

# Seminar 5: SMT & SPIN

CEGAR Refinement Step, SMT, Promela, SPIN

Di Marco, Okwieka

University of Rome "La Sapienza"

May 29th, 2023

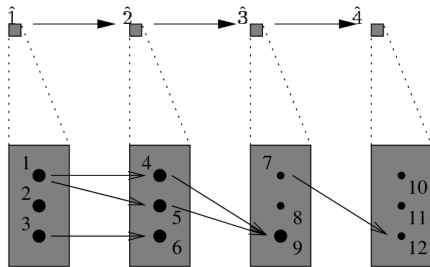
## CEGAR Refinement Step

# SplitPATH: Identifying spurious finite path counterexamples

Clarke et al. 2003 (figure and pseudocode copied verbatim)

Define  $h^{-1}(\hat{s}) = \{s \mid h(s) = \hat{s}\}$  and lift to paths  $\hat{T} = \langle \hat{s}_1, \dots, \hat{s}_n \rangle$

Concrete counterexample  $\iff h_{path}^{-1}(\hat{T}) \neq \emptyset \iff S_i \neq \emptyset$



## Algorithm SplitPATH

$S := h^{-1}(\hat{s}_1) \cap I$

$j := 1$

**while** ( $S \neq \emptyset$  and  $j < n$ ) {  
     $j := j + 1$   
     $S_{\text{prev}} := S$   
     $S := \text{Img}(S, R) \cap h^{-1}(\hat{s}_j)$  }

**if**  $S \neq \emptyset$  **then** output counterexample

**else** output  $j, S_{\text{prev}}$

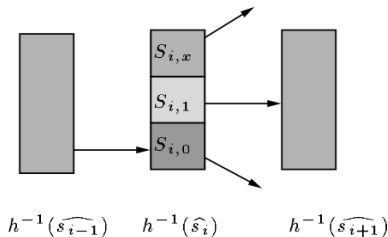
# PolyRefine: Refinement Step

Clarke et al. 2003 (figure copied verbatim)

For  $i$  so that  $Img(S_i, R) \cap h^{-1}(\widehat{s_{i+1}}) = \emptyset$  and  $S_i$  reachable, we partition  $h^{-1}(\widehat{s_i})$  into  $S_{i,0}$  (reachable),  $S_{i,1}$  (outgoing),  $S_{i,x}$  (isolated)

$$\text{Spurious } \widehat{s_i} \rightarrow \widehat{s_{i+1}} \iff S_{i,1} \neq \emptyset$$

Split abstraction relation  $\equiv$  into coarsest  $\equiv'$  s.t.  $S_{i,0}$  and  $S_{i,1} \cup S_{i,x}$  are separate.  
Can be done in polynomial time if no need for optimal (coarsest) refinement.



# PolyRefine: Refinement Step

Clarke et al. 2003

For every equivalence class  $h^{-1}(\hat{s}_j) = E_j$  of abstraction  $\equiv$ , define  $\equiv_j$  as equivalence relation with one class,  $E_j$ .

Want to find refinement for each. Initially  $\equiv'_j \leftarrow \equiv_j$ .

For all  $a, b \in E_j$ : if  $a \in S_{i,0}$  and  $b \in S_{i,1} \cup S_{i,x}$ , remove  $(a, b)$  from  $\equiv'_j$ .

$\implies$  Splits  $E_j$  into two subclasses, assign each to new abstract state.

SMT

# Satisfiability Modulo Theories

**Classic SAT:** “find Boolean variable assignment s.t. Boolean formula holds”

Conjunctive Normal Form of OR-*clauses* of *literals* connected with AND:

$$(x_1 \vee \neg x_2 \vee \dots \vee x_n) \wedge (x_{n+1} \vee \dots \vee \neg x_{n+m}) \wedge \dots$$

**Literal in Boolean formula:** variable or its negation

**Literal in SMT formula:** formula in quantifier-free First-Order Logic (FOL) theory, e.g. Linear Rational Arithmetic:

$$(\neg(2v_1 + v_2 \leq 3) \vee A_1) \wedge (3v_2 - 2v_1 < 6) \vee \dots$$

# Quantifier-free First-Order (= Predicate) Logic and Theories

**FOL:** non-Boolean (“domain”) variables, predicates over them, quantifiers

**FOL Signature  $\Sigma$ :** set of symbols (functions, predicates, constants)

**$\Sigma$ -FOL Theory  $\mathcal{T}$ :** set of  $\Sigma$ -formulas constrained by *axioms*

$\implies$  axioms *interpret* the symbols of the signature

**Decidable Theory:** there is an efficient procedure for “Is a formula included in the theory?”

Usually more interested in “Is this conjunction of  $\mathcal{T}$ -literals *consistent*?”



