

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 ϕ
 - We denote this set as $\{\phi\}$
 - Our algorithm is recursive on the structure of ϕ
 - 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 μ and ν

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

-Where $uZ.f$ is recursively defined as

$$-u_0 Z.f = 0$$

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

- $p \models vZ.f$ 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 ϕ into a formula f^ϕ of the relational mu-calculus s.t. f^ϕ 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 ϕ

-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 ϕ as a function f^ϕ in relational mu-calculus is given inductively:

- $f^x = x$ for vars x

- $f^\perp = 0$

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

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

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

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

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

- 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