

# The BDD Domain and Applications to Constraint Satisfaction

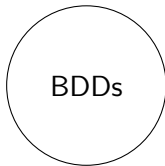
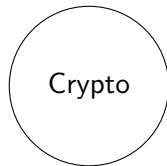
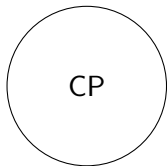
Mathieu Vavrille

ENS de Lyon

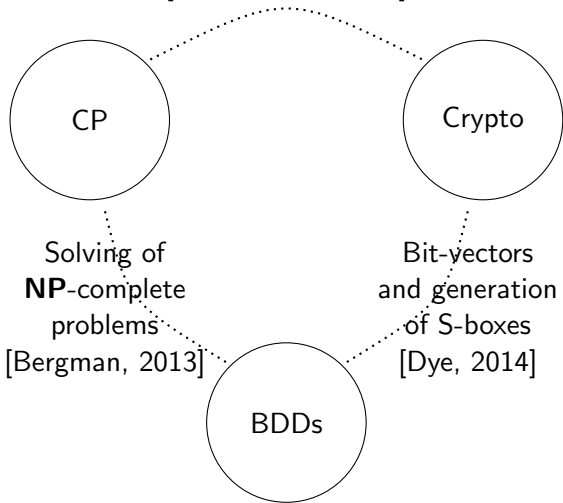
M2 internship supervised by Charlotte Truchet  
in the University of Nantes (LS2N, TASC team)

July 03, 2019

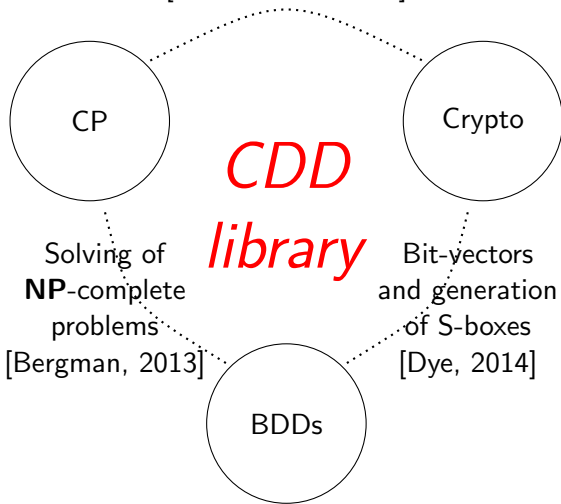




Analysis of attacks  
[Minier et al., 2014]



Analysis of attacks  
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# Constraint programming

- ◇ Define a problem declaratively with variables and constraints, and find a solution.
- ◇ A field between Artificial Intelligence and Operations Research.
- ◇ Many applications on solving NP-hard problems in planning, logistics, arts, computational sustainability and, very recently, cryptography.

## Constraint satisfaction problem (CSP)

A CSP is the data  $(X, \mathcal{D}, C)$  where  $X$  is a set of variables,  $\mathcal{D}$  is a function from variables to their domains (values they can take),  $C$  is a set of constraints defined in:

- ◇ intension: high level language to express the constraint, or,
- ◇ extension: list of allowed solutions (called table constraints).

# Constraint solving

## Algorithm: propagate and search

Find the solutions in two steps:

- ◇ propagation: delete values from the domains of some variables that are inconsistent with respect to some constraint
- ◇ search: enumerate the domains to find the solutions.

Example of propagation:

$$\begin{aligned}x, y, z &\in \{1, 2, 3\} \\ x + y &\leq z\end{aligned}$$

We find that the value  $z = 1$  is not consistent (not possible).

# Context

## A limitation of CP solvers

- ◇ Constraint with divisions or modulo
- ◇ Constraints on bit-vectors (vectors of bits)

## Goal

Investigate the use of another structure on these constraints: BDDs

## Existing domains

- ◇ BDD [Bryant, 1986]: a data structure initially used for boolean functions
- ◇ MDD [Andersen et al., 2007]: the consistency
- ◇ bit-vector domain [Michel and Van Hentenryck, 2012]: the computations

# Internship contribution <sup>1</sup>

## Contribution : CDD library

A library based on a BDD structure to represent domains in a constraint-friendly way,

- ◇ with two types of BDDs: unrestricted and limited-width,
- ◇ implemented in OCaml within the AbSolute constraint solver,
- ◇ experimented on a cryptography problem (AES).