# Examples of Decidable FOL Theories

## Equality and Uninterpreted Functions

$$f(g(x, y)) = h(y, f(x))$$

- Equality axioms  $a = a$  and  $a = b \iff b = a$  and  $a = b \wedge b = c \implies a = c$
- Uninterpreted functions: only defined by name and arity (argument number)

## Linear Arithmetic over the Integers

$$(a + b < 3) \wedge (b = 2) \wedge (a > 0) \implies (a = 1)$$

- Clauses are (in-)equalities of arithmetic expressions
- Arithmetic expressions: constant numbers, operators, variables

# Building a SMT Solver: Lazy Approach

**SMT formula**  $\varphi$   $\mathcal{T}$ -SAT  $\iff \varphi^p \wedge \tau^p$  SAT where

- $\varphi^p$  Boolean abstraction of  $\varphi$ : every  $\mathcal{T}$ -atom corresponds to a Boolean variable  $B_i$
- $\tau^p$  Boolean abstraction of all  $\mathcal{T}$ -lemmas on atoms in  $\varphi$  (not explicitly constructed as Boolean formula)

**SAT solver:** find  $\mu^p \models \varphi^p$ , then ask  $\mathcal{T}$ -solver whether  $\mu^p \models \tau^p$

**$\mathcal{T}$ -solver( $\mu$ ):** determine consistency of  $\mathcal{T}$ -literals conjunction  $\mu$ , give *conflict set*  $\eta \subseteq \mu$  if inconsistent (i.e. return falsified clauses)

# Classic SAT: DPLL Algorithm

Backtracking algorithm with eager rules reducing search space. Observations:

$$(0 \vee x_2 \vee \dots \vee x_n) \text{ SAT} \iff (x_2 \vee \dots \vee x_n) \text{ SAT} \quad (1)$$

$$(1 \vee x_2 \vee \dots \vee x_n) \text{ SAT trivially} \quad (2)$$

**Unit propagation:** for unit clause  $x_i$ , assign  $x_i \leftarrow 1$ . Conversely for  $\neg x_i$ .

Remove all clauses containing this literal (2), remove complement from all other clauses (1).

**Pure literal elimination:** if variable  $x_i$  only occurs as literal  $x_i$  or as  $\neg x_i$  throughout formula, assign it appropriately and drop containing clauses (2).

**Termination:** SAT if all clauses satisfied, UNSAT if there is an unsatisfied clause for all variable assignments (exhaustive search)

