC markers. The algorithm is the following:

BUILD_MARKERS.Simple_Condition_Decision The sub-decision is a single condition decision, it is of the form:
$DEC ::= C$
In this case:
$eval(C, e) = e(C)$
and, for markers:

- if C is evaluated to True in e:
  $markers(DEC, e) = \{t(C)\}$

- if C is evluated to False:
  $markers(DEC, e) = \{f(C)\}$

- and otherwise (if e(C) is not evaluated):
  $markers(DEC, e) = \{\}$

BUILD_MARKERS.Not The sub-decision is of the form:
$DEC ::= not\,DEC1$
In this case:
$eval(DEC, e) = not\,eval(DEC1)$
$markers(DEC, e) = markers(DEC1)$

BUILD_MARKERS.Short_Circuit The sub-decision is of the form:
$DEC ::= DECL \star DECR$
In this case:
$eval(DEC, e) = eval(DECL, e) \star eval(DECR, e)$
and, for markers:

- if $eval(DEC, e) = eval(DECL, e) = eval(DECR, e)$:
  $markers(DEC, e) = markers(DECL, e) + markers(DECR, e)$
  (we will call this case "both")

- if $eval(DEC, e) = eval(DECL, e) \neq markers(DECR, e)$:
  $markers(DEC, e) = markers(DECL, e)$
  (we will call this case "left")

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- if *eval(DEC, e) = eval(DECR, e) ≠ markers(DECL, e)*:
  markers(DEC, e) = markers(DECR, e)
  (we will call this case "right")

The fact that eval(DEC, e) corresponds the evaluations of the (sub-)decision is pretty obvious.

In BUILD_MARKERS.Short_Circuit, "left", "right" and "both" cover all possible cases ; we have either eval(DECL, e) or eval(DECR, e) (or both) that have the same value as eval(DEC, e). That's a obvious property of and, or, and then, or else... Let just maps the three cases on the three possible evaluations, just to make this clear:

eval(DECL, e)	eval(DECR, e)	eval(DEC,e)	case
not SC	not SC	not SC	"both"
not SC	SC	SC	"right"
SC	X	SC	"left"

Now let us prove this property that states the relation between these markers and MC/DC:

THEOREM 9 *Let eT, eF be two evaluations of DEC such that:*

- *eT(C) = T*

- *eF(C) = F*

(eT, eF) verifies Masking MC/DC for C in DEC if and only if:

- *t(C) is in markers(DEC, eT)*

- *f(C) is in markers(DEC, eF)*

This will be proved by induction on the decision structure, and we will only consider the only case that is not totally trivial: BUILD_MARKERS.Short_Circuit. So we have *DEC ::= DECL ⋆ DECR*, we assume that the property is true for DECL and DECR and show that is true for DEC as well. Let use consider the two sides of this equivalence separately:

- Let us suppose that (eT, eF) proves Masking MC/DC for a condition C, and let us prove that t(C)/f(C) are in markers(DEC, eT)/markers(DEC, eF):

  If C is in DECL: the root of DEC is colored with True in the influence tree, and the root of DECL is colored also with True in the influence tree, otherwise there would not be a path of True from C to the root of DEC.

  DEC's root and DECL's being both colored with True, they have both changed between eT and eF. This means that one of these two vectors is in the "both" case, and the second is in the "left" case. In both cases, the markers of DECL is a subset of the markers of DEC.

  The root of DECL being colored with True, we have Masking MC/DC for C in DECL. By using the induction property on DECL, we can deduce that t(C) is in markers(DECL, eT), and f(C) in markers(DECL, eF). As we just proved that the markers of DECL are a subset of the marker of DEC, we therefore have t(C) in markers(DEC, eT) and t(F) in markers(DEC, eF).

  If C is in DECR: the same reasoning shows that (eT, eF) are "right" and "both". The same conclusion follows.