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<sup>1</sup><https://github.com/MathieuVavrille/cdd>



```
module type CDD = sig
  type bdd
  val bdd_of : bdd -> bdd -> bdd
  (*** Set operations ***)
  val intersection : bdd -> bdd -> bdd
  (* and others: difference, union, etc *)
  (*** Bitwise operations ***)
  val xor : bdd -> bdd -> bdd
  (* and others: and, or, not, etc *)
  (*** BDD manipulations ***)
  val prefix : bdd -> int -> bdd
  val suffix : bdd -> int -> Bddset.t
  val concat : bdd -> bdd -> bdd
  (*** Solving techniques ***)
  val multiple_mdd_consistency : bdd -> Bddset.t -> bdd
  val refined_consistency : bdd -> bdd -> int -> bdd
  val split : bdd -> (bdd * bdd)
```

## 1 Introduction

## 2 Definitions

- Reduced BDDs
- Implementation

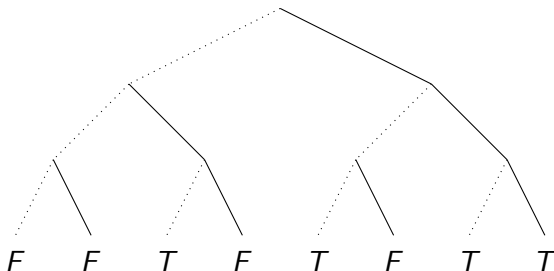
## 3 Computations

## 4 Propagation

## 5 Test case: AES

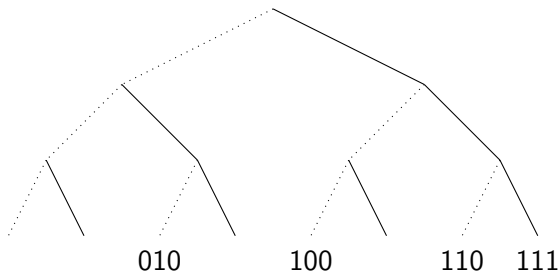
# Representation of bit-vectors

- ◇ Bit-vectors are represented by the paths from the root to the leaves  $T$ .
- ◇ The set represented is  $\{010, 100, 110, 111\} \Leftrightarrow \{2, 4, 6, 7\}$



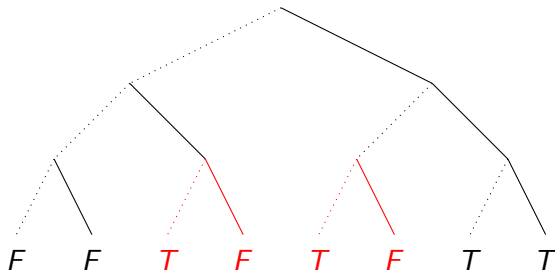
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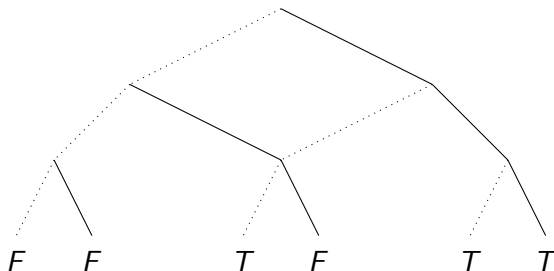
## BDD: Reduction of binary tree

We now want a smaller representation



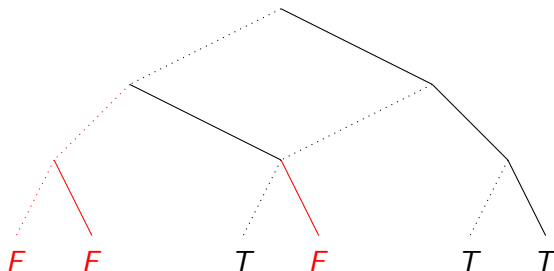
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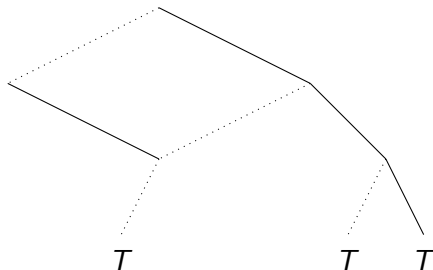
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# BDD: Reduction of binary tree

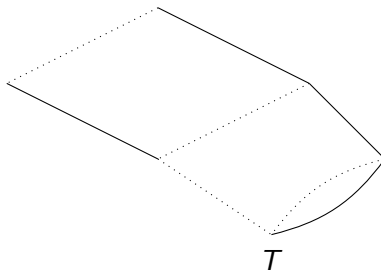
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# BDD: Reduction of binary tree

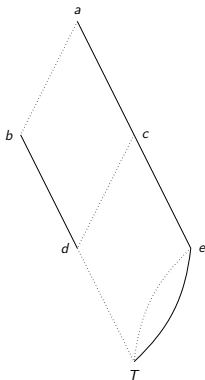
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# Implementation

