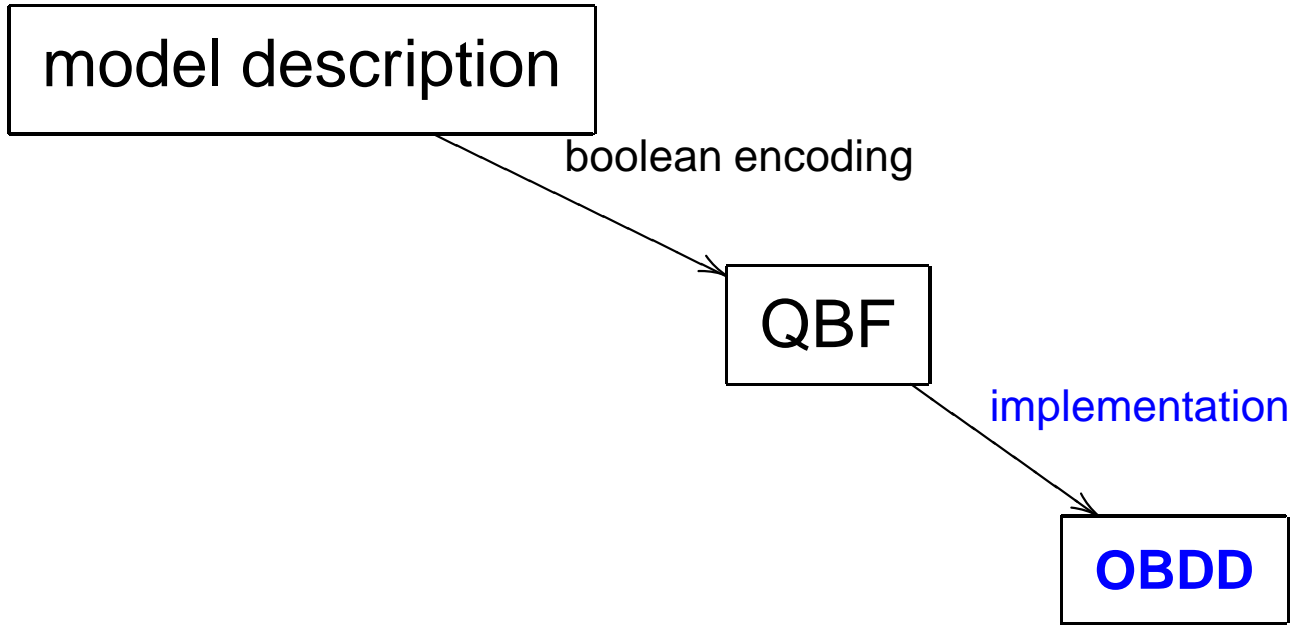
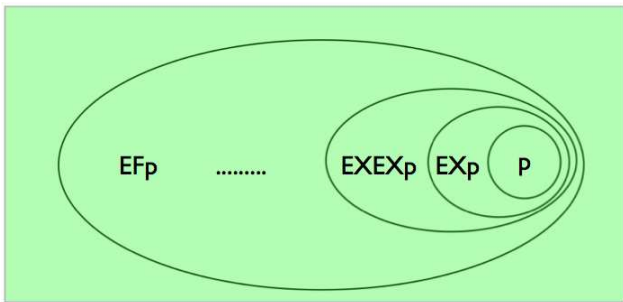


Computer aided verification

Lecture 8: Bounded model checking



model checking = operations on OBDDs



Example: $\text{EF } p$ (safety)

pre-SMC:

$$\dots \supseteq p \cup \mathbf{EX}(p \cup \mathbf{EX} p) \supseteq p \cup \mathbf{EX} p \supseteq p \supseteq \text{false}$$

$$\dots \supseteq p \cup \text{pre}(p \cup \text{pre}(p)) \supseteq p \cup \text{pre}(p) \supseteq p \supseteq \text{false}$$

