### Planning and Optimization G2. Symbolic Search: BDD Operations and Breadth-First Search

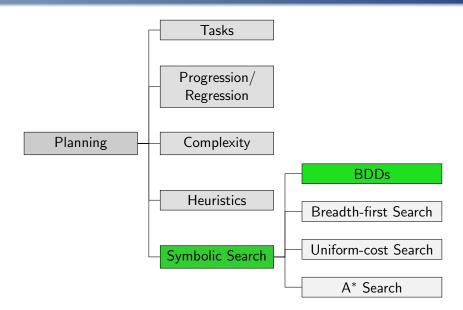
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### Content of this Course



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# **BDD** Operations

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### Reminder: BDD Implementation - Data Structures

#### Data Structures

- Every BDD (including sub-BDDs) B is represented by a single natural number *id*(B) called its ID.
   The zero BDD has ID −2, the one BDD ID −1.
- There are three global vectors to represent the decision variable, the 0- and the 1-successor of non-sink BDDs:
- There is a global hash table *lookup* which maps, for each ID *i* ≥ 0 representing a BDD in use, the triple ⟨*var*[*i*], *low*[*i*], *high*[*i*]⟩ to *i*.

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### **BDD** Operations: Notations

For convenience, we introduce some additional notations:

- We define **0** := *zero*(), **1** := *one*().
- We write var, low, high as attributes:
  - **B.var** for var[B]
  - B.low for low[B]
  - **B.high** for *high*[B]

### Essential vs. Derived BDD Operations

#### We distinguish between

- essential BDD operations, which are implemented directly on top of zero, one and bdd, and
- derived BDD operations, which are implemented in terms of the essential operations.

# **Essential BDD Operations**

We study the following essential operations:

- bdd-includes(B, s): Test  $s \in r(B)$ .
- bdd-equals(B, B'): Test r(B) = r(B').
- bdd-atom(v): Build BDD representing  $\{s \mid s(v) = 1\}$ .
- bdd-state(s): Build BDD representing {s}.
- bdd-union(B, B'): Build BDD representing  $r(B) \cup r(B')$ .
- bdd-complement(*B*): Build BDD representing  $\overline{r(B)}$ .
- bdd-forget(*B*, *v*): Described later.

### Essential Operations: Memoization

- The essential functions are all defined recursively and are free of side effects.
- We assume (without explicit mention in the pseudo-code) that they all use dynamic programming (memoization):
  - Every **return** statement stores the arguments and result in a memo hash table.
  - Whenever a function is invoked, the memo is checked if the same call was made previously. If so, the result from the memo is taken to avoid recomputations.
- The memo may be cleared when the "outermost" recursive call terminates.
  - The bdd-forget function calls the bdd-union function internally. In this case, the memo for bdd-union may only be cleared once bdd-forget finishes, not after each bdd-union invocation finishes.

Memoization is critical for the mentioned runtime bounds.

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# Essential BDD Operations: bdd-includes

#### Test $s \in r(B)$

```
def bdd-includes(B, s):

if B = 0:

return false

else if B = 1:

return true

else if s[B.var] = 1:

return bdd-includes(B.high, s)

else:

return bdd-includes(B.low, s)
```

- Runtime: O(k)
- This works for partial or full valuations *s*, as long as all variables appearing in the BDD are defined.

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### Essential BDD Operations: bdd-equals

#### Test r(B) = r(B')

**def** bdd-equals(B, B'): return B = B'

• Runtime: O(1)

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### Essential BDD Operations: bdd-atom

### Build BDD representing $\{s \mid s(v) = 1\}$

def bdd-atom(v): return bdd(v, 0, 1)

• Runtime: O(1)

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# Essential BDD Operations: bdd-state

### Build BDD representing $\{s\}$

```
def bdd-state(s):

B := 1

for each variable v of s, in reverse variable order:

if s(v) = 1:

B := bdd(v, 0, B)

else:

B := bdd(v, B, 0)

return B
```

- Runtime: O(k)
- Works for partial or full valuations s.

