

CIS 673: Lecture 6: Sept 25
Symbolic Reachability

- Instead of enumerating states, compute with regions (state-sets) that are represented symbolically (eg. $20 \leq x \leq 99$)
- Implicit representation of transition relation and reachable states
- For propositional or enumerated modules, binary decision diagrams serve as a compact representation
 - Binary decision diagrams: representation for boolean functions due to Bryant [87]
 - Application to model checking: tool SMV by McMillan [91]
 - Popular in hardware applications

Symbolic Data Types

- Symbolic regions support following operations
 - $\cup: \mathbf{symreg} \times \mathbf{symreg} \mapsto \mathbf{symreg}$
 - $\cap: \mathbf{symreg} \times \mathbf{symreg} \mapsto \mathbf{symreg}$
 - $=: \mathbf{symreg} \times \mathbf{symreg} \mapsto \mathbf{bool}$
 - $\subseteq: \mathbf{symreg} \times \mathbf{symreg} \mapsto \mathbf{bool}$
 - *EmptySet*: \mathbf{symreg}
- Note: No enumeration, so number of states in a region is not an issue
- Symbolic transition graph supports
 - *InitReg*: $\mathbf{symgraph} \mapsto \mathbf{symreg}$
 - *PostReg*: $\mathbf{symreg} \times \mathbf{symgraph} \mapsto \mathbf{symreg}$

Symbolic Search

Input: a transition graph G , and a region σ^T

Output: answer to reachability problem (G, σ^T) .

input G : symgraph; σ^T : symreg;

local σ^R : symreg;

begin

$\sigma^R := \text{InitReg}(G)$;

repeat

if $\sigma^R \cap \sigma^T \neq \text{EmptySet}$ then

return Yes;

if $\text{PostReg}(\sigma^R, G) \subseteq \sigma^R$ then

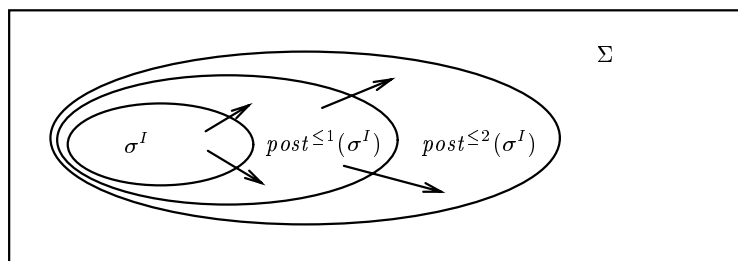
return No;

$\sigma^R := \sigma^R \cup \text{PostReg}(\sigma^R, G)$

forever

end.

Symbolic Search



Like breadth-first search, guaranteed to terminate if the diameter of the graph is bounded: there is i such that every reachable state is reachable within i transitions from some initial state

Detour: Fixpoint Theory

- Partial order \preceq : reflexive, transitive, and anti-symmetric relation
- Least upper bound a of a set B wrt \preceq , denoted $\vee B$
 - $b \preceq a$ for all $b \in B$
 - if $b \preceq c$ for all $b \in B$, then $a \preceq c$
- Similarly, greatest lower bound $\wedge B$ defined
- Complete partial order: all sets have lub and glb
- The powerset of a set with \subseteq is a CPO:
 - lub corresponds to set-union
 - glb corresponds to set-intersection

Fixpoint Theory

- Function f is monotonic: $a \preceq b$ then $f(a) \preceq f(b)$
- a is fixpoint of f : $f(a) = a$
- Knaster-Tarski fixpoint theorem:
If f is monotonic then f has least and greatest fixpoints: μf and νf
- f is \vee -continuous: $f(\vee B) = \vee f(B)$
- Kleene fixpoint theorem: If f is monotonic and \vee -continuous, then $\mu f = \vee \{f^\kappa(\perp) \mid \kappa \in \mathbb{O}\}$
- For countable sets: $\mu f = \vee \{f^i(\perp) \mid i \in \mathbb{N}\}$
- Summary: Least (or greatest) fixpoints can be approximated by starting from \perp , and applying f repeatedly