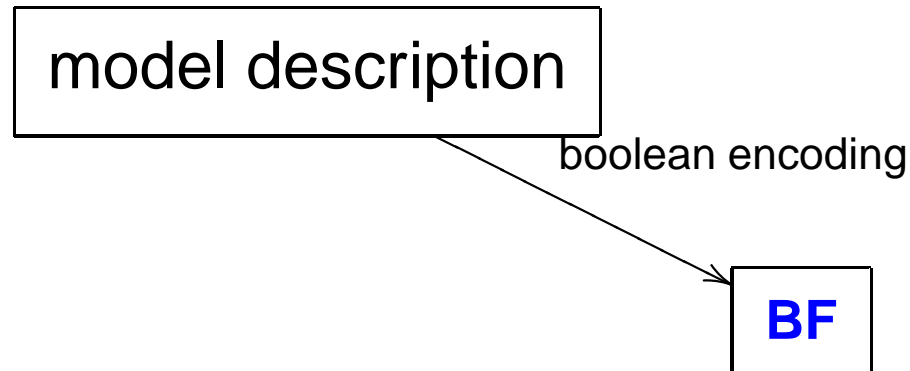
post-SMC:

$$S_0 \subseteq S_0 \cup \text{post}(S_0) \subseteq S_0 \cup \text{post}(S_0 \cup \text{post}(S_0)) \subseteq \dots$$

$$\exists \vec{z}_0 \exists \vec{z}_1 \dots \exists \vec{z}_k \quad \vec{z}_0 \longrightarrow \vec{z}_1 \longrightarrow \vec{z}_2 \longrightarrow \dots \longrightarrow \vec{z}_{k-1} \longrightarrow \vec{z}_k$$

$$\vec{z} = \vec{z}_0 \vee \vec{z} = \vec{z}_1 \vee \dots \vee \vec{z} = \vec{z}_k$$

Bounded model checking (BMC)



model checking = satisfiability of boolean formula

$$\vec{z}_0 \longrightarrow \vec{z}_1 \longrightarrow \vec{z}_2 \longrightarrow \dots \longrightarrow \vec{z}_{k-1} \longrightarrow \vec{z}_k$$

$$p(\vec{z}_0) \vee p(\vec{z}_1) \vee p(\vec{z}_2) \vee \dots \vee p(\vec{z}_k)$$

- searching for a counterexample of **bounded** length
- symbolic path
- application of **SAT-solvers**

I. BMC

$$M \models \mathbf{E} \phi \quad (\phi \in \mathbf{LTL}^+)$$

e.g.. instead of $M \models \mathbf{AG} \phi$, we check $M \models \mathbf{EF} \neg \phi$

M described by boolean formulas as usual:

$$R(\vec{z}, \vec{z}'), \quad L_p(\vec{z}), \quad S_0(\vec{z})$$

$$R(z_1, \dots, z_m, z'_1, \dots, z'_m), \quad L_p(z_1, \dots, z_m), \quad S_0(z_1, \dots, z_m)$$

Idea: Horizon $k \geq 0$

k system steps

$\Pi \vDash_k \phi$ k steps are sufficient to decide that $\Pi \vDash \phi$

Lem.: $\Pi \vDash_k \phi \implies \Pi \vDash \phi$

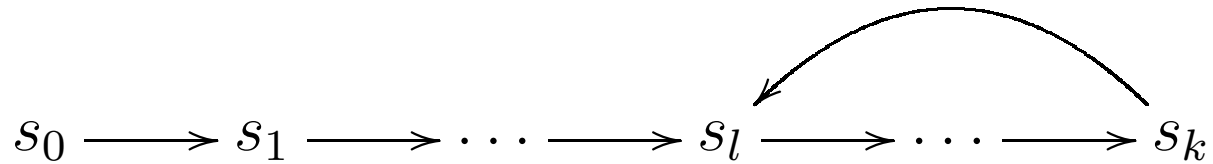
Lem.: $M \vDash \mathbf{E} \phi \implies \exists k \geq 0. M \vDash_k \mathbf{E} \phi$

loops!