# Classic SAT: DPLL Algorithm Example (1.0)

**Action to do:** Choose  $a = 0$ .

$$c_1 : \neg a \vee b \vee c$$

$$c_2 : a \vee c \vee d$$

$$c_3 : a \vee c \vee \neg d$$

$$c_4 : a \vee \neg c \vee d$$

$$c_5 : a \vee \neg c \vee \neg d$$

$$c_6 : \neg b \vee \neg c \vee d$$

$$c_7 : \neg a \vee b \vee \neg c$$

$$c_8 : \neg a \vee \neg b \vee c$$

# Classic SAT: DPLL Algorithm Example (1.1)

$$c_2 : c \vee d$$

$$c_3 : c \vee \neg d$$

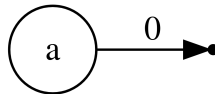
$$c_4 : \neg c \vee d$$

$$c_5 : \neg c \vee \neg d$$

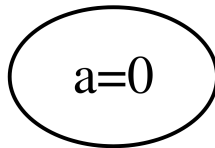
$$c_6 : \neg b \vee \neg c \vee d$$

**Action taken:** Choose  $a = 0$ .

**Tree:**



**Implication Graph:**



# Classic SAT: DPLL Algorithm Example (1.2)

$$c_2 : c \vee d$$

$$c_3 : c \vee \neg d$$

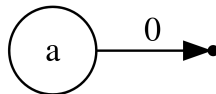
$$c_4 : \neg c \vee d$$

$$c_5 : \neg c \vee \neg d$$

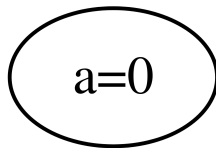
$$c_6 : \neg b \vee \neg c \vee d$$

**Action to do:** Pure literal elim.  $b = 0$ .

**Tree:**



**Implication Graph:**



# Classic SAT: DPLL Algorithm Example (1.3)

$$c_2 : c \vee d$$

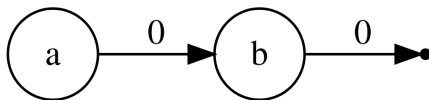
$$c_3 : c \vee \neg d$$

$$c_4 : \neg c \vee d$$

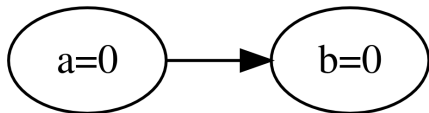
$$c_5 : \neg c \vee \neg d$$

**Action taken:** Pure literal elim.  $b = 0$ .

**Tree:**



**Implication Graph:**



# Classic SAT: DPLL Algorithm Example (2.0)

$$c_2 : c \vee d$$

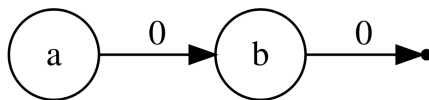
$$c_3 : c \vee \neg d$$

$$c_4 : \neg c \vee d$$

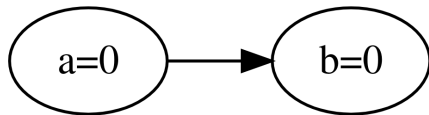
$$c_5 : \neg c \vee \neg d$$

**Action to do:** Choose  $c = 0$ .

**Tree:**



**Implication Graph:**





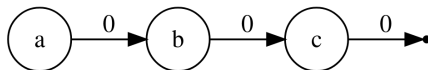
# Classic SAT: DPLL Algorithm Example (2.1)

$c_2 : d$

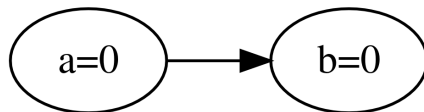
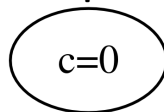
$c_3 : \neg d$

**Action taken:** Choose  $c = 0$ .

**Tree:**



**Implication Graph:**



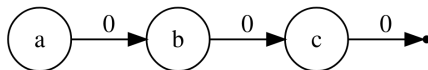
# Classic SAT: DPLL Algorithm Example (2.2)

$c_2 : d$

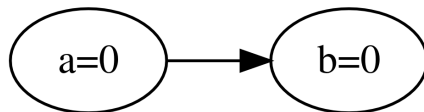
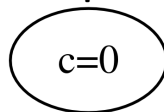
$c_3 : \neg d$

**Actions to do:** Unit prop.  $d = 1$  and  $d = 0$ .

**Tree:**



**Implication Graph:**



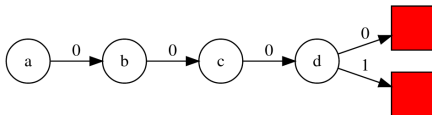
# Classic SAT: DPLL Algorithm Example (2.3)

$c_4 : d$

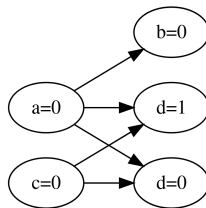
$c_5 : \neg d$

**Actions not taken:** Unit prop.  $d = 1$  and  $d = 0$ .

**Tree:**



**Implication Graph:**



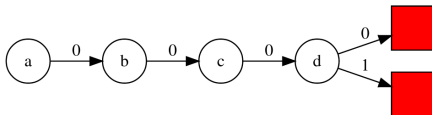
# Classic SAT: DPLL Algorithm Example (2.4)