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### Essential BDD Operations: bdd-state Example

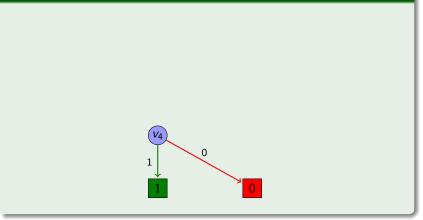
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### Essential BDD Operations: bdd-state Example



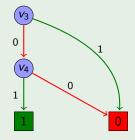
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### Essential BDD Operations: bdd-state Example



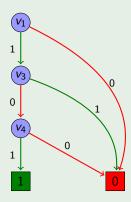
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### Essential BDD Operations: bdd-state Example



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### Essential BDD Operations: bdd-state Example



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# Essential BDD Operations: bdd-union

#### Build BDD representing $r(B) \cup r(B')$

```
def bdd-union(B, B'):
     if B = \mathbf{0} and B' = \mathbf{0}: return \mathbf{0}
    else if B = 1 or B' = 1: return 1
    else if B = 0: return B'
    else if B' = 0: return B
     else if B_{var} < B'_{var}
          return bdd(B.var, bdd-union(B.low, B'),
                              bdd-union(B.high, B'))
    else if B.var = B'.var:
          return bdd(B.var, bdd-union(B.low, B'.low),
                              bdd-union(B.high, B'.high))
    else if B_{var} > B'_{var}:
          return bdd(B'.var, bdd-union(B, B'.low),
                              bdd-union(B, B'.high))
```

• Runtime:  $O(||B|| \cdot ||B'||)$ 

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# Essential BDD Operations: bdd-complement

#### Build BDD representing r(B)

```
def bdd-complement(B):
    if B = 0:
        return 1
    else if B = 1:
        return 0
    else:
        return bdd(B.var, bdd-complement(B.low),
            bdd-complement(B.high))
```

• Runtime: O(||B||)

# Essential BDD Operations: bdd-forget (1)

The last essential BDD operation is a bit more unusual, but we will need it for defining the semantics of operator application.

### Definition (Existential Abstraction)

Let V be a set of propositional variables, let S be a set of variable assignments over V, and let  $v \in V$ . The existential abstraction of v in S, in symbols  $\exists v.S$ , is the set of valuations

$$\{s': (V\setminus\{v\}) 
ightarrow \{0,1\} \mid \exists s \in S: s' \subset s\}$$

over  $V \setminus \{v\}$ .

Existential abstraction is also called forgetting.

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# Essential BDD Operations: bdd-forget (2)

#### Build BDD representing $\exists v.r(B)$

```
def bdd-forget(B, v):

if B = 0 or B = 1 or B.var \succ v:

return B

else if B.var \prec v:

return bdd(B.var, bdd-forget(B.low, v),

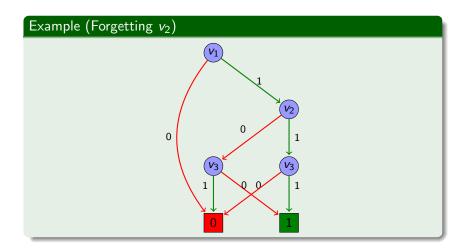
bdd-forget(B.high, v))
```

else:

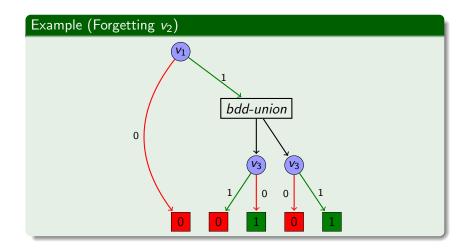
**return** *bdd-union*(*B*.low, *B*.high)

• Runtime:  $O(||B||^2)$ 

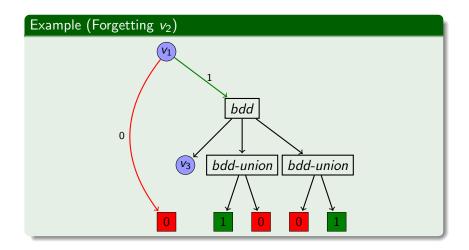
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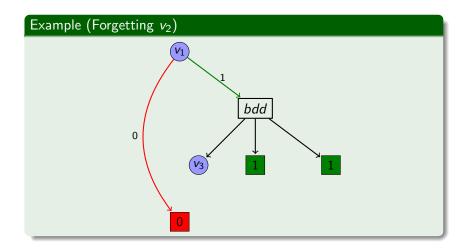
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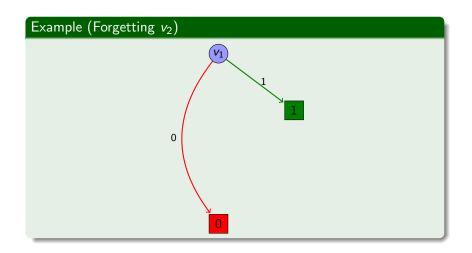
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# Derived BDD Operations