## Global hashtable

- ◇ keys are (0-child, 1-child), value is the BDD represented
- ◇ one entry  $(F, F)$  with value  $F$ .



name in the tree	ref in the table	key in the table	value in the table
-	$h_0$	$F, F$	$F$
$d$	$h_1$	$T, F$	$T$
$b$	$h_2$	$F, h_1$	$h_1$
$e$	$h_3$	$T, T$	$T$
$c$	$h_4$	$h_1, h_3$	$h_1 \vee h_3$
$a$	$h_5$	$h_2, h_4$	$h_2 \vee h_4$

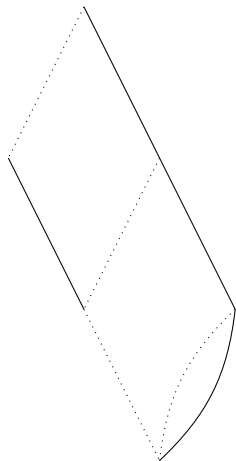
# Implementation

```
1 Function BDD_OF( $a, b$ ) returns the BDD Node( $a, b$ ) where we shared  
   equivalent subtrees  
2   if  $(a, b) \in \text{global\_hash}$  then  
3     return FIND( $\text{global\_hash}, (a, b)$ )  
4   else  
5      $\text{new\_ref} \leftarrow \text{Node}(a, b)$   
6     ADD_TO_HASH( $\text{global\_hash}, (a, b), \text{new\_ref}$ )  
7     return  $\text{new\_ref}$   
8   end
```

If we always use this algorithm, all the BDDs will be reduced. This is the core of the implementation.

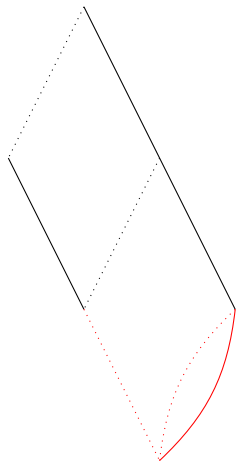
- 1 Introduction
- 2 Definitions
- 3 Computations**
- 4 Propagation
- 5 Test case: AES

# Extraction of prefixes



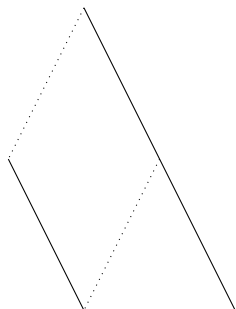
- Prefixes of size  $k$  are the first  $k$  bits of the BDD  $B$ .

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# Extraction of prefixes



- ◇ Prefixes of size  $k$  are the first  $k$  bits of the BDD  $B$ .
- ◇ The set  $Pref(B, k)$  is  $\{01, 10, 11\}$
- ◇ We have that  $\{2, 4, 6, 7\} // 2 = \{1, 2, 3\}$

1 Introduction

2 Definitions

3 Computations

4 **Propagation**

- General propagation
- Div constraint
- Table constraint

5 Test case: AES



# General propagation

## And constraint

Let the constraint  $X \wedge Y = Z$ . We compute

$$And^{-1}(\mathcal{D}(Z), \mathcal{D}(X)) = \{y \mid x \wedge y = z, x \in \mathcal{D}(X), y \in \mathcal{D}(Y), z \in \mathcal{D}(Z)\}$$

Then the propagation on  $Y$  is simply

$$\mathcal{D}(Y) \leftarrow \mathcal{D}(Y) \cap And^{-1}(\mathcal{D}(Z), \mathcal{D}(X))$$

## What do we need ?

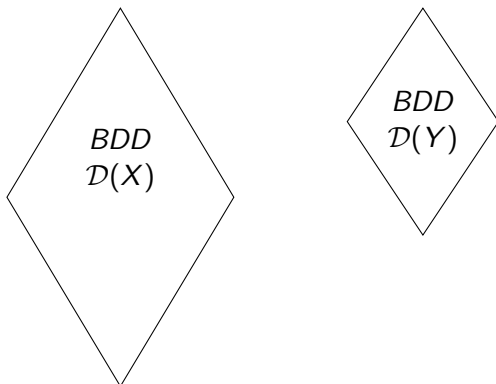
- ◇ Bitwise operations
- ◇ Intersection

Div constraint:  $X // 2^k = Y$

Condition on the depth:  $depth(X) = k + depth(Y)$

Propagation on  $Y$ :  $\mathcal{D}(Y) \leftarrow \mathcal{D}(Y) \cap Pref(\mathcal{D}(X), depth(X) - k)$