Thm.: $M \vDash \mathbf{E} \phi \iff \exists k \geq 0. M \vDash_k \mathbf{E} \phi$

$$\Pi = s_0 \rightarrow s_1 \rightarrow s_2 \rightarrow \dots$$

(k, l) - loop:



$$s_{k+1+i} = s_{l+i}$$

(k) - loop)

if there exists a back loop from s_k

no-loop:



Bounded semantics for k -loops

$$\Pi \vDash_k \phi \iff \Pi \vDash \phi$$

$$\Pi = s_0 \rightarrow s_1 \rightarrow s_2 \rightarrow \dots$$

Bounded semantics for no-loops

$$\Pi \models_k \phi \iff \Pi \models_k^0 \phi$$

$$\Pi = s_0 \rightarrow s_1 \rightarrow s_2 \rightarrow \dots$$

$$\models_k^i, \quad 0 \leq i \leq k$$

Bounded semantics for no-loops

$$\Pi \models_k \phi \iff \Pi \models_k^0 \phi$$

$$\Pi = s_0 \rightarrow s_1 \rightarrow s_2 \rightarrow \dots$$

$$\models_k^i, \quad 0 \leq i \leq k$$

$$\Pi \models_k^i p \iff p \in L(s_i)$$

$$\Pi \models_k^i \neg p, \quad \phi_1 \wedge \phi_2, \quad \phi_1 \vee \phi_2 \iff \dots$$

$$\Pi \models_k^i \mathbf{X} \phi \iff i < k \wedge \Pi \models_k^{i+1} \phi$$

$$\Pi \models_k^i \mathbf{F} \phi \iff \exists j, i \leq j \leq k. \Pi \models_k^j \phi$$

$$\Pi \models_k^i \mathbf{G} \phi \quad \text{never}$$

$$\Pi \models_k^i \phi \mathbf{U} \psi \iff \exists j, i \leq j \leq k. \Pi \models_k^j \psi \wedge \forall l, i \leq l < j. \Pi \models_k^l \phi$$

$$\Pi \models_k^i \phi \mathbf{R} \psi \iff ?$$

Tw.: $M \models \mathbf{E} \phi \iff \exists k \geq 0. M \models_k \mathbf{E} \phi$

- what k is sufficiently big? graph diameter?
- no completeness !
- efficiency of SMC depends on graph diameter

$$M \vDash_k \mathbf{E} \phi \quad (\phi \in \mathbf{LTL}^+)$$

$$M, \phi, k \mapsto [M, \phi]_k$$

$$M \vDash_k \phi \iff [M, \phi]_k \text{ satisfiable}$$

M described by boolean formulas as usual:

$$R(\vec{z}, \vec{z}'), \quad L_p(\vec{z}), \quad S_0(\vec{z})$$

$$R(z_1, \dots, z_m, z'_1, \dots, z'_m), \quad L_p(z_1, \dots, z_m), \quad S_0(z_1, \dots, z_m)$$

Symbolic path

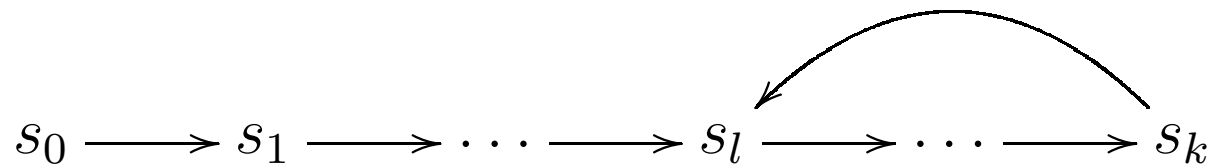


$$[M]_k := S_0(\vec{z}_0) \wedge \bigwedge_{i=0}^{k-1} R(\vec{z}_i, \vec{z}_{i+1})$$

$(k + 1) \cdot m$ boolean variables

$[M]_k$ is of size $\mathcal{O}(k \cdot |M|)$

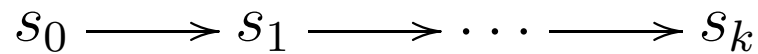
(k, l) -loop:



$${}_l L_k \equiv R(\vec{z}_k, \vec{z}_l)$$

$$L_k \equiv \bigvee_{l=0}^k {}_l L_k$$

no-loop:



$$[M, \phi]_k := [M]_k \wedge \left((\neg L_k \wedge [\phi]_k^0) \vee \bigvee_{l=0}^k (l L_k \wedge l[\phi]_k^0) \right)$$