- Let us assume that t(C)/f(C) are in markers(DEC, eT)/markers(DEC, eF), and let us prove that (eT, eF) proves Masking MC/DC for a condition C:

  t(C) and f(C) can only come from the sub-decision that contains C. This rules out the case ("right", "left"); indeed, in this case:
  markers(DEC, "right") = markers(DECR, "right")
  markers(DEC, "left") = markers(DECL, "left")
  ...which would mean that either markers(DEC, eT) or markers(DEC, eF) would contain no marker for C (markers from C come either from DECL, or from DECR, but not from both!), which contradicts the hypothesis. So we have only two possible cases: either ("both", "left") or ("both", "right").

  If C is in DECL: then t(C) comes from markers(DECL, eT) and f(C) comes from markers(DECL, eF). So we can use the induction hypothesis to deduce that (eT, eF) proves Masking MC/DC for C in DECL. This means that DECL takes two different values with eT and eF; as we ruled out the case ("right", "left"), it means that we are in the case ("both", "left"). In this case, DEC changes its value as well, so it is colored with True in the influence Tree. Which means that we have a path of True from C to the root of this influence tree, and as the influence set is a singleton (per Lemma 1.2.1), this tells us that (eT, eF) proves MC/DC for C in DEC.

  If C is in DECR: same thing with the case ("both", "right").

5.2 Marking a truth table

The previous algorithm was about determining the markers to associate to a given evaluation; this one will compute the same thing for the whole truth table. it is expressed here in a general formalism: Linear Logic. It should then be possible deduce, from this formalism, an implementation tailored to the tool that we target. There are some comments about implementation at the end of this section.

For now, we will be in the perfect world of Linear Logic expressions.

5.2.1 Linear Logic expressions

The idea would be to go through the structure of a decision and to produce an expression that records the relation between:

- a producer, that takes decision evaluations and produces markers t(c) or/and f(c);

- a consumer, a dual actor that gives evaluations and receives markers.

As we will see, this scheme fits quite neatly to both the structure of the decision and the truth table that we want to produce.

In the following, in order to easily talk of the forms of the expressions that the algorithm manipulates, we will gives some grammar fragments in BNF. First fragments: we will be talking about conditions, which will be atoms in this context, and the MC/DC markers that we introduced previously:

$$< condition > ::= A|B|C|...$$
$$< marker > ::= ('f('< condition >')')|('t('< condition >')')$$

It is not the purpose of this paper to give a comprehensive introduction to Linear Logic. Only a tiny understanding of this substructural logic is needed in our context. [DD92] is a good introduction; in particular, it gives an informal semantics that is good enough to understanding the algorithm that we will develop. We will just give here a similar informal interpretation in the context of our algorithm.

linear negation   The linear negation express the producer-consumer relationship: if a ressource is received on one side, it is given on the other side. Take for example a marker $t(C)$: $t(C)^{\perp}$ expresses the fact that this marker is expected (by the consumer); on the contrary, $t(C)$ expresses the fact that this marker in given up (by the producer).

As classic negation, linear negation verifies double negation elimination:

$$(a^\perp)^\perp = a$$

Multiplicative conjunction   We need to be able to talk about particular evaluations of a decision; for instance to say that condition A has been evaluated to False, then condition C has been evaluated to True. We will use the multiplicative conjunction[1] for that. To refer to this evaluation, we would say:

$$\{A = F\} \otimes \{C = T\}$$

In this particular case, we may also just note $\{A = F, C = T\}$ as a shortcut.

So the expressions that we just mentioned verifies the following BNF:

$$< condition\_eval > ::= ' \{' < condition >' =' \ ('T'|'F')'\}'$$
$$< decision\_eval > ::= < condition\_eval > [\otimes < decision\_eval >]$$

Multiplicative disjunction   Because of the producer-consumer duality, there is also a multiplicative disjunction for the other side of this relationship. They are related together by De Morgan's laws:

$$(a \otimes b)^\perp = a^\perp \,\invamp\, a^\perp$$

In our example, this gives us:

$$(\{A = F\} \otimes \{C = T\})^\perp = \{A = F\}^\perp \,\invamp\, \{C = T\}^\perp$$

...which would express what the producer expects before delivering marker f(A).