Symbolic reachability as fixpoint computation

- CPO: set of regions ordered by \subseteq
- Consider $f(\sigma) = \sigma^I \cup post_G(\sigma)$
- f is monotonic, \cup -continuous
- Least fixpoint of f : reachable region σ^R
- i -th approximation of reachable region: $f^i(\emptyset)$
- Many of our later symbolic algorithms will correspond to computing fixpoints (μ -calculus)

Symbolic Representation

- Type of variable x : \mathbb{T}_x
- Type of state over X : product type \mathbb{T}_X
- Type of transition: $\mathbb{T}_{X \cup X'}$
- Transition relation is, thus, a symbolic region over primed and unprimed variables
- Symbolic representation of a transition graph G
 - initial region $\{\sigma^I\}_s$ of type **symreg** $[\mathbb{T}_X]$
 - transition relation $\{\rightarrow\}_s$ of type **symreg** $[\mathbb{T}_{X \cup X'}]$
- We simply need to decide on how to represent regions

Computing Post

- Renaming: $Rename(x, y, \sigma)$ returns the renamed region $\sigma[x := y]$
- Renaming extends to variable-sets
- Existential quantifier elimination: For σ of type $\mathbf{symreg}[\mathbb{T}_X]$, $Exists(x, \sigma)$ returns the region

$$\{s \in \Sigma_{X \setminus \{x\}} \mid (\exists m. s[x := m] \in \sigma)\}$$

- Existential quantifier elimination extends to variable sets
- Computing $PostReg(\sigma)$:
 - conjunct σ with $\{\rightarrow\}_s$ to obtain the set of transitions originating in σ
 - project the result onto the set X'
 - rename each primed variable x' to x

$$PostReg(\sigma, \{G\}_s) = Rename(X', X, Exists(X, \sigma \cap \{\rightarrow\}_s))$$

Summary

To implement symbolic reachability, we need

- implementation the data type `symreg` that supports \cup , \cap , $=$, \subseteq , `Rename`, and `Exists`
- an efficient way to compute, from the module text, the symbolic representation of
 - the initial region $\{\sigma^I\}_s$
 - the transition region $\{G\}_s$

Symbolic search using propositional formulas

- Represent regions by (parse trees) of boolean expression
- Union, intersection easy
- Renaming: textual substitution
- Quantifier elimination in linear time:

$$\textit{Exists}(x, p) = (p[x := \textit{true}] \vee p[x := \textit{false}]).$$

- Inclusion (\subseteq) or equivalence ($=$):
 - checking validity of propositional formulas
 - hard: coNP-complete
- Obtaining symbolic representations of σ^I and \rightarrow is easy (no blow-up)

Symbolic reachability for Pete

- For module P_1 : $X = \{pc_1, pc_2, x_1, x_2\}$

- Initial region q_1^I : $pc_1 = out$

- Symbolic rep of transition relation q_1^T :

$$\begin{aligned} & \vee (pc_1 = out \wedge pc'_1 = req \wedge x'_1 = x_2) \\ & \vee (pc_1 = req \wedge (pc_2 = out \vee x_1 \neq x_2) \wedge pc'_1 = in \wedge x'_1 = x_1) \\ & \vee (pc_1 = in \wedge pc'_1 = out \wedge x'_1 = x_1) \\ & \vee (pc'_1 = pc_1 \wedge x'_1 = x_1). \end{aligned}$$

- One step of fixpoint computation:

$$PostReg(q_1^I) = Rename(X', X, Exists(X, q_1^I \cap q_1^T))$$

- Simplifies to $pc_1 = out \vee pc_1 = req$

Can we improve formula representation?