We study the following derived operations:

- bdd-intersection(B, B'): Build BDD representing  $r(B) \cap r(B')$ .
- bdd-setdifference(B, B'): Build BDD representing  $r(B) \setminus r(B')$ .
- bdd-isempty(B): Test  $r(B) = \emptyset$ .
- bdd-rename(*B*, *v*, *v*'):

Build BDD representing {rename(s, v, v') |  $s \in r(B)$  }, where rename(s, v, v') is the variable assignment s with variable v renamed to v'.

• If variable v' occurs in B already, the result is undefined.

# Derived Operations: bdd-intersection, bdd-setdifference

### Build BDD representing $r(B) \cap r(B')$

def bdd-intersection(B, B'):
 not-B := bdd-complement(B)
 not-B' := bdd-complement(B')
 return bdd-complement(bdd-union(not-B, not-B'))

### Build BDD representing $r(B) \setminus r(B')$

def bdd-setdifference(B, B'):
 return bdd-intersection(B, bdd-complement(B'))

- Runtime:  $O(||B|| \cdot ||B'||)$
- These functions can also be easily implemented directly, following the structure of *bdd-union*.

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### Derived BDD Operations: bdd-isempty

### Test $r(B) = \emptyset$

def bdd-isempty(B):
 return bdd-equals(B,0)

• Runtime: O(1)

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# Derived BDD Operations: bdd-rename

#### Build BDD representing {rename(s, v, v') | $s \in r(B)$ }

def bdd-rename(B, v, v'):
 v-and-v' := bdd-intersection(bdd-atom(v), bdd-atom(v'))
 not-v := bdd-complement(bdd-atom(v))
 not-v' := bdd-complement(bdd-atom(v'))
 not-v-and-not-v' := bdd-intersection(not-v, not-v')
 v-eq-v' := bdd-union(v-and-v', not-v-and-not-v')
 return bdd-forget(bdd-intersection(B, v-eq-v'), v)

• Runtime:  $O(||B||^2)$ 

### Derived BDD Operations: bdd-rename Remarks

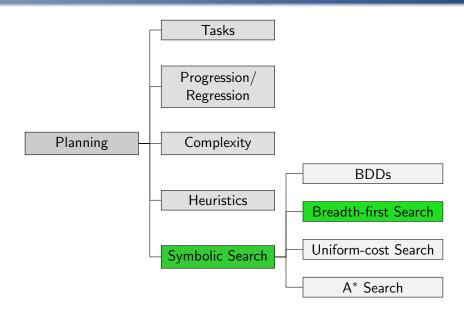
- Renaming sounds like a simple operation.
- Why is it so expensive?

This is **not** because the algorithm is bad:

- Renaming must take at least quadratic time:
  - There exist families of BDDs  $B_n$  with k variables such that renaming  $v_1$  to  $v_{k+1}$  increases the size of the BDD from  $\Theta(n)$ to  $\Theta(n^2)$ .
- However, renaming is cheap in some cases:
  - For example, renaming to a neighboring unused variable (e.g. from  $v_i$  to  $v_{i+1}$ ) is always possible in linear time by simply relabeling the decision variables of the BDD.
- In practice, one can usually choose a variable ordering where renaming only occurs between neighboring variables.