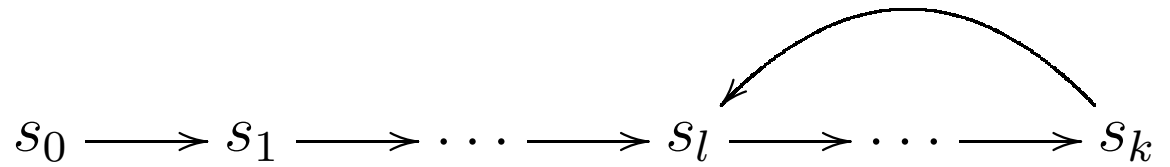
$l[\phi]_k^0$ – semantics of ϕ on (k, l) -loop

$[\phi]_k^0$ – semantics of ϕ on no-loop

$[M, \phi]_k$ is of size $\mathcal{O}(k \cdot (|M| + |\phi|))$

Boolean encoding

$$l[\phi]_k^i \quad 0 \leq l, i \leq k$$



$$l[p]_k^i \iff L_p(\vec{z}_i)$$

$$l[\neg p]_k^i \quad l[\phi \wedge \psi]_k^i$$

$$l[\phi \vee \psi]_k^i \iff l[\phi]_k^i \vee l[\psi]_k^i$$

$$l[X\phi]_k^i \iff l[\phi]_k^{\text{succ}(i)}$$

$$\text{succ}(i) = \begin{cases} i + 1 & i < k \\ l & i = k \end{cases}$$

$$\mathbf{F}\phi \equiv \phi \vee \mathbf{X}\mathbf{F}\phi$$

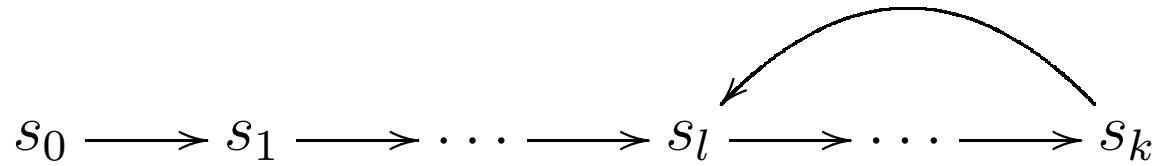
$$\phi \mathbf{U} \psi \equiv \psi \vee (\phi \wedge \mathbf{X}\phi \mathbf{U} \psi)$$

$$\mathbf{G}\phi \equiv \phi \wedge \mathbf{X}\mathbf{G}\phi$$

$$\phi \mathbf{R} \psi \equiv \psi \wedge (\phi \vee \mathbf{X}\phi \mathbf{R} \psi)$$

Boolean encoding

$$l[\phi]_k^i \quad 0 \leq l, i \leq k$$



$$l[p]_k^i \iff L_p(\vec{z}_i)$$

$$l[\neg p]_k^i \quad l[\phi \wedge \psi]_k^i \quad l[\phi \vee \psi]_k^i \iff l[\phi]_k^i \vee l[\psi]_k^i$$

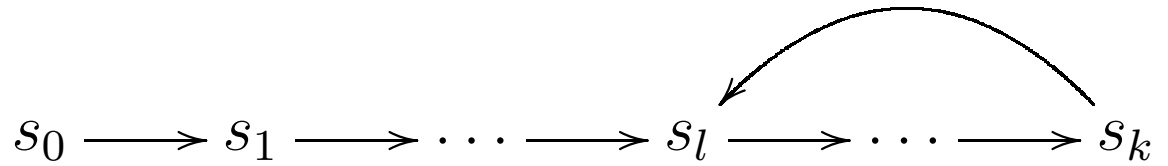
$$l[X \phi]_k^i \iff l[\phi]_k^{\text{succ}(i)} \quad \text{succ}(i) = \begin{cases} i + 1 & i < k \\ l & i = k \end{cases}$$

$$l[F \phi]_k^i \iff l[\phi]_k^i \vee l[F \phi]_k^{\text{succ}(i)}$$

$$l[G \phi]_k^i \iff l[\phi]_k^i \wedge l[G \phi]_k^{\text{succ}(i)}$$

$$l[\phi U \psi]_k^i \iff l[\psi]_k^i \vee (l[\phi]_k^i \wedge l[\phi U \psi]_k^{\text{succ}(i)})$$