$c_2 : d$

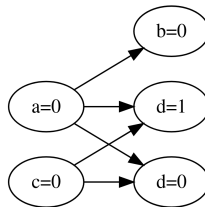
$c_3 : \neg d$

**Action to do:** Backtrack to last step.

**Tree:**



**Implication Graph:**



# Classic SAT: DPLL Algorithm Example (3.0)

$$c_2 : c \vee d$$

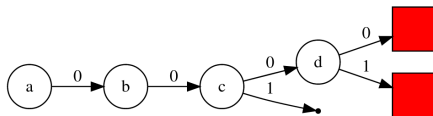
$$c_3 : c \vee \neg d$$

$$c_4 : \neg c \vee d$$

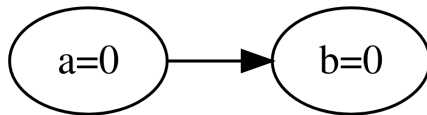
$$c_5 : \neg c \vee \neg d$$

**Action to do:** Choose  $c = 1$ .

**Tree:**



**Implication Graph:**



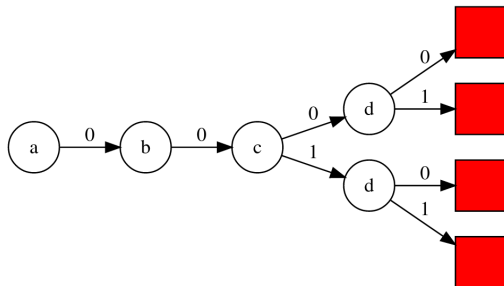
# Classic SAT: DPLL Algorithm Example (3.x)