Propagation on  $X$ :  $\mathcal{D}(X) \leftarrow \mathcal{D}(X) \cap \mathcal{D}(Y) \cdot COMPLETE(k)$

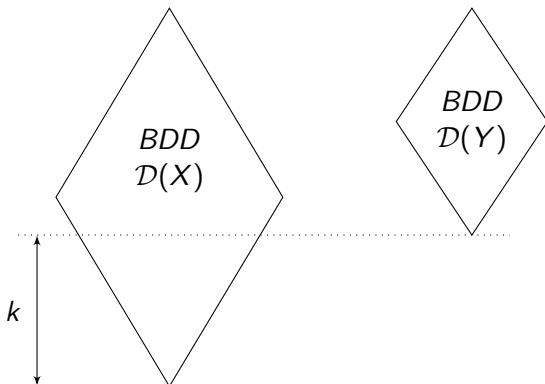


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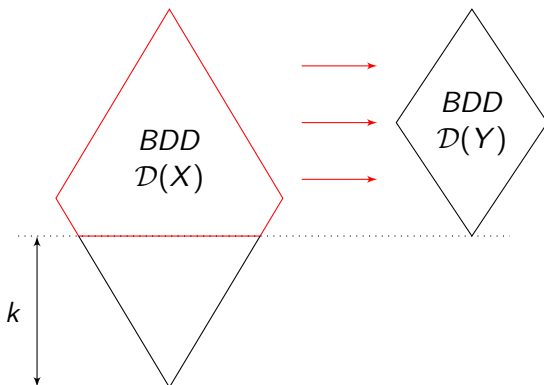


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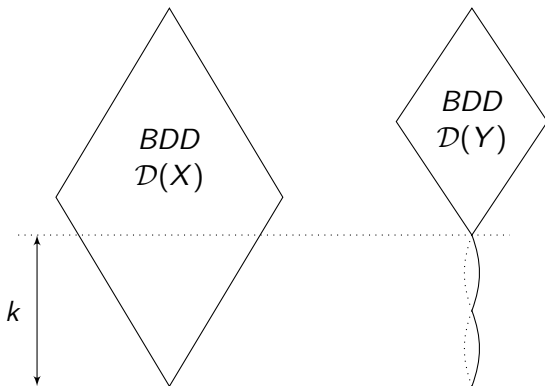


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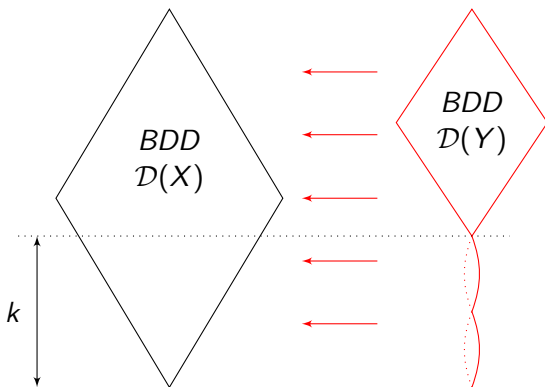


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# Table constraint

## Table constraint

A *table constraint* is a constraint defined on  $n$  variables in extension by the set of allowed  $n$ -uplet.