Linear implication   This duality allows to express contracts of the form: if you give me this, I'll give you that. Such a contract is expressed by the linear implication. It is defined as:

$$a \multimap b ::= a^\perp \,\invamp\, b$$

---

[1]Technically, the multiplicative conjunction is commutative in linear logic, so we lose the order in which these two conditions have been evaluated. It does not matter much, we may just order them in an actual implementation. We could also have used Cyclic Linear Logic, in which multiplicative operators are not commutative. But that would complicate a bit the expression of the algorithm, introducing exponentials; we chose to keep it simple.

This is similar to classical/intitionist logic where implication is defined as $\neg a \vee b$. In the following, we may also use the reciprocal form $\circ\!\!-$:

$$a \circ\!\!- b ::= a \,\invamp\, b^{\perp}$$

As the multiplicative disjunction is commutative, those are strictly equivalent, but the second form will be clearer in some expressions.

We will use this operator to attach a marker to an evalation. e.g.

$$\{A = F, C = T\} \multimap f(A)$$

which you would read as: "Give me A = F and C = T, and then I'll give f(A) (which is half of A's independence pair)". That's the producer talking.

We will give the BNF of such expressions as well (<mcdc_markers> is developped later):

$$< mcdc\_vector >::=< decision\_eval >{}' \multimap {}' < mcdc\_markers >$$

Additive conjunction   Now we know how to talk about one evaluation; but we need to be able to talk about several of them, to say something like "this evaluation *and* this evaluation proves MC/DC for this condition". This "and" is the additive conjunction &. e.g. to talk about vector (2) and (4) in our previous example, we would say:

$$(\{A = F, C = T\} \multimap t(C)) \,\&\, (\{A = T, B = F, C = T\} \multimap t(C))$$

i.e. if one wants to have a vector for which C is True and is the only condition that impacts the value of the decision, he may either choose {A=F, C=T}, or {A=T, B=F, C=T}.

The multiplicative disjunction is distributive over the additive conjunction; therefore, linear implication is left-distributive over it:

$$a \,\invamp\, (b \,\&\, c) = (a \,\invamp\, b) \,\&\, (a \,\invamp\, c)$$
$$a \multimap (b \,\&\, c) = (a \multimap b) \,\&\, (a \multimap c)$$

With this property, the additive conjunction may also be used to express the fact that several markers are associated to the same decision evaluation. e.g. eval (3) of our example has two markers f(A) and f(C), so it can be expressed as:

$$\{A = F, C = T\} \multimap (f(A) \,\&\, f(C))$$

As before, BNF for future references:

$$< mcdc\_evals > ::=< mcdc\_vector > [\,'\&'\, < mcdc\_evals >]$$
$$< mcdc\_markers > ::=< marker > [\,'\&'\, < mcdc\_markers >]$$

Additive disjunction   Additive conjunction also has its dual operator: it is the additive disjunction $\oplus$, which verify De Morgan's law:

$$(a \,\&\, b)^\perp = a^\perp \oplus a^\perp$$

The duality of $\oplus$ and $\&$ actually helps expressing the side of a choice in an alternative. e.g. in $A \,\&\, B$, you may choose either A or B; in $A \oplus B$, you may be given A or B, but the choice is on the other side. Obviously, when one side can give $A \,\&\, B$, the other side can receive $A \oplus B$. In our case, this says that the choice between alternative paths is made by the consumer, not by the producer.

The multiplicative conjunction is distributive over the additive disjunction; and, with De Morgan's law, linear implication is right-distributive over it:

$$(a \oplus b) \otimes c = (a \otimes c) \oplus (b \otimes c)$$
$$(a \oplus b) \multimap c = (a \multimap c) \,\&\, (b \multimap c)$$

So, to talk about the two vectors (2) and (4) previously, we may have factorized out t(C):

$$(\{A = F, C = T\} \oplus \{A = T, B = F, C = T\}) \multimap t(C)$$

And, as $\otimes$ is distributive over $\oplus$, this expression is equivalent to:

$$((\{A = F, C = T\} \oplus \{A = T, B = F\}) \otimes \{C = T\}) \multimap t(C)$$

$\{A = F, C = T\} \oplus \{A = T, B = F\}$ is another way to state a known property of Masking MC/DC: when proving independent effect of C, it does not matter which path to C has been taken.

