



# Specification and Verification

## Lecture 5: Binary decision diagrams

Guillermo A. Pérez

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# BDDs in short

## What are they?

A graph-based data structure for Boolean functions and sets

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## What are they?

A graph-based data structure for Boolean functions and sets

## Why use them?

They provide a **canonical**, sometimes space-efficient, representation of Boolean functions; allow to implement logical operations  $\vee, \wedge, \neg$  using **efficient** graph transformations.

# References

## Main references

- **Binary Decision Diagrams.** Randal E. Bryant. 2018. In Handbook of Model Checking. Springer, 2018.
- **The Art of Computer Programming, Vol. 4, Pre-Fascicle 1B: Binary Decision Diagrams.** Don Knuth. Addison-Wesley, 2008.

# Required and target competences

## What tools do we need?

Discrete mathematics, data abstraction and structures; algorithms and complexity

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- practice: manipulate BDD-represented functions & sets

## Why?

May be useful for professional work on circuits/hardware and

- data structures and graph algos
- artificial intelligence (knowledge representation)
- specification and verification

# Outline

1. Recap on Boolean functions and sets
2. Example: from characteristic functions to BDDs
3. History
4. Motivation and applications
5. Reduced ordered BDDs
6. BDD operations
7. Shared table and other implementation tricks
8. Conclusions



# Boolean functions

## Definition (Boolean or switching functions)

A Boolean function  $f$  is of the form  $\mathbb{B}^k \rightarrow \mathbb{B}$ , where  $\mathbb{B} = \{0, 1\}$  and  $k \in \mathbb{N}$  is the arity of  $f$ .

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## Example

Let us define the function  $g : \mathbb{B}^2 \rightarrow \mathbb{B}$  such that  $g(0,0) = 0$ ,  $g(0,1) = 1$ ,  $g(1,0) = 1$ , and  $g(1,1) = 0$ . Do you recognize the function?

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$x_1$	$x_2$	XOR
0	0	0
0	1	1
1	0	1
1	1	0

# Defining Boolean functions

## How can we define a Boolean function?

- They are expressively equivalent to
  - truth tables
  - propositional formulas
- They can encode sets via their characteristic function
- They can be represented as
  - And-inverter graphs
  - Negation normal forms
  - Propositional directed acyclic graphs
  - Binary decision diagrams

# Propositional formulas & Bool functions

## Propositional formulas

For a set  $P = \{p_1, \dots, p_k\}$  of **propositions**, a propositional formula  $\varphi$  over  $P$  is constructed using the logical connectives  $\vee, \wedge, \neg$ . E.g.  $(p_1 \wedge \neg p_2) \vee (\neg p_1 \wedge p_2)$ .

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- $\varphi$  induces a Boolean function  $f_\varphi$  that maps  $(x_1, \dots, x_k)$  to the truth value of  $\varphi$  under the truth-value assignment  $p_1 = x_1, \dots, p_k = x_k$ .

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## Completeness

Every Boolean function  $g$  has a corresponding propositional formula  $\varphi$ , i.e.  $g = f_\varphi$  for some  $\varphi$ . (Graphical proof via truth tables, see board).

# Subsets and Boolean functions

## Characteristic functions of subsets

Let  $S = \{s_1, \dots, s_k\}$  be a finite set of elements. Every subset  $P \subseteq S$  induces a function  $\chi_P : S \rightarrow \mathbb{B}$

$$\chi_P(s) = \begin{cases} 1 & \text{if } s \in P \\ 0 & \text{otherwise.} \end{cases}$$



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## Binary encoding of the set

Let  $\ell = \lceil \log_2 k \rceil$  and consider a mapping  $\beta : \mathbb{B}^\ell \rightarrow S \cup \{\perp\}$  such that  $\beta|_{\beta^{-1}(S)}$  is injective and surjective. Every subset  $P \subseteq S$  induces a function  $\chi_P^\beta : \mathbb{B}^\ell \rightarrow \mathbb{B}$

$$\chi_P^\beta(x_1, \dots, x_\ell) = \begin{cases} 1 & \text{if } \beta(x_1, \dots, x_\ell) \in P \\ 0 & \text{otherwise.} \end{cases}$$

## Example: binary encoding of a set

1.  $S = \{\text{Belgium, Panama, Tunisia, England, Honduras}\}$

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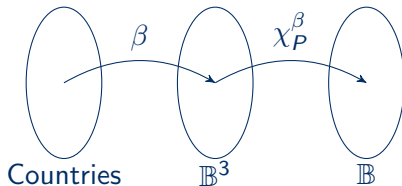
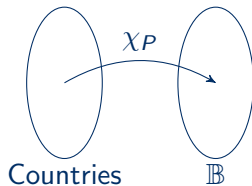
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## Example: binary encoding of a set