Example:



$${}_l[\mathbf{F} p]_k^0 \iff L_p(\vec{z}_0) \vee \dots \vee L_p(\vec{z}_k)$$

$${}_l[\mathbf{G} p]_k^0 \iff L_p(\vec{z}_0) \wedge \dots \wedge L_p(\vec{z}_k)$$

$${}_l[p \mathbf{U} q]_k^0 \iff L_q(\vec{z}_0) \vee (L_p(\vec{z}_0) \wedge (L_q(\vec{z}_1) \vee (\dots \\ L_q(\vec{z}_{k-1}) \vee (L_p(\vec{z}_{k-1}) \wedge L_q(\vec{z}_k)) \dots))))$$

$${}_l[\mathbf{F} \mathbf{G} p]_k^0 \iff (L_p(\vec{z}_0) \wedge L_p(\vec{z}_1) \wedge \dots \wedge L_p(\vec{z}_k)) \vee \\ (L_p(\vec{z}_1) \wedge \dots \wedge L_p(\vec{z}_k)) \vee \dots \vee (L_p(\vec{z}_l) \wedge \dots \wedge L_p(\vec{z}_k)) \\ \iff (L_p(\vec{z}_l) \wedge \dots \wedge L_p(\vec{z}_k))$$

$I[\phi]_k^0$ is of size $\mathcal{O}(k \cdot |\phi|)$

Boolean encoding

$$[\phi]_k^i \quad 0 \leq i \leq k$$

$$s_0 \longrightarrow s_1 \longrightarrow \dots \longrightarrow s_k$$

$$[p]_k^i \quad [\neg p]_k^i \quad [\phi \wedge \psi]_k^i \quad [\phi \vee \psi]_k^i \iff \dots \text{ j. w.}$$

$$[X\phi]_k^i \iff [\phi]_k^{i+1}$$

$$[\phi]_k^{k+1} \iff \text{false}$$

$$\mathbf{F}\phi \equiv \phi \vee \mathbf{X}\mathbf{F}\phi$$

$$\phi \mathbf{U}\psi \equiv \psi \vee (\phi \wedge \mathbf{X}\phi \mathbf{U}\psi)$$

$$\mathbf{G}\phi \equiv \phi \wedge \mathbf{X}\mathbf{G}\phi$$

$$\phi \mathbf{R}\psi \equiv \psi \wedge (\phi \vee \mathbf{X}\phi \mathbf{R}\psi)$$

Example:

$$s_0 \longrightarrow s_1 \longrightarrow \dots \longrightarrow s_k$$

$$[\mathbf{F} p]_k^0 \iff L_p(\vec{z}_0) \vee \dots \vee L_p(\vec{z}_k) \vee \mathbf{false}$$

$$[\mathbf{G} p]_k^0 \iff L_p(\vec{z}_0) \wedge \dots \wedge L_p(\vec{z}_k) \wedge \mathbf{false} \equiv \mathbf{false}$$

$$[p \mathbf{U} q]_k^0 \iff L_q(\vec{z}_0) \vee (L_p(\vec{z}_0) \wedge (L_q(\vec{z}_1) \vee (\dots \\ L_q(\vec{z}_{k-1}) \vee (L_p(\vec{z}_{k-1}) \wedge (L_q(\vec{z}_k) \vee (L_p(\vec{z}_k) \wedge \mathbf{false})))) \dots)))$$

$$[\mathbf{F} \mathbf{G} p]_k^0 \iff \mathbf{false}$$

$$[M, \phi]_k := [M]_k \wedge \left((\neg L_k \wedge [\phi]_k^0) \vee \bigvee_{l=0}^k ({}_l L_k \wedge {}_l [\phi]_k^0) \right)$$

Thm.: $M \models_k \mathbf{E} \phi \iff [M, \phi]_k$ is satisfiable.

Repeat:

(1) $k := k_0$

(2) if $[M, \phi]_k$ satisfiable then stop

$M \models \mathbf{E} \phi$

(3) increment k and continue

Other boolean encodings

- hardware verification:
 - circuit structure kept, designated SAT-solvers
- LTL with past
- software verification (e.g. CBMC)
 - loop unfolding, heap model
- concurrent programs
- ...