The algorithm will indeed use $\oplus$ to talk about the alternative paths to reach a condition.

Other properties $\invamp$, $\&$, $\otimes$ and $\oplus$ are all commutative and associative. Thanks to associativity of $\invamp$ and De Morgan's law, we have the following property:

$$a \multimap (b \multimap c) = (a \otimes b) \multimap c$$

5.2.2   Algorithm

The algorithm forms two kinds of expressions:

- two of them representing any path from root of BDD to True: paths_to (T, Dec) and paths_to (F, Dec);

- two of them representing one half of any masking MC/DC coverage; MCDC(T, Dec) represents the evaluations that exits on True, MCDC(F, Dec) represents the evaluations that exits on False; The final result being: $MCDC(T, Dec) \& MCDC(F, Dec)$

The algorithm is defined by induction on the structure of the decision:

BUILD_MCDC.Simple_Condition_Decision

$$paths\_to(T, DEC) = \{C = T\}$$
$$paths\_to(F, DEC) = \{C = F\}$$
$$MCDC(T, DEC) = \{C = T\} \multimap t(c)$$
$$MCDC(F, DEC) = \{C = F\} \multimap f(c)$$

BUILD_MCDC.NOT

$$paths\_to(T, DEC) = paths\_to(F, DEC1)$$
$$paths\_to(F, DEC) = paths\_to(T, DEC1)$$
$$MCDC(T, DEC) = MCDC(F, DEC1)$$
$$MCDC(F, DEC) = MCDC(T, DEC1)$$

BUILD_MCDC.Short_Circuit

$$paths\_to(\neg SC, DEC) = paths\_to(\neg SC, DECL) \otimes paths\_to(\neg SC, DECR)$$
$$paths\_to(SC, DEC) = paths\_to(SC, DECL)$$
$$\oplus paths\_to(\neg SC, DECL) \otimes paths\_to(SC, DECR)$$
$$MCDC(\neg SC, DEC) = MCDC(\neg SC, DECL) \multimapinv paths\_to(\neg SC, DECR)$$
$$\& paths\_to(\neg SC, DECL) \multimap MCDC(\neg SC, DECR)$$
$$MCDC(SC, DEC) = MCDC(SC, DECL)$$
$$\& paths\_to(\neg SC, DECL) \multimap MCDC(SC, DECR)$$

A few comments before proving that the results of BUILD_MCDC do represents what they are meant to be... to get an intuition of the linear operators. Consider the case of

(a) $DL \star DR$
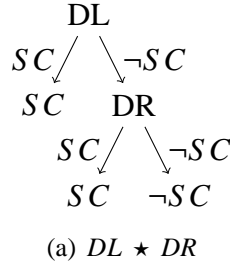
BUILD_MCDC.Short_Circuit (as always, the only one that is really substantial). DEC's BDD can be represented as on figure 3(a)

Paths to SC in DEC contains exactly the paths to SC in DECL and the concatenations of all paths to $\neg SC$ in DECL with all paths to SC in DECR, which corresponds to this expression:

$$paths\_to(SC, DEC) = paths\_to(SC, DECL)$$
$$\oplus\, paths\_to(\neg SC, DECL) \otimes paths\_to(SC, DECR)$$

The concatenations are expressed by $\otimes$ and is assured by its distributivy over $\oplus$, as we will show below.

Similarly, MCDC evaluations that reach SC in DEC are exactly MCDC evaluations that reach SC in DECL and the concatenations of all paths to $\neg SC$ (all paths: in Masking MC/DC, one does not care about the path to reach the condition for which we are proving the independence) with MCDC evaluations that reach SC in DECR:

$$MCDC(SC, DEC) = MCDC(SC, DECL)$$
$$\&\, paths\_to(\neg SC, DECL) \multimap MCDC(SC, DECR)$$

This time, the concatenation is expressed by $\multimap$ instead of $\otimes$. And the list of evaluations is constituted by $\&$ instead of $\oplus$. But that's the same idea. It's just that there is a duality between $\&$ and $\oplus$ (resp. $\otimes$ and $⅋$) and that MCDC(...) and paths_to(...) just happen to be on a different side of this duality. Otherwise, these operators work pretty much the same way.