1.  $S = \{\text{Belgium, Panama, Tunisia, England, Honduras}\}$
2.  $P = \{\text{Belgium, Panama, Tunisia, England}\}$
3. We choose **some**  $\beta$ , for example:

$x_1$	$x_2$	$x_3$	$s \in S \cup \{\perp\}$
0	0	0	Belgium
0	0	1	Panama
0	1	0	$\perp$
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2.  $P = \{\text{Belgium, Panama, Tunisia, England}\}$

$$\chi_P^\beta(0, 0, 0) = 1$$

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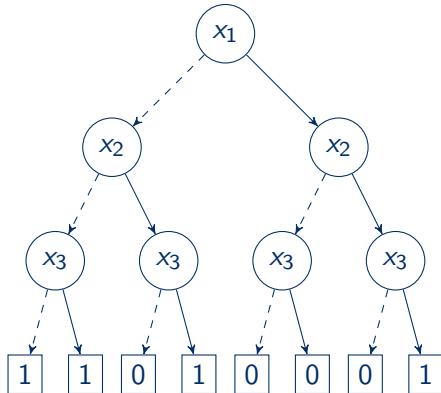
and all other combinations map to 0. Also,  $\chi_P^\beta = f_\varphi$  with

$$\varphi = (\neg x_1 \wedge \neg x_2) \vee (x_2 \wedge x_3).$$

## Example: truth tables to dec. trees

$x_1$	$x_2$	$x_3$	$s \in S \cup \{\perp\}$	$\chi_P^\beta$
0	0	0	Belgium	1
0	0	1	Panama	1
0	1	0	$\perp$	0
0	1	1	Tunisia	1
1	0	0	$\perp$	0
1	0	1	Honduras	0
1	1	0	$\perp$	0
1	1	1	England	1

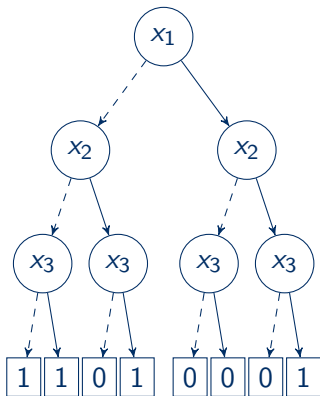
- dashed: assignment  $x_1 = 0$ ; solid arrows,  $x_1 = 1$
- leaves: value of  $\chi_P^\beta$





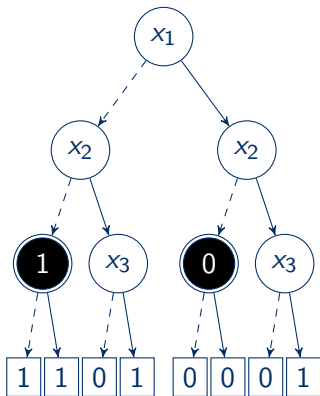
# Example: decision trees to ROBDDs

1. Identify **useless sub-graphs**: those whose children are equivalent



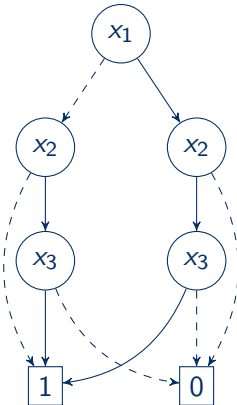
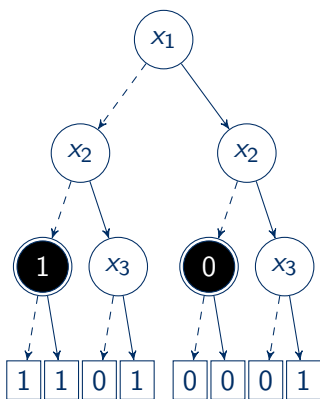
# Example: decision trees to ROBDDs

1. Identify **useless sub-graphs**: those whose children are equivalent
2. Merge equivalent leaves and remove useless sub-trees