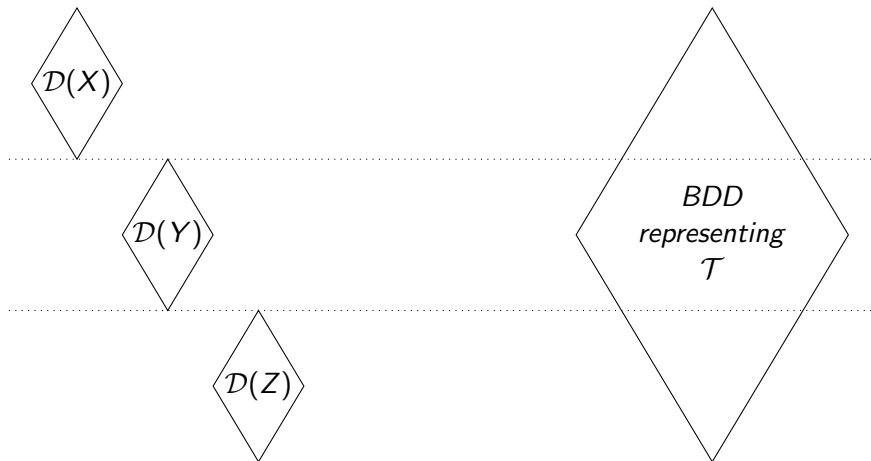
## Transformation to BDD

Let  $\mathcal{T}$  the set representing a table constraint on 3 variables. Let

$$S = \{b^1 \cdot b^2 \cdot b^3 \mid (b^1, b^2, b^3) \in \mathcal{T}\}$$

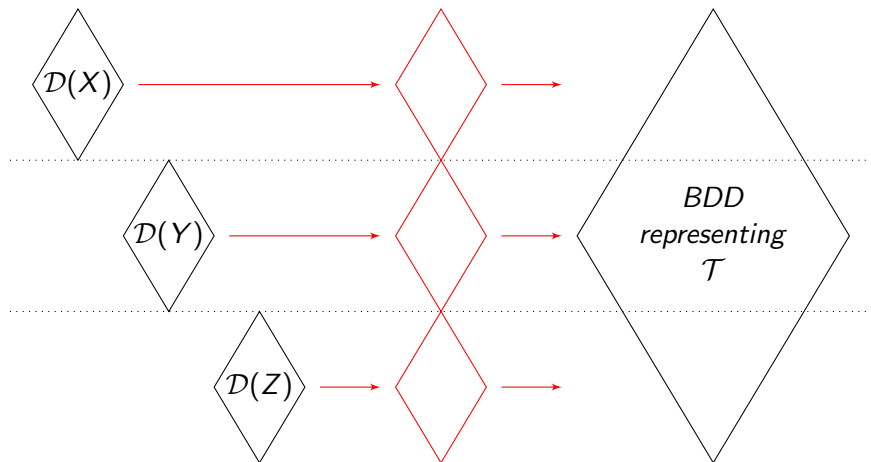
$S$  is a set of bit-vectors, thus it can be represented by a BDD.

## Table constraint: $Table(X, Y, Z, \mathcal{T})$

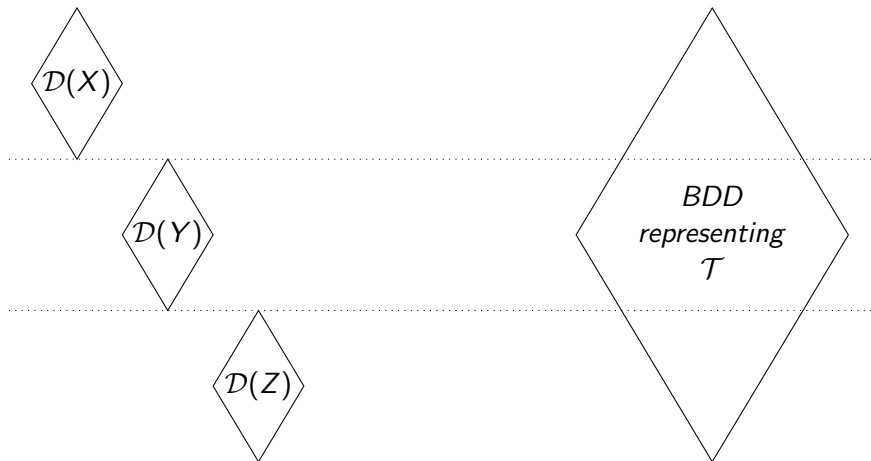




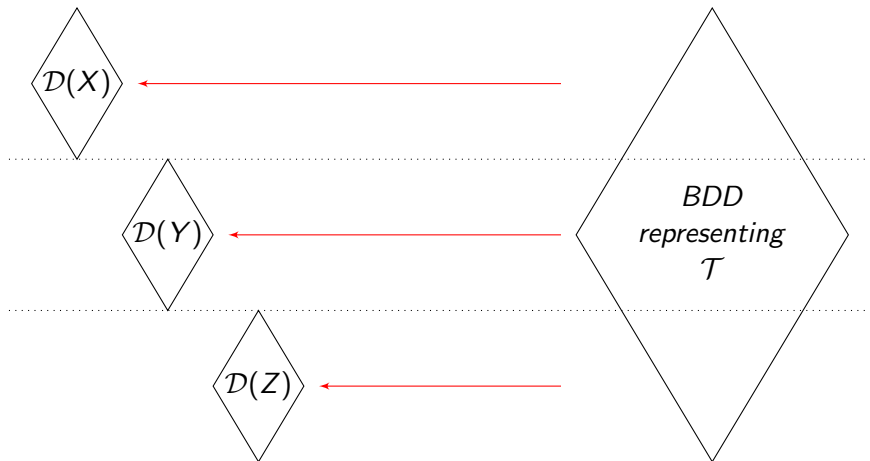
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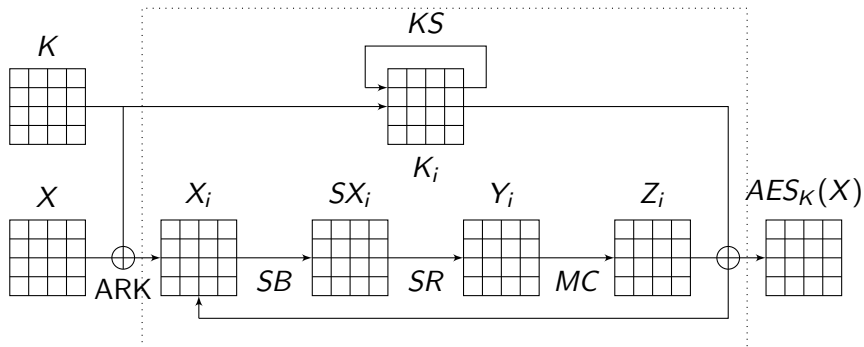
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## Differential analysis