$c_4 : d$

$c_5 : \neg d$

**Actions taken:** Choose  $c = 1$ , and continue (omitted).

**Tree:**



# Classic SAT: DPLL Algorithm Example (4.0)

$$c_1 : \neg a \vee b \vee c$$

$$c_2 : a \vee c \vee d$$

$$c_3 : a \vee c \vee \neg d$$

$$c_4 : a \vee \neg c \vee d$$

$$c_5 : a \vee \neg c \vee \neg d$$

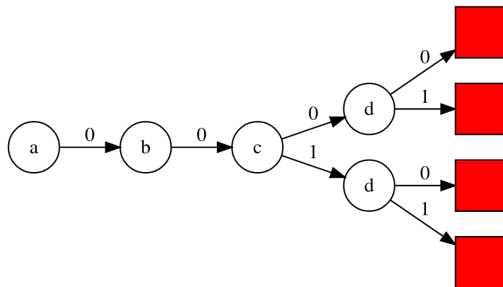
$$c_6 : \neg b \vee \neg c \vee d$$

$$c_7 : \neg a \vee b \vee \neg c$$

$$c_8 : \neg a \vee \neg b \vee c$$

**Actions taken:** Backtrack to  $a$ , since  $b$  pure.

**Tree:**



# Classic SAT: DPLL Algorithm Example (4.1)

$$c_1 : b \vee c$$

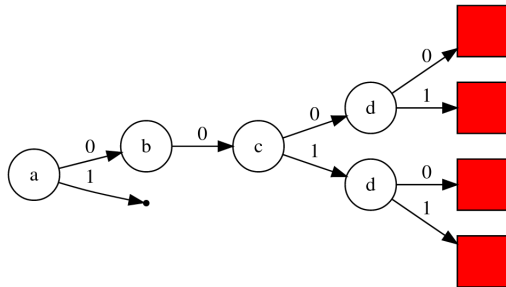
$$c_6 : \neg b \vee \neg c \vee d$$

$$c_7 : b \vee \neg c$$

$$c_8 : \neg b \vee c$$

**Actions taken:** Choose  $a = 1$ .

**Tree:**





# Classic SAT: DPLL Algorithm Example (4.2)

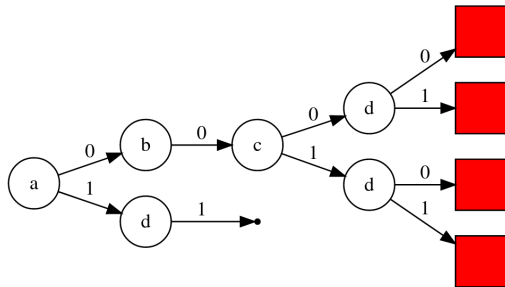
$$c_1 : b \vee c$$

$$c_7 : b \vee \neg c$$

$$c_8 : \neg b \vee c$$

**Actions taken:** Pure literal elim.  $d = 1$ .

**Tree:**



# Classic SAT: DPLL Algorithm Example (5.x)

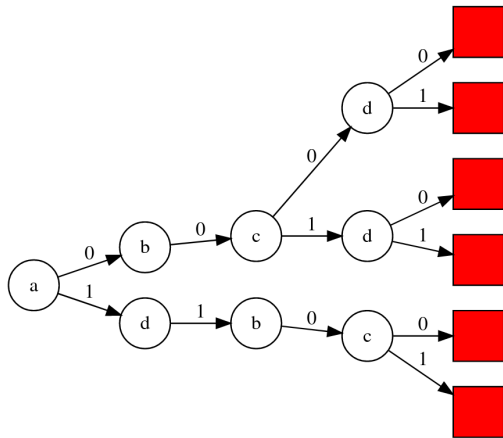
$$c_1 : b \vee c$$

$$c_7 : b \vee \neg c$$

$$c_8 : \neg b \vee c$$

**Actions taken:** Choose  $b = 0$ , see it is unsat., backtrack.

**Tree:**

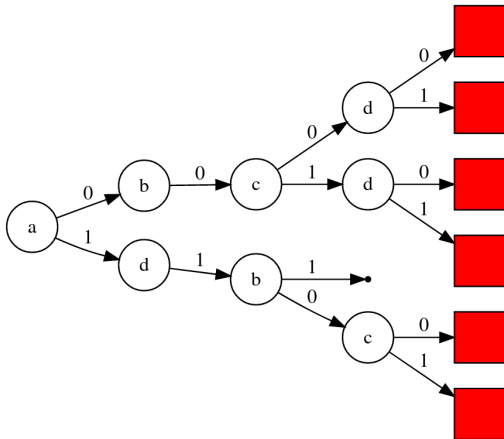


# Classic SAT: DPLL Algorithm Example (6.1)

**Action taken:** Choose  $b = 1$ .

**Tree:**

$C_8 : C$

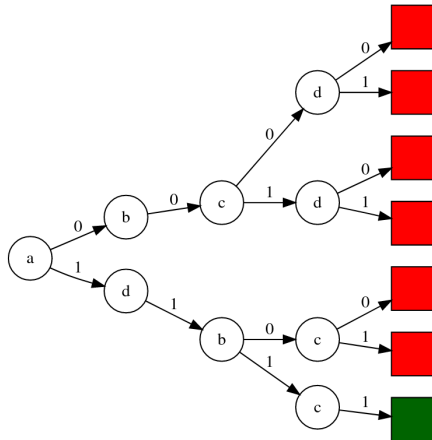


# Classic SAT: DPLL Algorithm Example (6.1)

**Action taken:** Unit prop.  $c = 1$ .

**Tree:**

$C_8 : C$



# Classic SAT: Conflict-Driven Clause Learning (CDCL)

## Improvement on DPLL

- 1 Variable choosing, unit propagation, pure literal elim. as before.
- 2 Construct implication graph to find conflicts.
- 3 If conflict, find responsible assignments (predecessors of conflicting states).
- 4 Add negation of these assignments as new *conflict clause*.
- 5 *Non-chronological backjumping*: Jump back in tree as far as possible to a responsible assignment.

# From CDCL to CDCL( $\mathcal{T}$ )

Early pruning: call  $\mathcal{T}$ -solver for partial assignments

Then:  $\mathcal{T}$ -solver( $\mu$ ) should be incremental (reuse  $\mu_1$  results for  $\mu_1 \cup \mu_2$  computation) and backtrack-capable (undo)  $\implies$  Stack-based interface

$\mathcal{T}$ -solver( $\mu$ ) returns conflict set  $\eta \subseteq \mu \implies$  conflict clauses

$\mathcal{T}$ -propagation:  $\mathcal{T}$ -solver( $\mu$ ) can give deduction clauses  $\neg\mu' \models \eta$  where  $\mu' \subseteq \mu$  causes assignment  $\eta$  to previously unassigned atom in  $\varphi$ .

