

# CSEP590 – Model Checking and Automated Verification

Lecture outline for July 30, 2003

- We will first finish up Timed Automata from the last lecture...
- The fixed point characterization of CTL
  - We discuss this issue to motivate a proof of correctness of our model checking algorithm for CTL
  - This also provides necessary background for discussing the relational mu-calculus and its applications to model checking
- Recall: given a Model  $M = (S, \rightarrow, L)$ , our algorithm computes all  $s \in S$  s.t.  $M, s \models \phi$  for a CTL formula  $\phi$ 
  - We denote this set as  $\{\phi\}$
  - Our algorithm is recursive on the structure of  $\phi$
  - For boolean operators it is easy to find  $\{\phi\}$  via combinations of subsets using Union, Intersection, etc
  - An interesting case though is a formula involving a temporal operator (such as  $EX \phi$ )
    - We compute the set  $\{\phi\}$ , then compute the set of all states with transitions to a state in  $\{\phi\}$
  - How do we reason about EU, AF, and EG? – we are iterating a labelling policy until stabilised!

-But how do we know that such iterations will terminate and even return the correct sets?? How can we argue this?

-Defn: let  $S$  be a set of states and  $F: P(S) \rightarrow P(S)$  be a function on the power set of  $S$  (where  $P(S)$  denotes power set of  $S$ ). Then,

-1)  $F$  is monotone if  $x \subseteq Y$  implies that  $F(X) \subseteq F(Y)$  for all subsets  $X$  and  $Y$  of  $S$

-2) A subset  $X$  of  $S$  is called a fixed point of  $F$  if  $F(X) = X$

-We'll see an example in class of fixed points and monotone functions. Indeed, a greatest fixed point is a subset  $X$  that is a fixed point and has the largest size. A least fixed point can be defined similarly

-Why are we exploring monotone functions?

-They always have a least and greatest fixed point

-The meanings of EG, AF, EU can be expressed via greatest and least fixed points of monotone function on  $P(S)$  ( $S$  = set of states)

-Fixed points are easily computed

- Notation:  $F^i(X) = F(F(\dots F(X)\dots)) \Rightarrow$  a function  $F$  applied  $i$  times
- Theorem: Let  $S$  be a set  $\{s_0, s_1, \dots, s_n\}$  with  $n+1$  elements. If  $F: P(S) \rightarrow P(S)$  is a monotone function, then  $F^{n+1}(\emptyset)$  is the least fixed point of  $F$ , and  $F^{n+1}(S)$  is the greatest fixed point of  $F$ .
  - Proof: in book on page 207
  - This theorem provides a recipe for computing fixed points!  
Indeed, the method is bounded at  $n+1$  iterations.
- Now, we can prove the correctness of our model checking algorithm
  - Proof that EG algorithm is correct:
    - We could say that  $EG \phi = \phi \wedge EXEG \phi$  (call this (1))
    - Also,  $\{EG \phi\} = \{s \mid \text{exists } s' \text{ s.t. } s \rightarrow s' \text{ and } s' \in \{\phi\}\}$
    - Thus, we can rewrite (1) as
      - $\{EG \phi\} = \{\phi\} \cap \{s \mid \text{exists } s' \text{ s.t. } s \rightarrow s' \text{ and } s' \in \{EG \phi\}\}$
      - Thus, we calculate  $\{EG \phi\}$  from  $\{EG \phi\}$  – this sounds like a fixed point operation!
    - Indeed,  $\{EG \phi\}$  is a fixed point of the function
      - $F(X) = \{\phi\} \cap \{s \mid \text{exists } s' \text{ s.t. } s \rightarrow s' \text{ and } s' \in X\}$

- F is monotone, and  $\{EG \phi\}$  is its greatest fixed point
  - (Formal proof is in book on pg. 209)
- $\{EG \phi\}$  can be computed using our theorem for fixed points, applied iteratively
  - ie,  $\{EG \phi\} = F^{n+1}(S)$  where  $n+1=|S|$
- Thus, correctness of EG procedure is proved and it is guaranteed to terminate in at most  $|S|$  iterations

-The book gives similar fixed point analysis for the EU operator, showing that its algorithm is also correct

-This, when combined with the correctness of EX and the boolean operators, completes proof of correctness of our CTL model checking algorithm

-Now, let's discuss the relational mu-calculus and how model checking can be performed in it