Start with messages  $X, X'$ , and keys  $K, K'$ . Compute  $E_K(X)$  and  $E_{K'}(X')$ . What is the probability to get  $E_K(X) \oplus E_{K'}(X')$  knowing  $X \oplus X'$  and  $K \oplus K'$  ?

# Constraint program solving

## Translation

- Variables:  $\delta X = X \oplus X', \delta K = K \oplus K'$ .
- Constraints: representation of the computations (Xor, mix column, S-box, ...)

Representation from [Minier et al., 2014, Gerault et al., 2016, Gerault et al., 2017, Gerault et al., 2018]

## Phase 1 (SAT solver)

Abstract the bytes with variables

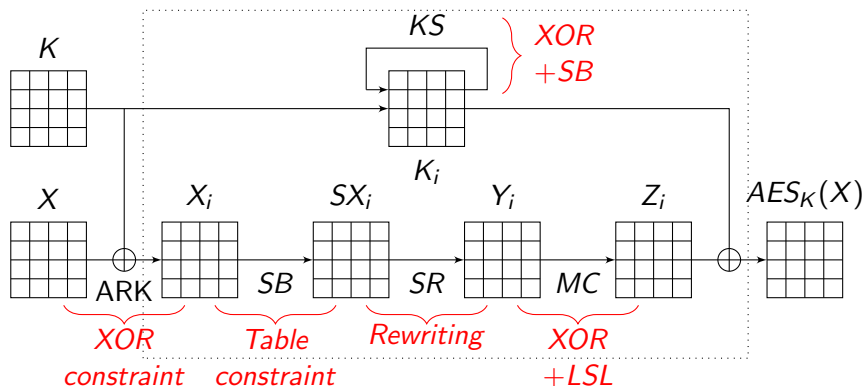
$$\Delta X = 0 \Leftrightarrow \delta X = 0^8$$

$$\Delta X = 1 \Leftrightarrow \delta X \in [1, 255]$$

## Phase 2 (Constraint solver)

Find the real values of the bytes  $\delta X$  knowing  $\Delta X$

# Can we solve it with BDDs ?



# Conclusion, what I didn't talk about

Design and implementation of a BDD library for CP.

- ◇ Limited-width-BDDs,
- ◇ Bitwise propagators,
- ◇ Set operations

Theoretical work on BDDs

- ◇ Definition of merge value
- ◇ Analysis of the width
- ◇ Equivalence between domains




# Future work

## On BDDs:

- ◇ Integer domain: algebraic computations
- ◇ Global constraints: `all_different`, `global_cardinality`, `regular`
- ◇ Comparison with other domains (bit-vector domain)
- ◇ Improvement of heuristics
- ◇ General constraint  $X \equiv Y \pmod{c}$  (not  $c = 2^k$ )

About cryptography: project ANR Decrypt about CP and cryptography

- ◇ Analysis of other protocols
- ◇ Usage of BDDs to design protocols



Thank you  
for your attention



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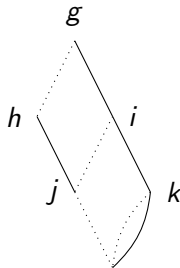
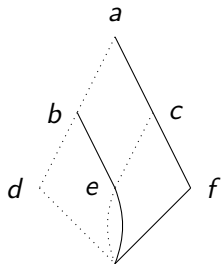


Minier, M., Solnon, C., and Reboul, J. (2014).

Solving a symmetric key cryptographic problem with constraint programming.

In *13th International Workshop on Constraint Modelling and Reformulation (ModRef)*, in conjunction with CP, volume 14, pages 1–13.