- Propositional formulas not very satisfactory:
 - Each step of the computation may be expensive (equality test)
 - More importantly, size of $PostReg^{\leq i}(q^I)$ may grow with i , with no good heuristics for simplification
- Desirable properties of representation:
 - All operations (and hence, a single iteration) should be efficient
 - Should maintain compactness whenever possible: canonicity
 - Representing q^I and q^T from module text should be efficient

Ordered Binary Decision Graphs

- Representation for predicates over a set X of boolean vars
- The variables in X are totally ordered by \prec
- BDG is a directed acyclic graph
- Paths in the graph encode assignments to variables in X
- Terminal vertices classify paths into accepting and rejecting

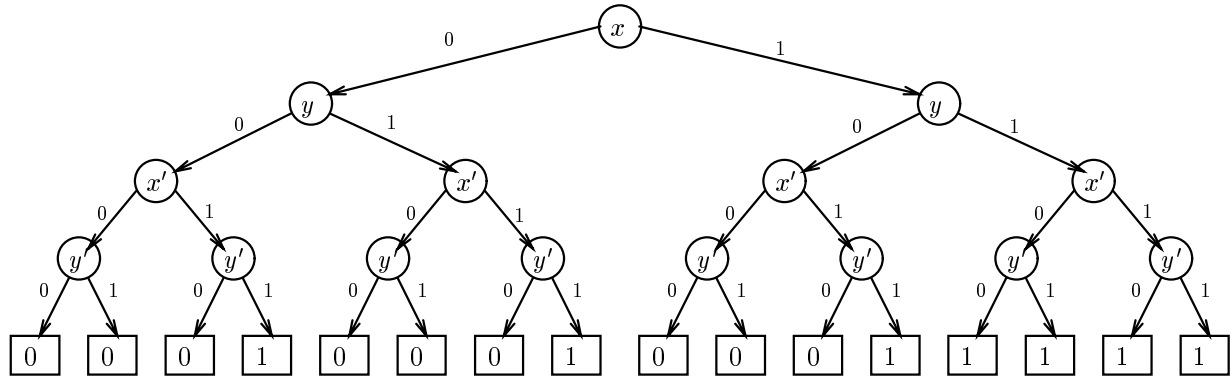
Definition of BDG wrt (X, \prec)

- Vertices: a finite set V that is partitioned into two sets; internal vertices V^N and terminal vertices V^T
- Root: a root vertex v^I in V
- Labeling: a labeling function $label$ that labels each internal vertex with a variable in X , and each terminal vertex with a constant in **bool**
- Left edges: a function $left : V^N \mapsto V$ that maps each internal vertex v to a vertex $left(v)$ such that if $left(v)$ is an internal vertex then $label(v) \prec label(left(v))$
- Right edges: a function $right : V^N \mapsto V$ that maps each internal vertex v to a vertex $right(v)$ such that if $right(v)$ is an internal vertex then $label(v) \prec label(right(v))$.

Sample BDG

Function: $(x \wedge y) \vee (x' \wedge y')$

Ordering: $x \prec y \prec x' \prec y'$



Boolean Function of a BDG

- On every path from root to a terminal vertex, a variable appears 0 or 1 times
- With every vertex v , associate a boolean function $r(v)$
 - if v is terminal then $r(v)$ equals its label (0 or 1)
 - if v is internal then $r(v)$ equals
$$(\neg \text{label}(v) \wedge r(\text{left}(v))) \vee (\text{label}(v) \wedge r(\text{right}(v)))$$
- BDG B represents the function $r(B) = r(v^I)$ associated with its root
- To check whether a state s satisfies $r(B)$ simply follow path according to values assigned by s

Isomorphism and Equivalence

- Two BDGs B and C are isomorphic if the corresponding labeled graphs are isomorphic
- Two BDGs B and C are equivalent if the boolean expressions $r(B)$ and $r(C)$ are equivalent.
- Deciding isomorphism is easy
- Deciding equivalence is difficult (so this is not what we want for symbolic reachability)
- Equivalence does not imply isomorphism
- Every vertex v can be viewed as a BDD (subgraph rooted at v), so we can talk about isomorphism and equivalence of vertices