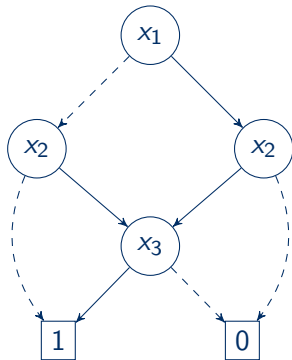
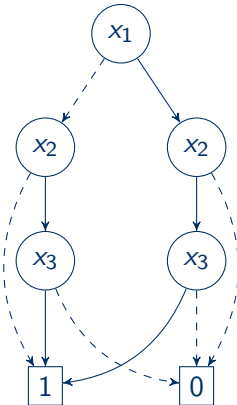
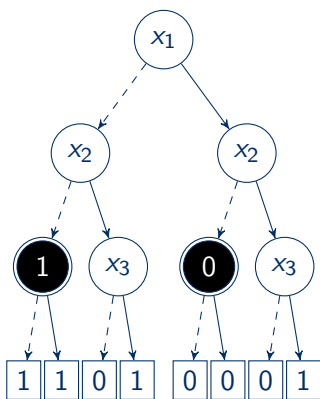
# Example: decision trees to ROBDDs

1. Identify **useless sub-graphs**: those whose children are equivalent
2. Merge equivalent leaves and remove useless sub-trees
3. Merge **sub-graph-equivalent inner vertices**



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## Exercise: from a subset to its ROBDD

- $S = \{\text{Brussels, Antwerp, Leuven, Amsterdam}\}$
- We choose  $\beta$  as follows

$x_1$	$x_2$	$s \in S \cup \{\perp\}$
0	0	Brussels
0	1	Antwerp
1	0	Amsterdam
1	1	Leuven

- $P = \{\text{Brussels, Leuven, Antwerp}\}$
- What is the ROBDD for  $\chi_P^\beta$ ?

# History

## From Boole to Lee

- The Shannon expansion (due to Boole) is the basic theoretical idea behind BDDs

$$f(x_1, \dots, x_k) = (\neg x_i \wedge f|_{x_i \leftarrow 0}) \vee (x_i \wedge f|_{x_i \leftarrow 1})$$

for any  $x_i$ .

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## BDDs become efficient

- Randal E. Bryant introduces sharing (for compression) and order fixing (for canonicity) [1986–1990]
- Don Knuth: “one of the only really fundamental data structures that came out in the last ... years”

# Motivation and applications

## Any $f(x_1, \dots, x_k)$ for which you can build a BDD

- we can evaluate  $f(x)$  in at most  $k$  steps,
- we can find the (lexicographically) smallest  $x$  such that  $f(x) = 1$ ,
- we can count and/or list the number of solutions to  $f(x) = 1$ ,
- ...

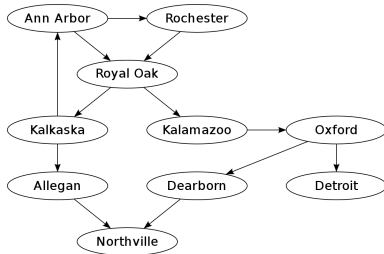
## Applications

- Computer-aided design of circuits
- Formal verification
- Bayesian reasoning
- ...



# Sample application: the geography game

- Alternate naming cities, no repetition allowed
- The next city's name should start with the same letter as the last letter of the previously named city



- If I start, can I win or make sure the game lasts for at least three rounds (i.e. I play at least 2 cities)?

# ROBDDs: the formal definition (1/4)

## Reduced ordered BDDs

A (RO)BDD is a rooted directed acyclic graph (DAG) with **inner decision vertices** and **constant-value leaves** (with values 0 or 1).

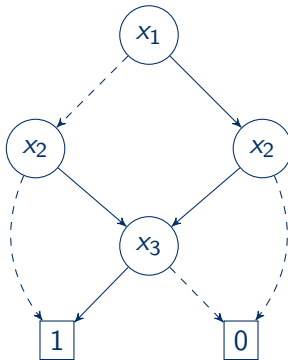
- Every decision vertex  $v$  is labelled by a Boolean variable  $\text{VAR}(v)$  and has two children:  $\text{LO}(v)$  and  $\text{HI}(v)$  respectively representing a truth-value assignment of 0 or 1 for  $v$ .

# ROBDDs: the formal definition (1/4)

## ROBDDS, continued

- All paths  $v_1 \dots v_\ell$  are such that  $\text{VAR}(v_i) < \text{VAR}(v_j)$  for all  $i < j$  (for **some ordering** of the variables).
- There are at most two constant-value leaves with distinct values.
- No two vertices  $u, v$  induce **isomorphic graphs** ( $\text{VAR}(u) = \text{VAR}(v)$ ,  $\text{LO}(u) = \text{LO}(v)$ , and  $\text{HI}(u) = \text{HI}(v)$ ) and
- all vertices  $v$  have **non-isomorphic children**, i.e.  $\text{LO}(v) \neq \text{HI}(v)$ .