5.2.3 Properties

We'll start with some simple syntactic properties:

- paths_to(...) can be normalized to match the following syntax:

$$< paths\_nf >::=< decision\_eval > ['\oplus' paths\_nf]$$

This is proven by induction; in the proof, and in all other proofs on this algorithm, only the case of BUILD_MCDC.Short_Circuit is non-trivial. This case is proven by using the distributivity of $\otimes$ over $\oplus$.

Example:

$paths\_to(T, (A \text{ and then } B) \text{ or else } C)$
$$=\{A = T, B = T\} \oplus (\{A = F\} \oplus \{A = T, B = F\}) \otimes \{C = T\}$$
$$=\{A = T, B = T\} \oplus \{A = F, C = T\} \oplus \{A = T, B = F, C = T\}$$

- MCDC(...) can be normalized to match the following syntax:

$$< mcdc\_nf >::=< mcdc\_vector > ['\&' < mcdc\_nf >]$$

This is proven as the previous property, by structural induction, focusing on expressions of the form $paths\_to(...) \multimap MCDC(...)$ in BUILD_MCDC.Short_Circuit. These expressions are reduced by:

  - dismembering paths_to(...), using right-distributivity of $\multimap$ over $\oplus$;

  - dismembering MCDC(...), using left-distributivity of $\multimap$ over $\&$;

  - merging evals of paths_to(...) with evals of MCDC(...), using associativity of $\mathfrak{N}$.

Example:

$MCDC(F, (A \text{ and then } B) \text{ or else } C)$

$$= (\{A = T, B = F\} \multimap f(B) \& \{A = F\} \multimap f(A)) \circ\!\!- \{C = F\}$$
$$\& (\{A = F\} \oplus \{A = T, B = F\}) \multimap (\{C = F\} \multimap f(C))$$

$$= (\{A = T, B = F\} \multimap f(B) \& \{A = F\} \multimap f(A)) \circ\!\!- \{C = F\}$$
$$\& \{A = F\} \multimap (\{C = F\} \multimap f(C))$$
$$\& \{A = T, B = F\} \multimap (\{C = F\} \multimap f(C))$$

$$= (\{A = T, B = F\} \multimap f(B)) \circ\!\!- \{C = F\}$$
$$\& (\{A = F\} \multimap f(A)) \circ\!\!- \{C = F\}$$
$$\& \{A = F\} \multimap (\{C = F\} \multimap f(C))$$
$$\& \{A = T, B = F\} \multimap (\{C = F\} \multimap f(C))$$

$$= \{A = T, B = F, C = F\} \multimap f(B)$$
$$\& \{A = F, C = F\} \multimap f(A)$$
$$\& \{A = F, C = F\} \multimap f(C)$$
$$\& \{A = T, B = F, C = F\} \multimap f(C))$$

We can even factorize common evaluations using left-distributivity of $\multimap$ over $\&$:

$$= \{A = T, B = F, C = F\} \multimap (f(B) \& f(C)) \& \{A = F, C = F\} \multimap (f(A) \& f(C))$$

For now on, the normal form for MCDC(...) will refer to the form where only one occurence of each decision evaluation can be found in the formula. That will be the next item.

- Unicity of normal forms

  Let us define normal forms as follow:

  DEFINITION 6 *paths_to(...) is said to be in normal form if it verifies the syntax of <paths_nf>.*

  MCDC(...) is said to be in normal form if it verifies the syntax of <mcdc_nf> and if only one occurence of each decision can be found in the formula.

  Existence of normal form has been proven previously. We have a property of unicity (modulo commutativity) that is a general property of these expressions in linear logic:

THEOREM 10 *Two expressions in normal forms are say to represent the same truth table if you can reduce one to the other using only the commutativity of ⊗, ⊕, &, ⅋.*

Then, if an expression returned by BUILD_MCDC can be reduced to two different expressions in normal forms, then these two normals forms represent the same truth table.

This is a general property of linear logic, that can be proven using its sequent calculus. We will not detail this proof here.

