Abstraction and Interpolation

EECS 219C: Formal Methods

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Reviewing BMC

Given M = (S, I, R, L) and LTL property φ
We create a satisfiability instance that encodes

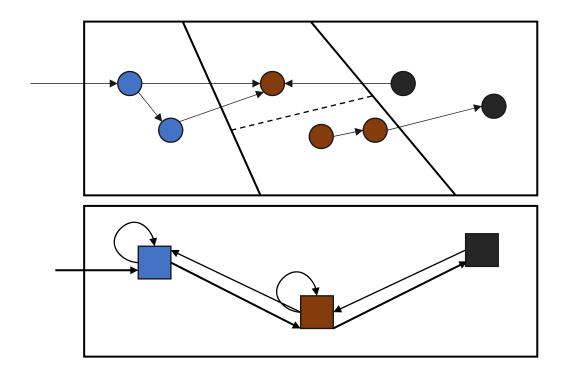
- System state transitions:
 - $I(s_0) \wedge R(s_0, s_1) \wedge ... \wedge R(s_1, s_{1+1}) \wedge ... \wedge R(s_{k-1}, s_k)$
- Set of loop-conditions: L(I, k)= R(s_k, s_l)
- Inductively constructed VC for negation of φ
 - $F \varphi : \pi_0^k \models F \varphi = S_0 \Rightarrow \varphi \lor \pi_1^k \models F \varphi$
 - $G\varphi: \pi_0^k \models G\varphi = s_0 \Rightarrow \varphi \land \pi_1^k \models G\varphi$; s_k must loop back to s_k

Reviewing Abstraction

Given $M = (S, I, R, L), h: S \to \hat{S}$ Suppose $\widehat{M} = (\hat{S}, \hat{I}, \widehat{R}, \widehat{L})$ where

- $\hat{I}(\hat{d})$ iff $\exists d. \left(h(d) = \hat{d} \land I(d)\right)$
- $\widehat{R}(\widehat{d_1}, \widehat{d_2})$ iff $\exists d_1 d_2. \left(h(d_1) = \widehat{d_1} \land h(d_2) = \widehat{d_2} \land R(d_1, d_2)\right)$
- $\hat{L}(\hat{d}) = \bigcup_{h(d)=\hat{d}} L(d)$

Abstraction Visualized



Cone-of-Influence Reduction

$$I \doteq x = 0 \land y = 0 \land z = 0 \land u = 0$$

$$R \doteq x' = x + y \land y' = y + z \land z' = z + 1 \land u' = u + 2$$

$$\phi \doteq G(x \ge 0)$$

Computing Col:

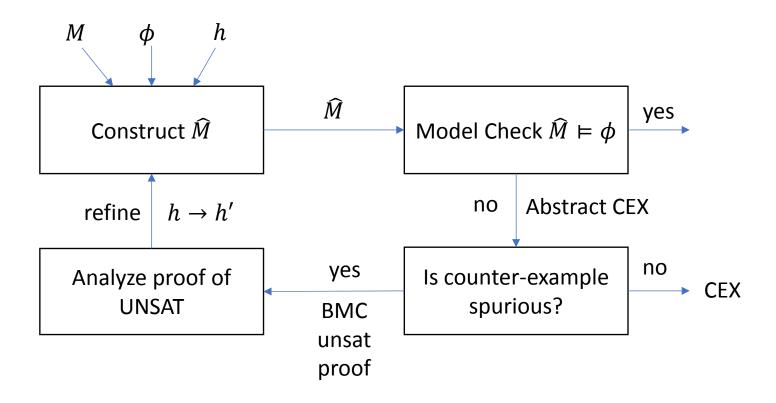
$$C_0 = \{x\}, C_1 = \{x, y\}, C_2 = \{x, y, z\}, C_3 = C_2$$