II. Completeness ?

Is completeness achievable ?

- completeness threshold for k (safety $G p$)
- simultaneous checks of ϕ and $\neg\phi$ (liveness $F p$)
- induction (safety $G p$)

Reachability diameter – bound for k

the least i such that

$$\forall \vec{z}_0, \dots, \vec{z}_n. \exists \vec{z}'_0, \dots, \vec{z}'_t, t \leq i. S_0(\vec{z}_0) \wedge \bigwedge_{j=0}^{n-1} R(\vec{z}_j, \vec{z}_{j+1}) \implies$$

$$S_0(\vec{z}'_0) \wedge \left(\bigwedge_{j=0}^{t-1} R(\vec{z}'_j, \vec{z}'_{j+1}) \right) \wedge \vec{z}'_t = \vec{z}_n$$

Reachability diameter – bound for k

the least i such that

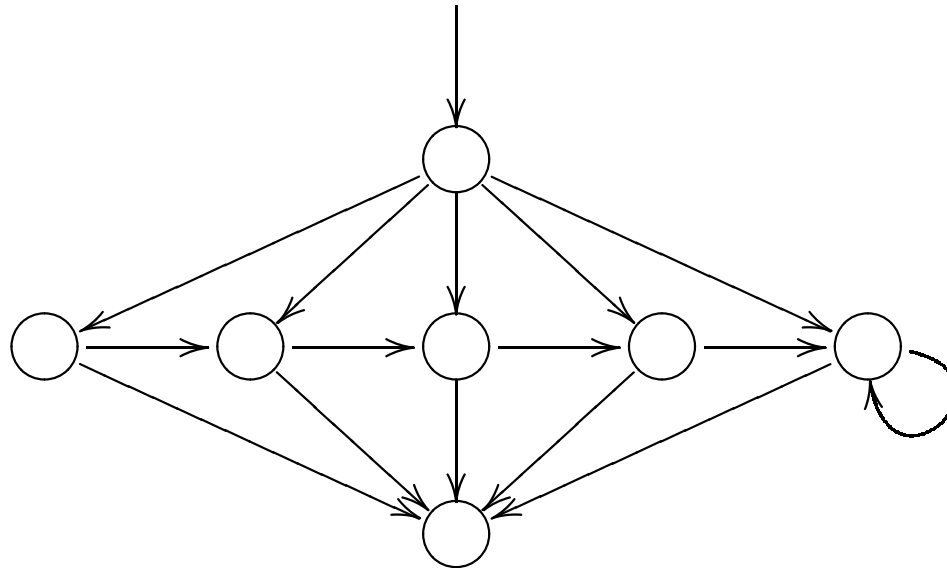
$$\forall \vec{z}_0, \dots, \vec{z}_{i+1}. \exists \vec{z}'_0, \dots, \vec{z}'_i. S_0(\vec{z}_0) \wedge \bigwedge_{j=0}^i R(\vec{z}_j, \vec{z}_{j+1}) \implies$$

$$S_0(\vec{z}'_0) \wedge \left(\bigwedge_{j=0}^{i-1} R(\vec{z}'_j, \vec{z}'_{j+1}) \right) \wedge \bigvee_{j=0}^i \vec{z}'_j = \vec{z}_{i+1}$$

the longest loop-free path – bound for k

the greatest i such that

$$\exists \vec{z}'_0, \dots, \vec{z}_i. S_0(\vec{z}_0) \wedge \left(\bigwedge_{j=0}^{i-1} R(\vec{z}_j, \vec{z}_{j+1}) \right) \wedge \bigwedge_{j=0}^{i-1} \bigwedge_{l=j+1}^i \vec{z}_j \neq \vec{z}_l$$



$M \models \mathbf{EG} \neg p \iff \exists k$ s.t. the following formula is satisfiable:

$$\mathbf{No}_k(\vec{z}_0, \dots, \vec{z}_{k+1}) \equiv S_0(\vec{z}_0) \wedge \bigwedge_{j=0}^{k-1} R(\vec{z}_j, \vec{z}_{j+1}) \wedge \left(\bigvee_{l=0}^k l L_k \right) \wedge \bigwedge_{j=0}^k \neg L_p(\vec{z}_j)$$