Many more advanced techniques (pure-literal filtering, preprocessing atoms)

SPIN

# Simple Promela Interpreter

- SPIN can be used for four main purposes:
  - ① as a **simulator**, allowing for rapid prototyping with a random, guided, or interactive simulations
  - ② as an exhaustive **verifier**, capable of proving the validity of user specified LTL properties
  - ③ as a **proof approximation system** that can validate large models with maximal coverage of the state space.
  - ④ as a driver for **swarm verification**, which can make optimal use of large numbers of available compute cores to leverage parallelism and search diversification techniques,



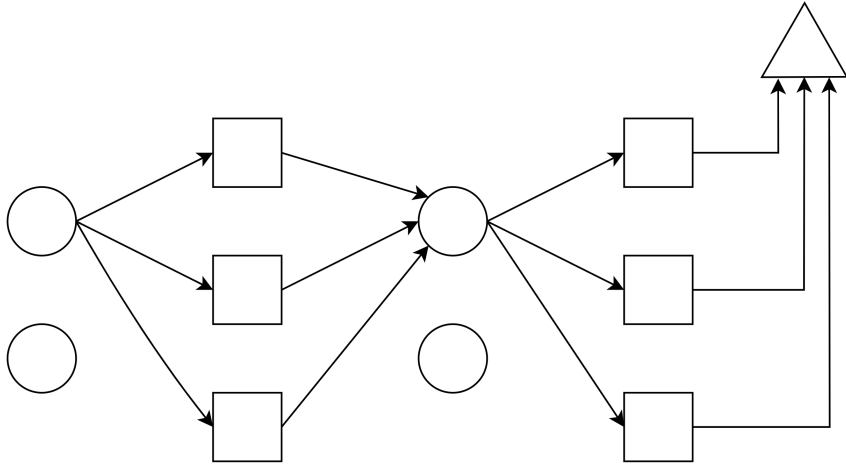
# Simple Promela Interpreter (2)

- Some of the many interesting features of SPIN are:
  - ① Spin targets efficient **software verification**, not hardware.
  - ② Provides direct support for the use of **embedded C code** as part of model specifications.
  - ③ Works **on-the-fly**, which means that it avoids the preemptive construction of a global state graph.
  - ④ Spin supports both rendezvous and buffered FIFO **message passing** between processes.
  - ⑤ To optimize the verification runs, Spin exploits **partial order reduction** techniques, and (optionally) BDD-like storage techniques.

# Process Meta Language

- Very similar to the C programming language
- Can specify **finite**-state systems.
  - Everything **must be bounded**
- A Promela model consists of the declaration of:
  - 1 Types
  - 2 Global variables
  - 3 Channels
  - 4 Processes
  - 5 `init`

# Case Study: Paxos



# Algorithm

---

**Algorithm 1** Paxos — Proposer  $p$ 

---

```
1: Constants:  
2:  $A, n$ , and  $f$ .           { $A$  is the set of acceptors.  $n = |A|$  and  
    $f = \lfloor (n - 1)/2 \rfloor$ .}  
3: Init:  
4:  $crnd \leftarrow -1$            {Current round number}  
5: on  $\langle \text{PROPOSE}, val \rangle$   
6:    $crnd \leftarrow \text{pickNextRound}(crnd)$   
7:    $cval \leftarrow val$   
8:    $P \leftarrow \emptyset$   
9:   send  $\langle \text{PREPARE}, crnd \rangle$  to  $A$   
10: on  $\langle \text{PROMISE}, rnd, vrnd, vval \rangle$  with  $rnd = crnd$  from  
    acceptor  $a$   
11:    $P \leftarrow P \cup (vrnd, vval)$   
12: on event  $|P| \geq n - f$   
13:    $j = \max\{vrnd : (vrnd, vval) \in P\}$   
14:   if  $j \geq 0$  then  
15:      $V = \{vval : (j, vval) \in P\}$   
16:      $cval \leftarrow \text{pick}(V)$  {Pick proposed value  $vval$  with  
        largest  $vrnd$ }  
17:   send  $\langle \text{ACCEPT}, crnd, cval \rangle$  to  $A$ 
```