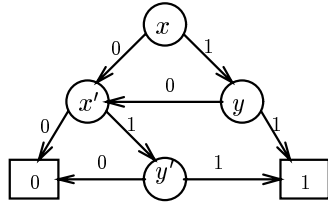
Ordered Binary Decision Diagrams

- BDD is obtained from a BDG by repeatedly applying following transformations:
 - Identify isomorphic subgraphs.
 - Eliminate internal vertices with identical left and right successors.
- An ordered binary decision diagram over a totally ordered set (X, \prec) is an ordered binary decision graph B over (X, \prec) with vertices V and root v^I such that
 - No isomorphic subgraphs: if v and w are two distinct vertices in V , then v is not isomorphic to w
 - No redundancy: for every internal vertex v , the two successors $left(v)$ and $right(v)$ are distinct.

Sample BDD

Function: $(x \wedge y) \vee (x' \wedge y')$

Ordering: $x \prec y \prec x' \prec y'$

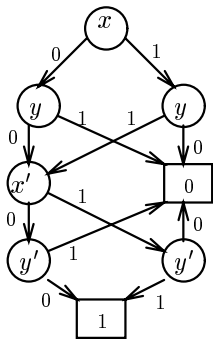


Properties of BDDs

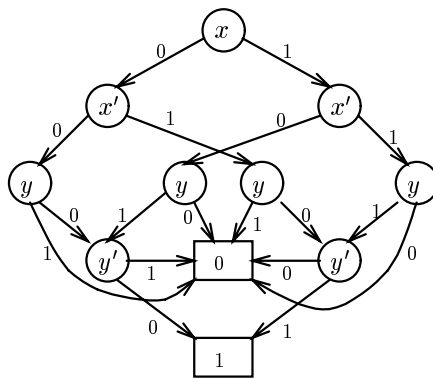
- Given an ordered set (X, \prec) , every boolean function has a unique, upto isomorphism, representation as a BDD (i.e. existence and canonicity)
- Isomorphism and equivalence are synonyms for BDDs
- Testing equivalence: easy (linear-time)
- $r(B)$ is valid iff B equals a single terminal vertex labeled with 1
- Testing validity (and satisfiability) is easy (constant-time)
- This implies translating a propositional formula to a BDD can cause an exponential blow-up
- Minimality: If B is a BDD and C is an equivalent BDG (with same ordering) then C has at least as many vertices as B

Ordering influences size

Two BDDs for $(x \leftrightarrow y) \wedge (x' \leftrightarrow y')$



Ordering: $x \prec y \prec x' \prec y'$



Ordering: $x \prec x' \prec y \prec y'$

More on variable ordering

- The size according to one ordering may be exponentially smaller than another ordering
- Figuring out the optimal ordering of variables is hard
- Theorem: The problem of checking, given a BDD B over (X, \prec) , whether the ordering \prec is optimal for $r(B)$, is coNP-complete.
- Some functions have small size independent of ordering, eg. parity
- Some functions have large size independent of ordering, multiplication

BDD Implementation

- Vertices of all BDDs stored in a central pool
- Each BDD is simply an index into this global data structure
- Invariant property: no two vertices in the pool are isomorphic
- Advantages:
 - To check equivalence, compare indices
 - Sharing of subgraphs across nonisomorphic BDDs
- Performance depends on efficient garbage collection

BDD Data Structures

- Type of BDDs: `bdd`. Pointer to global data structure *BddPool*.

- Type of *BddPool*: **set of `bddnode`**

- Type of vertices: (*k* is number of variables)

$$\mathbf{bddnode} = ([1..k] \times \mathbf{bdd} \times \mathbf{bdd}) \cup \mathbf{bool}$$

- **`bddnode`** supports

- *Label*: $\mathbf{bddnode} \mapsto [1..k]$

- *Left*: $\mathbf{bddnode} \mapsto \mathbf{bdd}$

- *Right*: $\mathbf{bddnode} \mapsto \mathbf{bdd}$

- *BddPool* has following operations

- Insert and IsMember

- *Index*: $\mathbf{bddnode} \mapsto \mathbf{bdd}$

- *BddPool*[*B*] returns vertex pointed to by *B*

Insertions into *BddPool*

- *BddPool* is initialized to contain two terminal vertices 0 and 1
- Add vertices only via *MakeVertex* to ensure that no vertices are isomorphic

function *MakeVertex*

Input: $i: [1..k]$, $B_0, B_1: \mathbf{bdd}$.