## ROBDDs: an example (2/4)



■ for the order  $x_1 < x_2 < x_3$

# ROBDDs: from BDDs to functions (3/4)

## Given a BDD, what is its function/formula?

For a BDD with variables  $x_1, \dots, x_k$  and root  $u$ , the Boolean function  $f_u(x_1, \dots, x_k)$  induced by it is defined as follows. For all vertices  $v$

- if  $\text{VAR}(v) = 0$  then  $f_v(x_1, \dots, x_k) = 0$ ,
- if  $\text{VAR}(v) = 1$  then  $f_v(x_1, \dots, x_k) = 1$ ,
- otherwise, by the **Shannon expansion**,

$$f_v(x_1, \dots, x_k) = (\neg \text{VAR}(v) \wedge f_{\text{LO}(v)}) \vee (\text{VAR}(v) \wedge f_{\text{HI}(v)}) .$$

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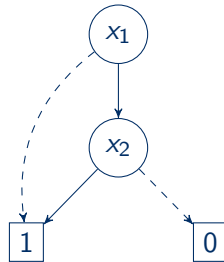
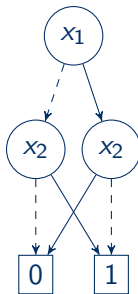
$$f_v(x_1, \dots, x_k) = (\neg \text{VAR}(v) \wedge f_{\text{LO}(v)}) \vee (\text{VAR}(v) \wedge f_{\text{HI}(v)}) .$$

## Theorem (Canonicity [Bryant '86])

For all Boolean functions  $f(x_1, \dots, x_k)$ , for every ordering of  $x_1, \dots, x_k$ , there is a **unique** BDD with root  $u$  such that  $f = f_u$ .

# Exercise: from BDDs to functions

- What are the formulas for the following BDDs?



# ROBDDs: on the ordering (4/4)

## What order should one use?

Depending on the ordering of  $x_1, \dots, x_k$ , the size of the BDD representing  $f(x_1, \dots, x_k)$  may vary **exponentially**!

- Deciding whether a given order is size-minimal is an NP-complete problem.
- In practice, heuristics are used to **dynamically choose** a “good” ordering.



# ROBDDs: on the ordering (4/4)

## Example

$x_1$	$x_2$	$x_3$	$f$
0	0	0	0
0	0	1	0
0	1	0	1
0	1	1	1
1	0	0	0
1	0	1	1
1	1	0	1
1	1	1	1

$x_2$	$x_1$	$x_3$	$f$
0	0	0	0
0	0	1	0
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# ROBDDs: on the ordering (4/4)

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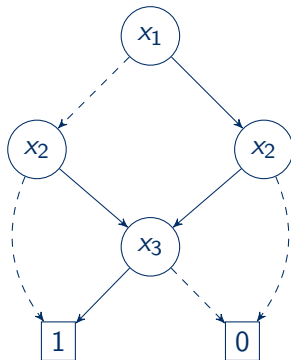
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0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	1
1	0	1	1
1	1	0	1
1	1	1	1

- with large BDDs the effect is more evident

## Exercise: a different order

- construct the ROBDD for Belgium's World Cup group using the order  $x_2 < x_3 < x_1$  and compare against the one we had before

$x_1$	$x_2$	$x_3$	$s \in S \cup \{\perp\}$
0	0	0	Belgium
0	0	1	Panama
0	1	0	$\perp$
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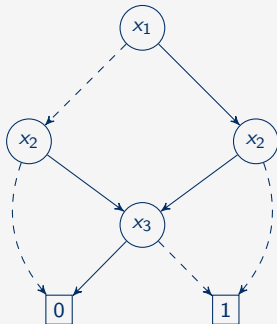
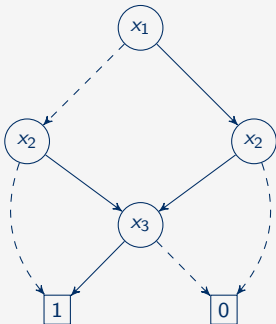
# BDD operations: logical negation (1/4)

## Negating a BDD

Given a BDD representing the Boolean function  $f(x_1, \dots, x_k)$ , we can obtain a BDD for  $g(x_1, \dots, x_k) = \neg f(x_1, \dots, x_k)$  by simply replacing the 0 and 1 constant-value leaves.