---

---

**Algorithm 2** Paxos — Acceptor  $a$ 

---

```
1: Constants:  
2:  $L$                                {Set of learners}  
3: Init:  
4:  $rnd \leftarrow -1$   
5:  $vrnd \leftarrow -1$   
6:  $vval \leftarrow -1$   
7: on  $\langle \text{PREPARE}, prnd \rangle$  with  $prnd > rnd$  from proposer  $p$   
8:    $rnd \leftarrow prnd$   
9:   send  $\langle \text{PROMISE}, rnd, vrnd, vval \rangle$  to proposer  $p$   
10: on  $\langle \text{ACCEPT}, i, v \rangle$  with  $i \geq rnd$  from proposer  $p$   
11:    $rnd \leftarrow i$   
12:    $vrnd \leftarrow i$   
13:    $vval \leftarrow v$   
14:   send  $\langle \text{LEARN}, i, v \rangle$  to  $L$ 
```

---

---

**Algorithm 3** Paxos — Learner  $l$ 

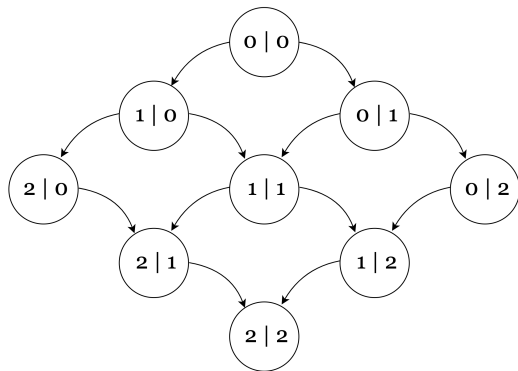
---

```
1: Init:  
2:  $V \leftarrow \emptyset$   
3: on  $\langle \text{LEARN}, (i, v) \rangle$  from acceptor  $a$   
4:    $V \leftarrow V \uplus (i, v)$   
5: on event  $\exists i, v : |\{(i, v) : (i, v) \in V\}| \geq n - f$   
6:    $v$  is chosen
```

---

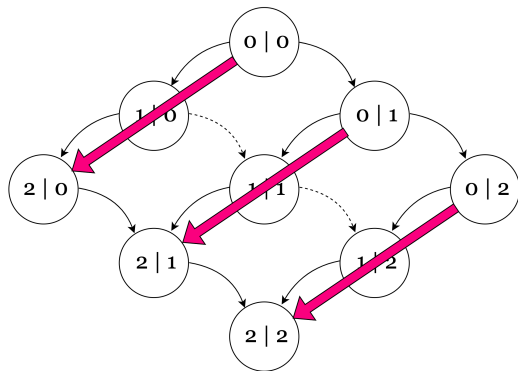
# Atomic

```
active proctype P1() {  
    a; b;  
}  
active proctype P2() {  
    c; d;  
}
```



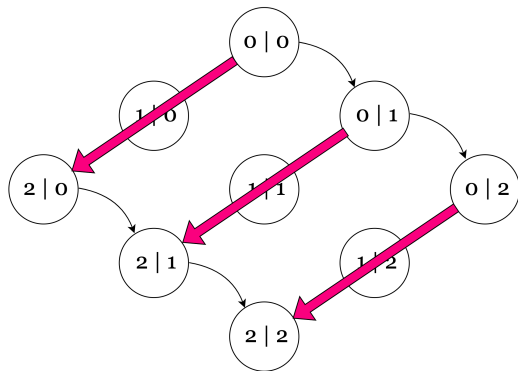
## Atomic (2)

```
active proctype P1() {  
    atomic { a; b; }  
}  
active proctype P2() {  
    c; d;  
}
```



## Atomic (3)

```
active proctype P1() {  
    d_step { a; b; }  
}  
active proctype P2() {  
    c; d;  
}
```



*Fin.*