Output: $B: \mathbf{bdd}$ such that $r(B)$ equivalent
to $(\neg x_i \wedge r(B_0)) \vee (x_i \wedge r(B_1))$.

begin

if $B_0 = B_1$ then return B_0 fi;

if $\neg \text{IsMember}((i, B_0, B_1), \text{BddPool})$ then

$\text{Insert}((i, B_0, B_1), \text{BddPool})$ fi;

return $\text{Index}((i, B_0, B_1))$

end.

Operations on BDDs

- Algorithms for boolean operations on BDDs
- Conjunction:
 - Input: Two pointers B_0 and B_1 into *BddPool*
 - Output: A pointer B such that the BDD represented by *BddPool*[B] should be the conjunction of the BDDs represented by *BddPool*[B_0] and *BddPool*[B_1]
- Recursive function *Conj* to take conjunction of vertices

Algorithm for Conjunction

```
function Conj
input  $B_0, B_1$  : bdd
output  $B$  : bdd
local  $v_0, v_1$  : bddnode;  $i, j$  :  $[1 \dots k]$ 
       $B, B_{00}, B_{01}, B_{10}, B_{11}$  : bdd
begin
   $v_0 := BddPool[B_0]$ ;
   $v_1 := BddPool[B_1]$ ;
  if  $v_0 = 0$  or  $v_1 = 1$  then return  $B_0$  fi;
  if  $v_0 = 1$  or  $v_1 = 0$  then return  $B_1$  fi;
   $i := Label(v_0)$ ;  $j := Label(v_1)$ ;
   $B_{00} := Left(v_0)$ ;  $B_{01} := Right(v_0)$ 
   $B_{10} := Left(v_1)$ ;  $B_{11} := Right(v_1)$ ;
  if  $i = j$  then
     $B := MakeVertex(i, Conj(B_{00}, B_{10}), Conj(B_{01}, B_{11}))$ 
  if  $i < j$  then
     $B := MakeVertex(i, Conj(B_{00}, B_1), Conj(B_{01}, B_1))$ 
  if  $i > j$  then
     $B := MakeVertex(j, Conj(B_0, B_{10}), Conj(B_0, B_{11}))$ 
  return  $B$ 
end.
```

Avoiding Recomputation

- Store computed results in a table *Done*
- Keys for *Done*: pairs of the form **bdd** \times **bdd**
- Stores value of the type **bdd**
- Initially empty
- Before recursive calls, *Conj* checks if *Done* $[(B_0, B_1)]$ has a value
- Upon termination *Conj* stores the result in *Done*

Modified Algorithm

```
function Conj
begin
   $v_0 := BddPool[B_0]; v_1 := BddPool[B_1];$ 
  if  $v_0 = 0$  or  $v_1 = 1$  then return  $B_0$  fi;
  if  $v_0 = 1$  or  $v_1 = 0$  then return  $B_1$  fi;
  if  $Done[(B_0, B_1)] \neq \perp$  then return  $Done[(B_0, B_1)]$ 
  if  $Done[(B_1, B_0)] \neq \perp$  then return  $Done[(B_1, B_0)]$ 
   $i := Label(v_0); j := Label(v_1);$ 
   $B_{00} := Left(v_0); B_{01} := Right(v_0)$ 
   $B_{10} := Left(v_1); B_{11} := Right(v_1);$ 
  if  $i = j$  then
     $B := MakeVertex(i, Conj(B_{00}, B_{10}), Conj(B_{01}, B_{11}))$ 
  if  $i < j$  then
     $B := MakeVertex(j, Conj(B_{00}, B_{11}), Conj(B_{01}, B_{10}))$ 
  if  $i > j$  then
     $B := MakeVertex(j, Conj(B_0, B_{10}), Conj(B_0, B_{11}))$ 
   $Done[(B_0, B_1)] := B;$ 
  return  $B$ 
end.
```