$M \models \mathbf{AF} p \iff \exists k$ s.t. the following formula is a tautology:

$$\mathbf{Yes}_k(\vec{z}_0, \dots, \vec{z}_{k+1}) \equiv S_0(\vec{z}_0) \wedge \bigwedge_{j=0}^{k-1} R(\vec{z}_j, \vec{z}_{j+1}) \implies \bigvee_{j=0}^k L_p(\vec{z}_j)$$

$M \models \mathbf{EG} \neg p \iff \exists k$ s.t. the following formula is satisfiable:

$$\mathbf{No}_k(\vec{z}_0, \dots, \vec{z}_{k+1}) \equiv S_0(\vec{z}_0) \wedge \bigwedge_{j=0}^{k-1} R(\vec{z}_j, \vec{z}_{j+1}) \wedge \left(\bigvee_{l=0}^k lL_k \right) \wedge \bigwedge_{j=0}^k \neg L_p(\vec{z}_j)$$

$M \models \mathbf{AF} p \iff \exists k$ s.t. the following formula is unsatisfiable:

$$\neg \mathbf{Yes}_k(\vec{z}_0, \dots, \vec{z}_{k+1}) \equiv S_0(\vec{z}_0) \wedge \bigwedge_{j=0}^{k-1} R(\vec{z}_j, \vec{z}_{j+1}) \wedge \bigwedge_{j=0}^k \neg L_p(\vec{z}_j)$$

Repeat:

(1) $k := 0$

(2) if No_k satisfiable then stop

(3) if $\neg Yes_k$ unsatisfiable then stop

(4) increase k and continue

$M \models EG \neg \phi$

$M \models Fp$

Completeness for $G p$ – induction

– induction base

check that the following formula is unsatisfiable:

$$\mathbf{Base}_k(\vec{z}_0, \dots, \vec{z}_k) \equiv S_0(\vec{z}_0) \wedge \bigwedge_{j=0}^{k-1} R(\vec{z}_j, \vec{z}_{j+1}) \wedge \bigvee_{j=0}^k \neg L_p(\vec{z}_j)$$

– induction step

check that the following formula is unsatisfiable:

$$\mathbf{Step}_k(\vec{z}_0, \dots, \vec{z}_{k+1}) \equiv \bigwedge_{j=0}^k (L_p(\vec{z}_j) \wedge R(\vec{z}_j, \vec{z}_{j+1})) \wedge \neg L_p(\vec{z}_{k+1})$$

Why do we need $k > 1$?

Repeat:

(1) $k := 0$

(2) if Base_k satisfiable then stop

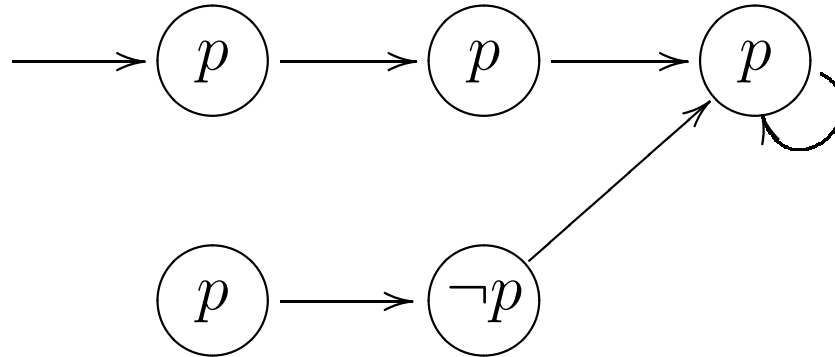
$$M \models \text{EF } \neg \phi$$

(3) if Step_k unsatisfiable then stop

$$M \models \text{G } p$$

(4) increase k and continue

Why do we need $k > 1$?