# BDD operations: logical negation (1/4)

## Example



## BDD operations: $\vee, \wedge$ (2/4)

### Divide and conquer using Shannon's expansion

For  $\text{op} \in \{\vee, \wedge\}$  we have that  $f(x_1, \dots, x_k) \text{ op } g(x_1, \dots, x_k)$  is equivalent to the following for all  $1 \leq i \leq k$

$$\left( \neg x_i \wedge \left( f|_{x_i \leftarrow 0} \text{ op } g|_{x_i \leftarrow 0} \right) \right) \vee \left( x_i \wedge \left( f|_{x_i \leftarrow 1} \text{ op } g|_{x_i \leftarrow 1} \right) \right).$$

# BDD operations: $\vee, \wedge$ (2/4)

## More graphically...

We do a DFS on both BDDs from the root stepping lexicographically and

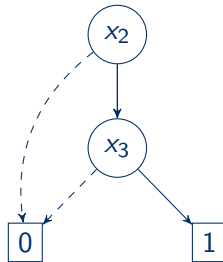
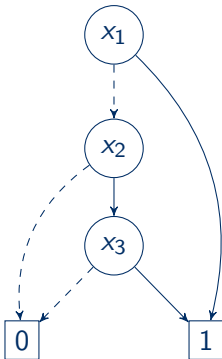
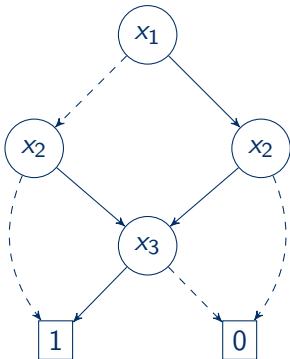
- **synchronously** if the current vertices  $u, v$  have the same label, i.e.  $\text{VAR}(u) = \text{VAR}(v)$ ,
- **asynchronously** from  $v$  if the current vertices  $u, v$  are such that  $\text{VAR}(u) > \text{VAR}(v)$ .<sup>a</sup>

For every vertex-pair whose children  $s, t$  have been visited, we add a new vertex  $r$  with  $\text{VAR}(r) = \min(\text{VAR}(u), \text{VAR}(v))$  and  $\text{LO}(r) = s, \text{HI}(r) = t$  to the resulting BDD.

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<sup>a</sup>We suppose  $\text{VAR}(t) \geq \text{VAR}(v)$  for all constant-value vertices  $t$  and all vertices  $v$ .

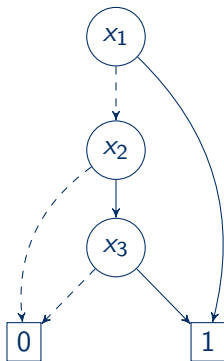
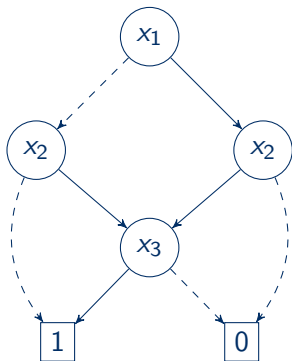
## BDD operations: example (3/4)



■ conjuncting the first two BDDs yields the third (from left to right)



## Exercise: disjunction of BDDs



?

- the disjunction of the first two BDDs yields. . .

# BDD operations: others (4/4)

## Useful classical operations

- **existential quantification** can be implemented using the equivalence

$$\exists x_1 : f(x_1, \dots, x_k) = f|_{x_1 \leftarrow 0} \vee f|_{x_1 \leftarrow 1},$$

- **universal quantification** via the equivalence

$$\forall x_1 : f(x_1, \dots, x_k) = f|_{x_1 \leftarrow 0} \wedge f|_{x_1 \leftarrow 1},$$

- **composition**  $f|_{x_i \leftarrow g}$  for  $f, g$  Boolean functions,
- **(partial) evaluation**  $f|_{x_i, \dots, x_j \leftarrow b_i, \dots, b_j}$
- ...

# Shared table and other tricks

## Implementing the data structure

Most implementations keep a **unique table**:

- every entry is of the form  $\langle \text{id}, v, \ell, h, \rangle$  and every “function” is a reference to an entry of the table.
- Some implementations also include flags to indicate whether  $\ell$  or  $h$  are **negated** (allows further sharing).

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## Algorithms for the data structure

- Most operations share some code that is usually refactored into the ITE (if-then-else) method.
- After every operation, the unique table is scanned to compute the **size of referenced BDDs**.
- Based on this, a **dynamic-reordering algorithm** may be called to attempt to reduce that size.

# Conclusions

## Theory

- Boolean functions can be represented by truth tables, propositional logic formulas, and BDDs.
- Reduced ordered BDDs are a **canonical** representation of Boolean functions for any fixed variable order.

## Practice

- BDDs can easily be manipulated using the usual logical connectives to obtain BDDs for more complex formulas.
- Under the hood, these operations are implemented using **graph algorithms**.
- The latter allows for many **useful operations** to be applicable to an **already constructed BDD**.