## Example: intersection



Intersections computed recursively:  $M \cap M' = a \cap g$

◇  $b \cap h$

◇  $d \cap F$

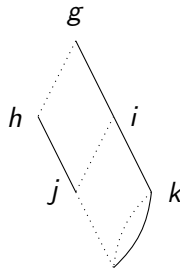
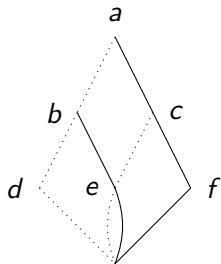
◇  $e \cap j$

◇  $c \cap i$

◇  $e \cap j$

◇  $f \cap k$

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◇  $d \cap F$

◇  $e \cap j$

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◇  $f \cap k$

# Implementation of intersection

## Idea

Use caching to compute only once the functions

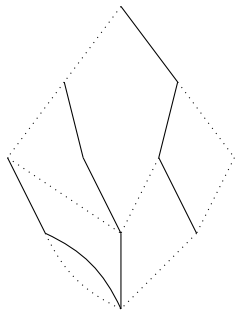
```
1 Function INTER( $B_1, B_2$ ) returns  $B_1 \cap B_2$ 
2   if ( $B_1, B_2$ )  $\in$  inter_hash then
3     return FIND(inter_hash, ( $B_1, B_2$ ))
4   else
5     if  $B_1 = F$  or  $B_2 = F$  then
6       result  $\leftarrow F$ 
7     else if  $B_1 = T$  or  $B_2 = T$  then
8       result  $\leftarrow T$ 
9     else
10      Let  $a_1, b_1$  such that  $B_1 = \text{Node}(a_1, b_1)$ 
11      Let  $a_2, b_2$  such that  $B_2 = \text{Node}(a_2, b_2)$ 
12      result  $\leftarrow$  BDD_OF(INTER( $a_1, a_2$ ), INTER( $b_1, b_2$ ))
13    end
14    ADD_TO_HASH(inter_hash, ( $B_1, B_2$ ), result)
15    return result
16  end
```



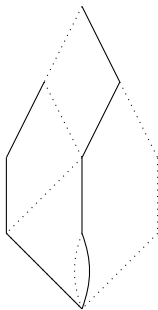
# Limited-width-BDDs

## Goal

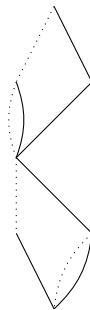
Having a polynomial size representation (depending on the depth)



(a)  
4-limited-width-BDD  
representing  $S$



(b)  
3-limited-width-BDD:  
 $S \cup \{15\}$

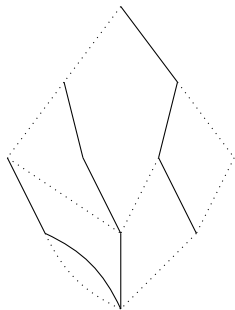


(c)  
2-limited-width-BDD:  
 $S \cup \{5, 6, 9, 15\}$

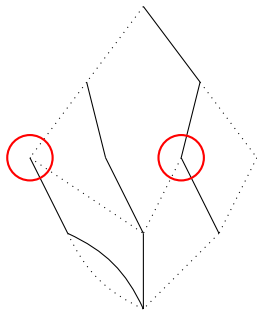
Figure 3: Representation of the set

$$S = \{1, 2, 3, 7, 8, 13, 14\} = \{0001, 0010, 0011, 0111, 1000, 1101, 1110\}$$

# How to generate them ?

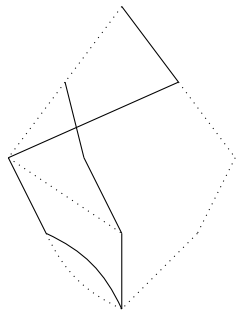


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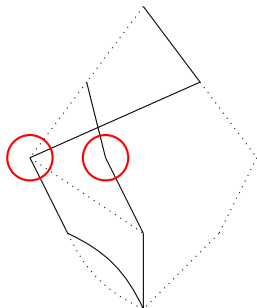
- ◇ Merging paths  $00 \cdot \{01, 10, 11\}$  and  $11 \cdot \{01, 10\}$

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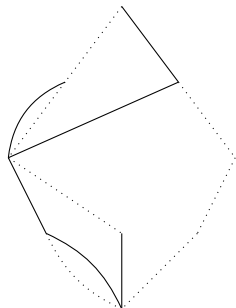
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  - ◇ Adds the bit-vector 1111