Repeat:

(1) $k := 0$

(2) if Base_k satisfiable then stop

$$M \models \text{EF } \neg \phi$$

(3) if Step_k unsatisfiable then stop

$$M \models \text{G } p$$

(4) increase k and continue

III. BDD or SAT ?

Comparison

bit	k	SMV ₂	MB	PROVER	MB
0	1	25	79	< 1	1
1	2	25	79	< 1	1
2	3	26	80	< 1	1
3	4	27	82	1	2
4	5	33	92	1	2
5	6	67	102	1	2
6	7	258	172	2	2
7	8	1741	492	7	3
8	9		> 1GB	29	3
9	10			58	3
10	11			91	3
11	12			125	3
12	13			156	4
13	14			186	4
14	15			226	4
15	16			183	5

[Biere, Cimatti, Clarke, Strichman, Zhu 2003]

- the methods seem complementary
- SAT more predictable
- BDD is memory-consuming and SAT time-consuming
- BDD is complete, SAT no
- unbounded model checking UMC:
 - CNF instead of BDD in SMC
- BDD + SAT

1981–1982: EMC — 10^5 states

1990–1992: SMV — 10^{100} states

2000': BMC + CEGAR — 10^{1000} states

IV. SAT-solvers

- CNF
- algorithm DPLL
 - searching through partial valuations tree
 - Boolean Constraint Propagation (BCP)
 - **conflicts** – pruning of the tree
- heuristics

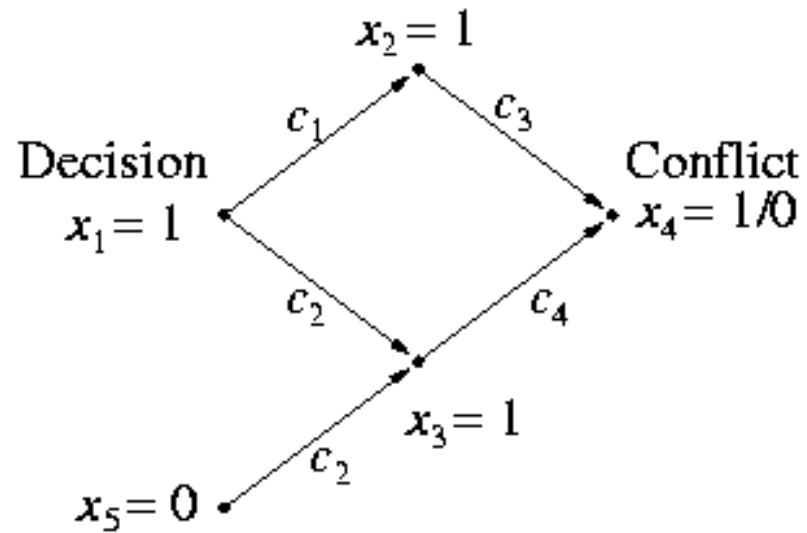
Implications graph and BCP

$$c_1 = (\neg x_1 \vee x_2)$$

$$c_2 = (\neg x_1 \vee x_3 \vee x_5)$$

$$c_3 = (\neg x_2 \vee x_4)$$

$$c_4 = (\neg x_3 \vee \neg x_4)$$



[Biere, Cimatti, Clarke, Strichman, Zhu 2003]

```
// Input arg: Current decision level  $d$ 
// Return value:
//   SAT():      {SAT, UNSAT}
//   Decide():   {DECISION, ALL-DECIDED}
//   Deduce():   {OK, CONFLICT}
//   Diagnose(): {SWAP, BACK-TRACK} also calculates  $\beta$ 
```

```
SAT ( $d$ )
{
 $l_1$ :   if (Decide ( $d$ ) == ALL-DECIDED) return SAT;
 $l_2$ :   while (TRUE) {
 $l_3$ :     if (Deduce( $d$ ) != CONFLICT) {
 $l_4$ :       if (SAT ( $d+1$ ) == SAT) return SAT;
 $l_5$ :       else if ( $\beta < d$  ||  $d == 0$ )
 $l_6$ :         { Erase ( $d$ ); return UNSAT; }
      }
 $l_7$ :     if (Diagnose ( $d$ ) == BACK-TRACK) return UNSAT;
    }
}
```

[Biere, Cimatti, Clarke, Strichman, Zhu 2003]

Why SAT-solvers are so fast?

- conflict learning – adding conflict clauses
- non-chronological backtracking
- heuristics for decisions
- efficient data structures
- incremental satisfiability