- Evaluations and normal forms

THEOREM 11 *Let us call evals(EXP) the set of decisions evals in the normal form of EXP. Let OUT be a boolean constant that represents the outcome of a decision. We have the following properties:*

- *evals(paths_to(OUT, Dec)) are exactly those that evaluates Dec to OUT;*
- *evals(paths_to(OUT, Dec)) = evals(MCDC(OUT, Dec)).*

The first part says that paths_to(OUT, Dec) does represent what it is meant to represent. That can be shown by structural induction; thanks to the distributivity of ⊗ over ⊕, it can be seen that:

$$evals((A \oplus B) \otimes C)) = evals(A) * evals(C) + evals(B) * evals(C)$$

∗ on the left being the cartesian product, + being the set union. After that, it is just a matter of checking that each paths to OUT is in evals(paths_to(OUT, Dec)); this offers no technical difficulty.

The second part says that all entries of the truth tables are annotated (which would eventually mean that all of them may be used to prove Masking MC/DC for a condition). That is one of the property that we introduced in our preliminary remarks.

This is also proven be induction, using distributivity properties of ⊸ . One may use the multiplicative disjunction and its distributity over & to make the proof simpler. One may notice how, in the key case BUILD_MCDC.Short_Circuit, the expression of paths_to(OUT, Dec) and MCDC(OUT, Dec) are similar. Again, no technical difficulty, just a mechanical check that each evals can be found in both expressions.

- MCDC vectors and normal forms

THEOREM 12 *Let us call vectors(EXP) the set of mcdc vectors in the normal form of EXP.*
Let OUT be a boolean constant that represents the outcome of a decision.
Then vectors(MCDC(OUT, Dec)) represents exactly the half of the annotated truth table of Dec in which Dec is evaluated to OUT.

Evals on the left of ⊸ operators, markers on right. This just says that MCDC(OUT, Dec) does represent what it is meant to be.

We already know that MCDC(OUT, Dec) contains all and only evals that such that the outcome of Dec is OUT: that was the second part of property 11. So, to prove the new property, we would just need to follow the markers.

To prove that the resulting set contains all MC/DC markers, we would use a property of masking MC/DC: if an evaluation can be used to prove t(c) for a condition c in Dec, then it restriction to a sub-decision (that contains c) proves t(c) for this sub-decision as well (that lemma can be proven fairly easily with a structural induction on influence trees). Same thing for f(c). With this property, we can show that the coverage that is computed cannot be incomplete, otherwise the coverage of one of the sub-decisions (DL or DR) would be incomplete, which would contradict the induction hypothesis.

5.2.4 Thoughts about implementation

From the properties of BUILD_MCDC, we can deduce a set of simplification for a real implementation.

- MCDC(OUT, Dec) can be represented by a list of evaluations colored with the t(c)/f(c) markers; such a structure would represent the normal form of the expression. This list is what we called the annotated truth table earlier.

- No point in maintaining two different structures for MCDC(OUT, Dec) and paths_to(OUT, Dec); the second one can be deduced from the first one by removing markers (property 11).

- In the context of our expressions, multiplicative operators are always distributive over the additive ones. So one may just understand $\otimes$ and $\multimap$ as a cartesian product over annotated truth tables. $\oplus$ and $\&$ can be thought of as union.

- We can have one entry per evaluation in the truth table, and several markers associated to them, thanks to left-distributivity of $\multimap$ over $\&$. Entries may be ordered to ease to factorize markers out: e.g. in (*A* `and then` *B*) `or else` *C*, evaluation

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{A=T, B=F, C=T} can be encoded in binary as 101. This gives us an index for ordering evaluations in the truth table. When in an evaluation a condition is not evaluated, the value of its digit in the index can be set to 0 (or 1, as long as the same convention is kept for all such evaluations). e.g. for {A=F, C=F}, index is 000.

- An implementation may not add some evaluations if it can prove (by an other mean) that coupling prevents them. This may avoid to build the whole truth table.

- In the modular implementation of BUILD_MCDC, when analyzing a sub-decision that contains all conditions on which we focus, it is possible to drop any evaluation that does not execute one of these conditions.

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