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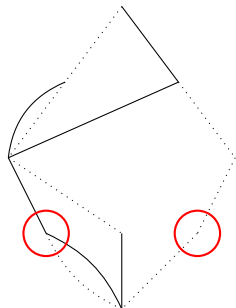
- ◇ Merging paths  $00 \cdot \{01, 10, 11\}$  and  $11 \cdot \{01, 10\}$ 
  - ◇ Adds the bit-vector 1111
- ◇ Merging paths  $00 \cdot \{01, 10, 11\}$  and  $01 \cdot 11$

# How to generate them ?



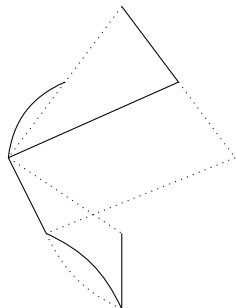
- ◇ Merging paths  $00 \cdot \{01, 10, 11\}$  and  $11 \cdot \{01, 10\}$ 
  - ◇ Adds the bit-vector 1111
- ◇ Merging paths  $00 \cdot \{01, 10, 11\}$  and  $01 \cdot 11$ 
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- ◇ Merging paths  $001 \cdot \{0, 1\}$  and  $100 \cdot 0$

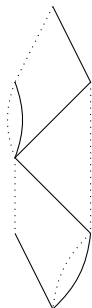
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- ◇ Merging paths  $001 \cdot \{0, 1\}$  and  $100 \cdot 0$ 
  - ◇ Adds the bit-vector 1001

# What is the over-approximation

## Definition (Merge Value)

The *merge value* of two nodes ( $mv(u, v)$ ) is the number of bit-vectors added when merging the two nodes  $u$  and  $v$ .

## Theorem

*Let  $B$  be a BDD of root  $r$ , and  $u$  and  $v$  two nodes of the same layer. Let  $p_u$  (resp  $p_v$ ) the number of paths that go from  $r$  to  $u$  (resp.  $v$ ).*

*The merge value is then*

$$\begin{aligned} mv(u, v) &= p_u |\gamma(v) \setminus \gamma(u)| + p_v |\gamma(u) \setminus \gamma(v)| \\ &= p_u |\gamma(v)| + p_v |\gamma(u)| - (p_u + p_v) |\gamma(u) \cap \gamma(v)| \end{aligned}$$

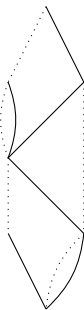
# Consistency for limited BDDs

## Issue

The intersection may increase the width

## Solution from MDD-consistency

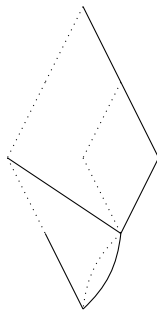
Consistency of  $M$  with respect to  $M'$ : delete edges of  $M$  that are not in paths of  $M'$ .



(a) BDD  $M$



(b) BDD  $M'$



(c) BDD  $M \cap M'$

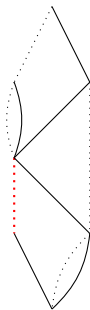
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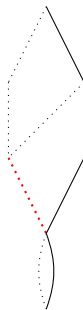
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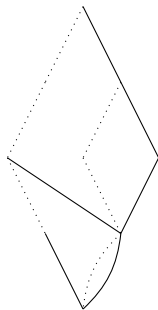
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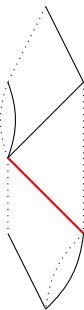
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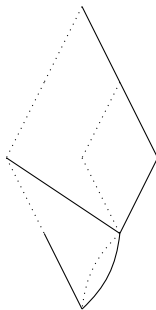
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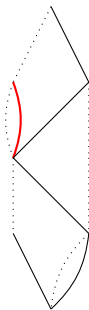
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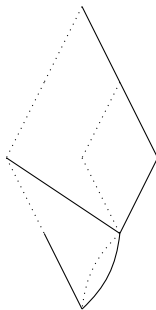
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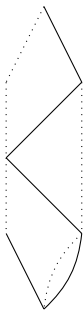
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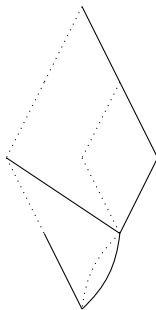
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(c) BDD  $M \cap M'$

# Multiple consistency

## Improvement

We often need to do the consistency of  $M$  w.r.t  $\bigcup_i M_i$

## Algorithm

Simply check if the edges are in paths of at least one BDD in the set

## Cost

If the BDDs come from the same original BDD, we share a lot of computations.



# Refining

## Improvement

The width will only decrease with this consistency. This limits the expressiveness.

## Solution 1

Choose nodes to split: have twice the same node in the BDD, and then do a propagation that will change one and not the other.

→ Not really great with a functional implementation.

## Solution 2

Increase the width during consistency.

→ Great compromise between computing the intersection and doing the consistency