The abstraction is:

$$\hat{I} \doteq x = 0 \land y = 0 \land z = 0$$

$$\hat{R} \doteq x' = x + y \land y' = y + z \land z' = z + 1$$

Counter-example Guided Abstraction Refinement



CEGAR: Example

$$I \doteq x = 0 \land y = 0 \land z = 0 \land u = 0$$

$$R \doteq x' = x + y \land y' = y + ite(z \ge 0, z, -z) \land z' = z + 1$$

$$\phi \doteq G(x \ge 0)$$

$$\hat{I} \doteq x = 0$$

$$\hat{R} \doteq x' = x + y$$

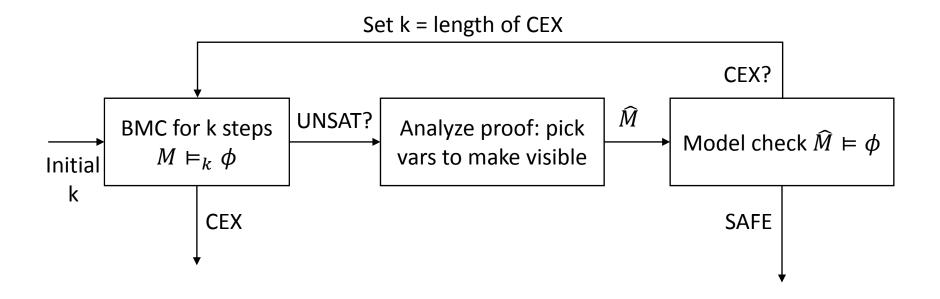
Abstract CEX: $(0, 0) \rightarrow (0, -1) \rightarrow (-1, 10)$

CEX is spurious, as y=0, y'=-1 is not possible, so we refine

$$\hat{I} \doteq x = 0 \land y = 0$$

$$\hat{R} \doteq x' = x + y \land y' = y + ite(z \ge 0, z, -z)$$

Proof-Based Abstraction



PBA: Example

$$I \doteq x = 0 \land y = 0 \land z = 0 \land u = 0$$

$$R \doteq x' = x + y \land y' = y + ite(z \ge 0, z, -z) \land z' = z + 1 \land u' = u + 2$$

$$\phi \doteq G(x \ge 0)$$

BMC for 2 steps gives us a proof involving x,y,z So we refine the abstraction to include these vars

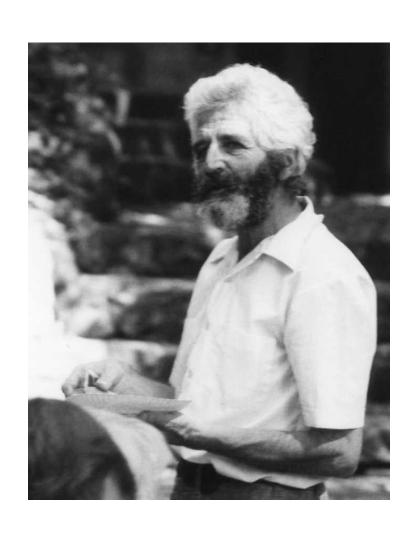
$$\hat{I} \doteq x = 0 \land y = 0 \land z = 0$$

$$\hat{R} \doteq x' = x + y \land y' = y + ite(z \ge 0, z, -z) \land z' = z + 1$$

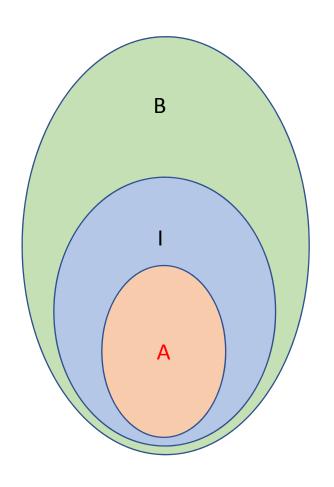
Complete model checking procedure proves ϕ on \widehat{M}

Unbounded Model Checking Using Craig Interpolants

William Craig (1918-2016)



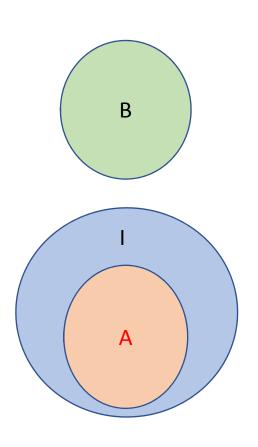
Craig Interpolants



For two formulas A and B such that $A \Rightarrow B$, a Craig interpolant is a formula I such that:

- 1. $A \Rightarrow I$
- 2. $I \Rightarrow B$
- 3. The nonlogical symbols in I occur in both A, B

Reverse Interpolants

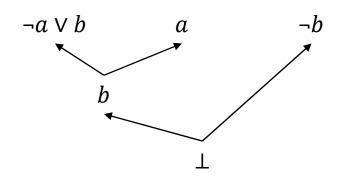


For two formulas A and B such that $A \wedge B$ is UNSAT, a reverse interpolant is a formula I such that:

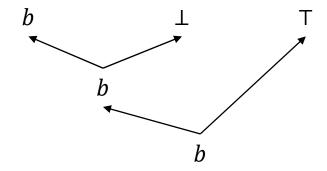
- 1. $A \Rightarrow I$
- 2. $I \wedge B$ is UNSAT
- 3. The nonlogical symbols in I occur in both A, B

Computing Interpolants from Resolution Proofs

Resolution proof for: $(\neg a \lor b)$, a, $\neg b$



Let $A = \{ (\neg a \lor b), a \} B = \{ \neg b \}$



Given a proof of UNSAT Π of $A \cup B$, for all vertices $c \in V_{\Pi}$ let p(c) be:

- If *c* is a root then:
 - If $c \in A$ then p(c) = g(c)
 - Else p(c) = T
- Else, let c_1 , c_2 be preds of c and let v be their pivot variable
 - If v is local to A, then $p(c) = p(c_1) \vee p(c_2)$
 - Else, $p(c) = p(c_1) \land p(c_2)$

Interpolant is $p(\bot)$

Interpolation based MC

k > 0; M = (Init, R, bad); P = Init

If $Init \wedge bad$ is SAT then return UNSAFE

repeat

•
$$M' = (P, R, bad)$$

$$PREF_h(M) = Init(s_{-l}) \land \bigwedge_{-h \le i \le 0} R(s_i, s_{i+1})$$

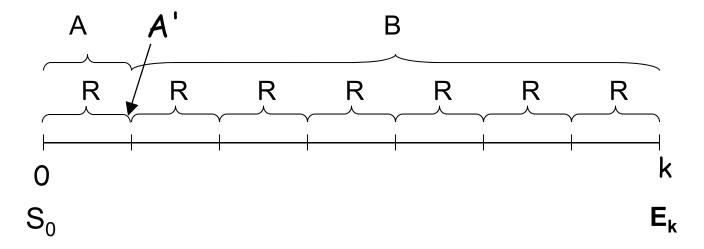
•
$$A = PREF_1(M')$$

•
$$B = SUFF_0^k(M')$$

$$SUFF_j^k(M) = \bigwedge_{0 \le i \le k} R(s_i, s_{i+1}) \wedge \bigwedge_{i \le i \le k} bad(s_i)$$

- If *A* ∧ *B* is SAT:
 - If P = Init return CEX
 - Else increase k; P = Init; continue
- Let Q be an interpolant for $A \wedge B$
- If $Q \Rightarrow P$ then **return** SAFE
- $P = P \vee Q$

Intuition



- A' tells us everything the prover deduced about the image of S₀ in proving it can't reach an error in k steps.
- Hence, A' is in some sense an abstraction of the image relative to the property *and* the bound k

The fixed point P is an inductive invariant

Refinement

- The procedure may be inconclusive for a fixed k
 - May add a state that reaches error in k steps (getting SAT in step 2 with $Z := S_0$)
- Refinement is just increasing k
 - How big can k get?

Interpolation based MC

For a fixed k:

- 1. Set Z initially to S_0
- 2. Do BMC starting from Z for k steps
 - If SAT: have we found a counterexample?
 - If UNSAT, continue
- Use interpolation to compute over-approximation of next states of Z and add them back into Z
 - Can newly added states lead to error states in k-1 steps? In k steps?
- 4. If Z does not increase
 - We've reached a fixed point Z=P. Is the property true?
- 5. Otherwise, back to step 2