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# Breadth-first Search with Progression and BDDs

#### Progression Breadth-first Search

```
def bfs-progression(V, I, O, \gamma):
      goal := formula-to-set(\gamma)
      reached_0 := \{I\}
      i := 0
      loop:
           if reached<sub>i</sub> \cap goal \neq \emptyset:
                 return solution found
            reached_{i+1} := reached_i \cup apply(reached_i, O)
           if reached_{i+1} = reached_i:
                 return no solution exists
            i := i + 1
```

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# Breadth-first Search with Progression and BDDs

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           i := i + 1
```

Use bdd-atom, bdd-complement, bdd-union, bdd-intersection.

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```

Use bdd-state.

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```

Use bdd-intersection, bdd-isempty.

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### Breadth-first Search with Progression and BDDs

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```

Use bdd-union.

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### Breadth-first Search with Progression and BDDs

#### Progression Breadth-first Search

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```

Use bdd-equals.

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### Breadth-first Search with Progression and BDDs

#### Progression Breadth-first Search

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      i := 0
      loop:
            if reached<sub>i</sub> \cap goal \neq \emptyset:
                  return solution found
            reached_{i+1} := reached_i \cup \frac{apply}{reached_i}, O
            if reached_{i+1} = reached_i:
                  return no solution exists
            i := i + 1
```

How to do this?

# The *apply* Function (1)

- We need an operation that, for a set of states *reached<sub>i</sub>* (given as a BDD) and a set of operators *O*, computes the set of states (as a BDD) that can be reached by applying some operator *o* ∈ *O* in some state *s* ∈ *reached*.
- We have seen something similar already...

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### Translating Operators into Formulae

#### Definition (Operators in Propositional Logic)

Let *o* be an operator and *V* a set of state variables. Define  $\tau_V(o) := pre(o) \land \bigwedge_{v \in V} (regr(v, eff(o)) \leftrightarrow v')$ .

States that o is applicable and describes when the new value of v, represented by v', is **T**.

# The apply Function (2)

- The formula τ<sub>V</sub>(o) describes the applicability of a single operator o and the effect of applying o as a binary formula over variables V (describing the state in which o is applied) and V' (describing the resulting state).
- The formula  $\bigvee_{o \in O} \tau_V(o)$  describes state transitions by any operator in O.
- We can translate this formula to a BDD (over variables V ∪ V') using bdd-atom, bdd-complement, bdd-union, bdd-intersection.
- The resulting BDD is called the transition relation of the planning task, written as  $T_V(O)$ .

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# The *apply* Function (3)

Using the transition relation, we can compute *apply(reached, O)* as follows:

### The apply function

```
def apply(reached, O):

B := T_V(O)

B := bdd-intersection(B, reached)

for each v \in V:

B := bdd-forget(B, v)

for each v \in V:

B := bdd-rename(B, v', v)

return B
```

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This describes the set of state pairs  $\langle s, s' \rangle$  where s' is a successor of s in terms of variables  $V \cup V'$ .

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Summary 00

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This describes the set of states s' which are successors of some state  $s \in$  reached in terms of variables V'.

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Using the transition relation, we can compute *apply(reached, O)* as follows:

### The apply function

```
\begin{array}{l} \textbf{def apply}(reached, \ O):\\ B := T_V(O)\\ B := bdd\text{-}intersection(B, reached)\\ \textbf{for each } v \in V:\\ B := bdd\text{-}forget(B, v)\\ \textbf{for each } v \in V:\\ B := bdd\text{-}rename(B, v', v)\\ \textbf{return } B \end{array}
```

Thus, *apply* indeed computes the set of successors of *reached* using operators *O*.

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### Plan Extraction

We can construct a plan from the BDDs *reached*<sub>i</sub> (set given as parameter *reached*<sub>\*</sub>):

#### Construct Plan

```
def construct_plan(I, O, \gamma, reached<sub>*</sub>, i_{max}):
      goal := BDD for \gamma
      s := arbitrary state from bdd-intersection(goal, reached_{imax})
      \pi := \langle \rangle
      for i = i_{max} - 1 to 0:
            for o \in O:
                  p := BDD for regr(s, o)
                  if c := bdd-intersection(p, reached_i) \neq \mathbf{0}:
                        s := arbitrary state from c
                        \pi := \langle o \rangle \pi
                        break
      return \pi
```

Remarks

BDDs can be used to implement a blind breadth-first search algorithm in an efficient way.

- For good performance, we need a good variable ordering.
  - Variables that refer to the same state variable before and after operator application (v and v') should be neighbors in the transition relation BDD.
- Use mutexes to reformulate as a multi-valued task.
  - Use  $\lceil \log_2 n \rceil$  BDD variables to represent a variable with *n* possible values.

# Summary



- Binary decision diagrams are a data structure to compactly represent and manipulate sets of valuations.
- They can be used to implement a blind breadth-first search algorithm in an efficient way.