-We introduce a syntax for referring to fixed points in the context of boolean formulas

-Formulas of the relational mu-calculus grammar:

- $t = x \mid Z$

- $f = 0 \mid 1 \mid t \mid !f \mid f_1 + f_2 \mid f_1 * f_2 \mid \exists x.f \mid \forall x.f \mid uZ.f \mid vZ.f \mid f[X=X']$

-Where  $x$  is a boolean variable,  $Z$  is a relational variable, and  $X$  is a tuple of variables

-A relational variable can be assigned a subset of  $S$  (set of states)

-In formulas  $uZ.f$  and  $vZ.f$  any occurrence of  $Z$  in  $f$  is required to fall within an even # of complementation symbols

-Such an  $f$  is called formally monotone in  $Z$

-Symbols  $u$  and  $v$  stand for least and greatest fixed point operators

-Thus,  $uZ.f$  means “least fixed point of function  $f$ ” (where the iteration is “occurring” on relational variable  $Z$ . The “returned”  $Z$  is the least fixed point of  $f$ )

- The formula  $f[X=X']$  expresses the explicit substitution forcing  $f$  to be evaluated using the values of  $x_i'$  rather than  $x_i$  (allows for notions of “next time” evaluations, like successors)
- A valuation  $p$  for  $f$  is an assignment of values 0 or 1 to all variables
- Define: satisfaction relation  $p \models f$  inductively over the structure of such formulas  $f$ , given a valuation  $p$
- We define  $\models$  for formulas without fixed point operators:
  - $p \not\models 0$ ,  $p \models 1$ ,  $p \models v$  iff  $p(v)=1$ ,  $p \models !f$  iff  $p \not\models f$ ,  $p \models f+g$  iff  $p \models f$  or  $p \models g$ ,  $p \models f*g$  iff  $p \models f$  and  $p \models g$ ,  $p \models \exists x.f$  iff  $p[x=0] \models f$  or  $p[x=1] \models f$ ,  $p \models \forall x.f$  iff  $p[x=0] \models f$  and  $p[x=1] \models f$ ,  $p \models f[X=X']$  iff  $p[X=X'] \models f$
  - Where  $p[X=X']$  is the valuation assigning the same values as  $p$  but for each  $x_i$  in  $X$ , it assigns  $p(x_i')$
  - We'll see a few examples in class that make all this jumbled notation clearer
- Now, we extend the  $\models$  definition to fixed point operators  $\mu$  and  $\nu$

- $p \models uZ.f$  iff  $p \models u_m Z.f$  for some  $m \geq 0$

-Where  $uZ.f$  is recursively defined as

- $u_0 Z.f = 0$

- $u_m Z.f = f[u_{m-1} Z.f/Z]$  (that is, replace all occurrences of  $Z$  in  $f$  with  $u_{m-1} Z.f$ )

- $p \models vZ.f$  iff  $p \models v_m Z.f$  for *all*  $m \geq 0$

-Where  $vZ.f$  is recursively defined as

- $v_0 Z.f = 1$

- $v_m Z.f = f[v_{m-1} Z.f/Z]$

-We'll see some examples in class that will make this intuitive. Essentially, these are just recursive definitions, they iterate to fixed points

-So now we can code CTL models and specifications

-Given a model  $M=(S, \rightarrow, L)$ , the  $u$  and  $v$  operators permit us to translate any CTL formula  $\phi$  into a formula  $f^\phi$  of the relational mu-calculus s.t.  $f^\phi$  represents the set of states  $s$  where  $s \models \phi$

-Then, given a valuation  $p$  (ie, a state), we can check if  $p \models f^\phi$ , meaning that the state satisfies  $\phi$



-Indeed, we can do this purely symbolically

-Recall that the transition relation  $\rightarrow$  can be represented as a boolean formula  $f^{\rightarrow}$  (from our symbolic model checking lecture 4). Also, sets of states can be encoded as boolean formulas

-Therefore, the coding of a CTL formula  $\phi$  as a function  $f^{\phi}$  in relational mu-calculus is given inductively:

- $f^x = x$  for vars  $x$

- $f^{\perp} = 0$

- $f^{!\phi} = !f^{\phi}$

- $f^{\phi \vee \varphi} = f^{\phi} * f^{\varphi}$

- $f^{\text{EX}\phi} = \exists X'. (f^{\rightarrow} * f^{\phi}[X=X'])$

-What the heck does that mean? “There exists a next state s.t. the transition relation holds from the current state AND  $f^{\phi}$  holds in this next state”

-We can also encode the formula for  $\text{EF}\phi$

- Note that  $EF\phi = \phi \vee EXEF\phi$
- Thus,  $f^{EF\phi}$  is equivalent to  $f^\phi + f^{EXEF\phi}$ , which is equivalent to  $f^\phi + \exists X'.(f \rightarrow^* f^{EF\phi}[X=X'])$
- Since EF involves computing the least fixed point, we obtain
  - $f^{EF\phi} = uZ.(f^\phi + \exists X'.(f \rightarrow^* Z[X=X']))$ , where  $Z$  is a relational variable.
  - Thus, we are getting the least fixed point of the formula that precisely encodes  $EF\phi = \phi \vee EXEF\phi$
- The book provides similar coding for AF and EG on page 368
- The important point is to see how we used the fixed point characterization of CTL to code CTL formulas in relational mu-calculus (which has a fixed point syntax!)
- Thus, we can model check in terms of these relational mu-calculus formulas and symbolic representations of states and the transition relation

-Our last topic today, time-permitting, is to discuss a few abstraction techniques in model checking

-Abstraction methods are a family of techniques used to simplify automata.

-It is probably “the most important technique for reducing the state explosion problem.” –EM Clarke

-Aim: given model as an automata  $A$ , we reduce a complex problem of  $A \models \phi$  into a much simpler problem  $A' \models \phi$

-Thus, this is another layer of abstraction on top of the abstraction of specifying a model to represent the system in question

-We'll look more at examples to illustrate abstraction as opposed to developing a formal theory (for those interested, see me after class or email)

-Why/when abstraction? Automata (model) is too big to check, of model checker doesn't handle certain details of the model

- We'll look at 2 techniques
  - Abstraction by state merging
  - Cone of influence reduction
- Abstraction by state merging
  - View some states as identical (ie, notions of folding states)
  - Merged states are put together into a super-state
  - Merging can be used for verifying safety properties, mainly because
    - 1) the merged automata  $A'$  has more behaviors than  $A$
    - 2) the more behaviors an automata has, the fewer safety properties it fulfills
    - 3) thus, if  $A'$  satisfies a safety property  $p$ , then so too does  $A$  satisfy  $p$
    - 4) if  $A'$  doesn't satisfy  $p$ , *no* conclusion can be drawn about  $A$
  - Why is this verification only one-way?
- There is a difficulty here though:

-How are atomic propositions labeling states gathered together into the super-state??

-In principle: never merge states that are labeled with different sets of atomic props

-But this is way too restrictive

-How weaken?

-Turns out that if merging is used to check property  $p$ , then only the propositions occurring in  $p$  are relevant

-Thus, if a proposition  $X$  only appears in positive form in  $p$  (each occurrence of  $X$  is within an even # of negation symbols), then we can merge states w/o the need for these to agree on the presence of  $X$

-The super-state then carries the label of  $X$  iff all merged states carry the label  $X$

-This rationale isn't obvious though...

-Abstraction via cone of influence reduction

-Suppose we are given a subset of the variables  $V' \subseteq V$  that are of interest with respect to a required spec

- Recall: system can be specified as a Kripke Structure using equations for transition relations, and an equation for the initial set of states of the system
- We want to simplify the system description by referring to only those variables  $V'$
- But, values of  $V'$  variables may depend on the values of variables not in  $V'$ 
  - For example, we'll consider the modulo 8 counter that we examined in lecture 2
- We define the cone of influence  $C$  for  $V'$  and use  $C$  for our reduction of the system
- Defn: the cone of influence  $C$  of  $V'$  is the minimal set of vars s.t.
  - 1)  $V'$  is a subset of  $C$
  - 2) if for some  $v_1 \in C$  its formula  $f_1$  depends on  $v_j$ , then  $v_j$  is also in  $C$
- Therefore, the reduced system is constructed by removing all transition equations whose left hand side variables do not appear in  $C$

- We'll see the full example of this technique in class using the Kripke Structure model for the modulo 8 counter
- We won't, however, go over the proof arguing that removal of such equations doesn't affect the equivalency of the model