Operations on BDDs

- If B_0 contains m nodes and B_1 has n nodes then *Conj* has time complexity $O(mn)$
- Quadratic bound tight: conjunction of two BDDs with n nodes can have n^2 nodes (like finite automata)
- Similar algorithms for union, negation, simplification (i.e. computing $B[x := 0]$)
- Now we can translate a propositional formula to its BDD representation
- Translation can cause exponential blow-up

Multi-valued Decision Diagrams (MDD)

- Just like BDD, except choice need not be binary
- Variables have enumerated types
- Branching degree at an internal vertex x is k if x can take k values:
 - An internal vertex labeled with x has a successor edge labeled with m for every $m \in \mathbb{T}_x$
- Terminal vertices have 0 or 1 label as before
- Useful for verification of enumerated modules

Symbolic Invariant Verification

To solve the invariant verification problem (P, p)

1. Choose variable ordering over $X_P \cup X'_P$
2. Build the BDDs for
 - Initial region of P
 - Target region (negation of invariant p)
 - Transition relation of P
3. Execute symbolic search algorithm

Partitioned Transition Relation

- Transition relation is conjunction of predicates
- Do not compute the conjunction a priori, but only as needed (on-the-fly symbolic representation)
- A conjunctively-partitioned representation of a boolean expression p is a set $\{B_1, \dots, B_k\}$ of BDDs such that p is equivalent to the conjunction $r(B_1) \wedge \dots \wedge r(B_k)$
- The transition relation q^T is maintained as a set $\{q_1^T, \dots, q_k^T\}$ can have much smaller BDD representation
- In each iteration, if q is the current reachable region, we need to compute $q \wedge q^T$
- In each iteration, we need compute the conjunction $q \wedge q_1^T \wedge \dots \wedge q_k^T$

- This is typically much smaller than BDD for q^T : only reachable transitions are considered (the domain of various conjuncts q_i^T is constrained, or simplified by q)
- For us, for every atom, there is a conjunct representing update command of the atom
- Further partitioning possible: eg. each conjunct is a disjunction of guarded commands, maintain this in a disjunctively partitioned format

Early Quantifier Elimination

- If p does not depend upon x then

$$\exists x. p \wedge q \equiv p \wedge \exists x. q$$

- BDD for $\exists x. q$ is likely to be smaller than BDD of q
- In computing *PostReg*, we need to compute

$$\exists X. (q \wedge q_1^T \wedge \dots \wedge q_k^T)$$

- Strategy: take one conjunction at a time, eliminate (via existential quantification) as many variables as possible
- Quantifier elimination has a precedence over conjunction

Ordering of Conjuncts

- Effectiveness of early quantifier elimination depends on the ordering of conjuncts
- In 3-bit counter example, we want to compute

$$\exists\{out_0, out_1, out_2\}. (p \wedge q_0^T \wedge q_1^T \wedge q_2^T)$$

- If we compute $p \wedge q_0^T$ first, no existential quantification can be applied
- If we compute $p \wedge q_2^T$ first, out_2 can be eliminated
- Optimal computing strategy:

$$\exists out_0. (q_0^T \wedge \exists out_1. (q_1^T \wedge \exists out_2. (q_2^T \wedge p)))$$

Dynamic Variable Reordering

- Symbolic computation computes successive approximation q_i to the reachable region
- What if, at some iteration, BDD representation of q_i becomes too large?
- One solution: try to reorder variables
- Swapping adjacent variables, say x_i and x_{i+1} , can be done reasonable efficiently:
 - BDD structure and above level i and below level $i + 1$ stays unchanged
- Greedy heuristic: find the level at which the BDD is fattest, and try swapping with adjacent level. Repeat if desired
- All BDDs of current relevance must be updated: expensive step