

Compositional Reasoning

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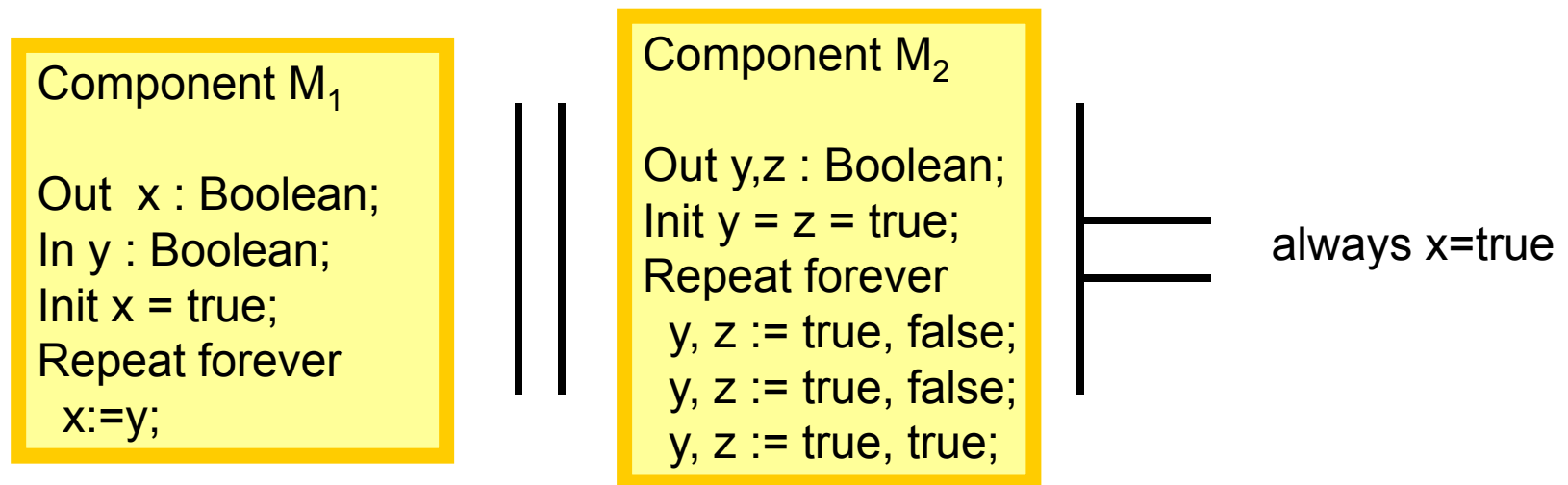
(original created by Yu-Fang Chen)

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Verification of Parallel Compositions

- **Verification Task:** verify if the system composed of components M_1 and M_2 satisfies a property P , i.e., $M_1 \parallel M_2 \models P$.
- M_1 and M_2 may rely on each other to satisfy P .

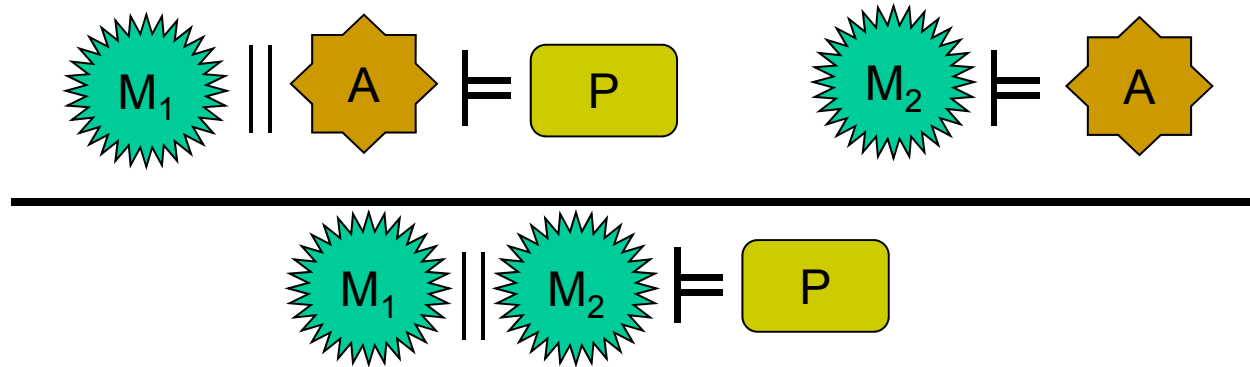


M_1 alone does not guarantee “always $x = \text{true}$ ”!

- Can the construction of $M_1 \parallel M_2$ be avoided?

Compositional Reasoning

- An **Assume-Guarantee (A-G)** rule:



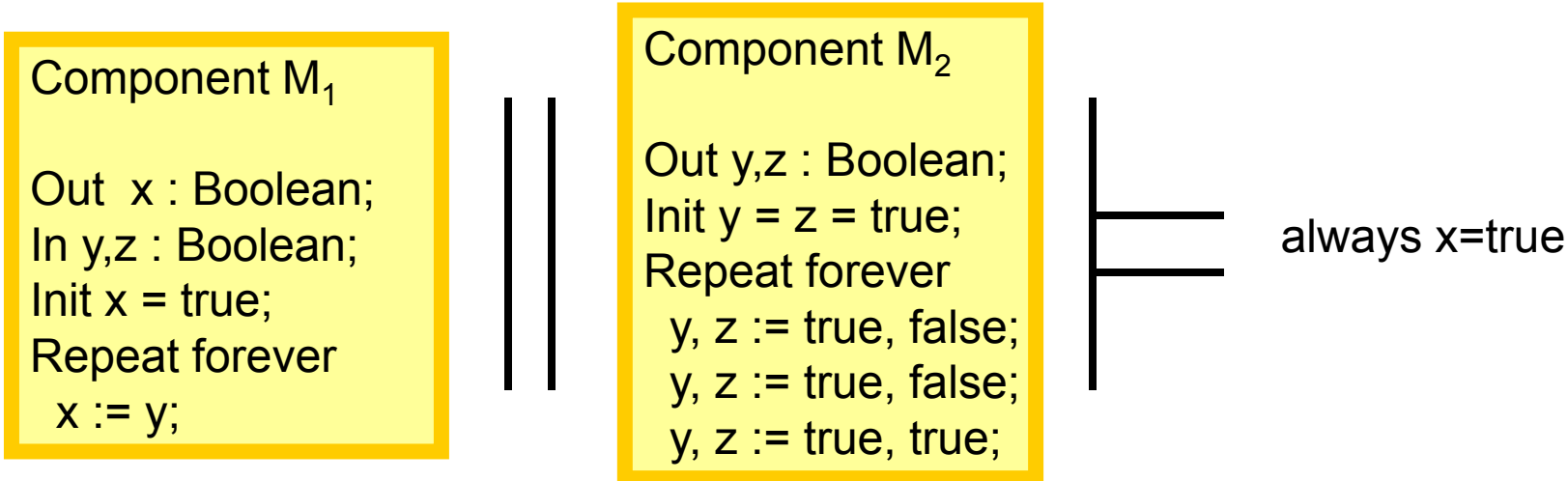
- If a small *contextual assumption* A (an abstraction of M_2) exists, then the overall verification task may become easier.



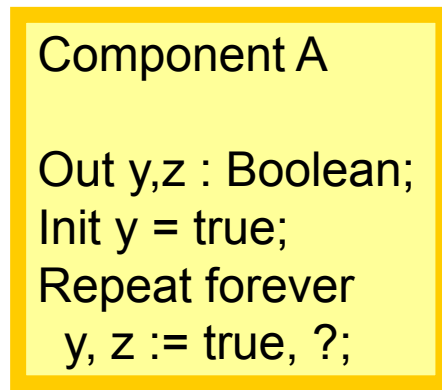
But, how to find A automatically?

- It is possible when M_1 , M_2 , A , and P are finite automata.

Compositional Reasoning (cont.)

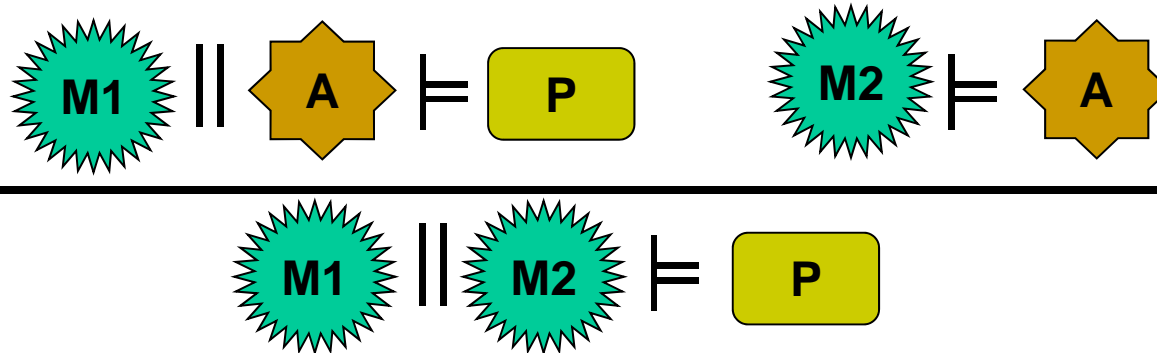


A suitable contextual assumption  :

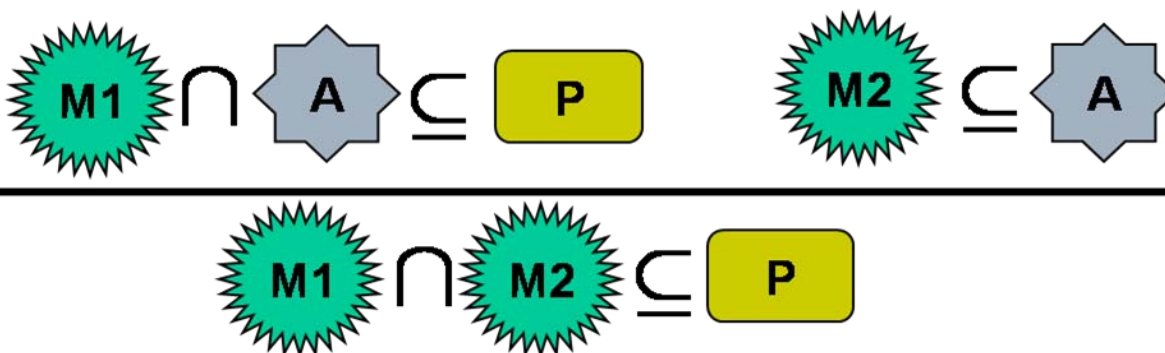


Component A has fewer states (automaton locations) than M_2 .

Setting the Stage



- The **behaviors of components** and **properties** are described as **regular languages**.
- **Parallel composition** is presented by **the intersection of the languages**.
- **A system satisfies a property** if **the language of the system is a subset of the language of the property**.

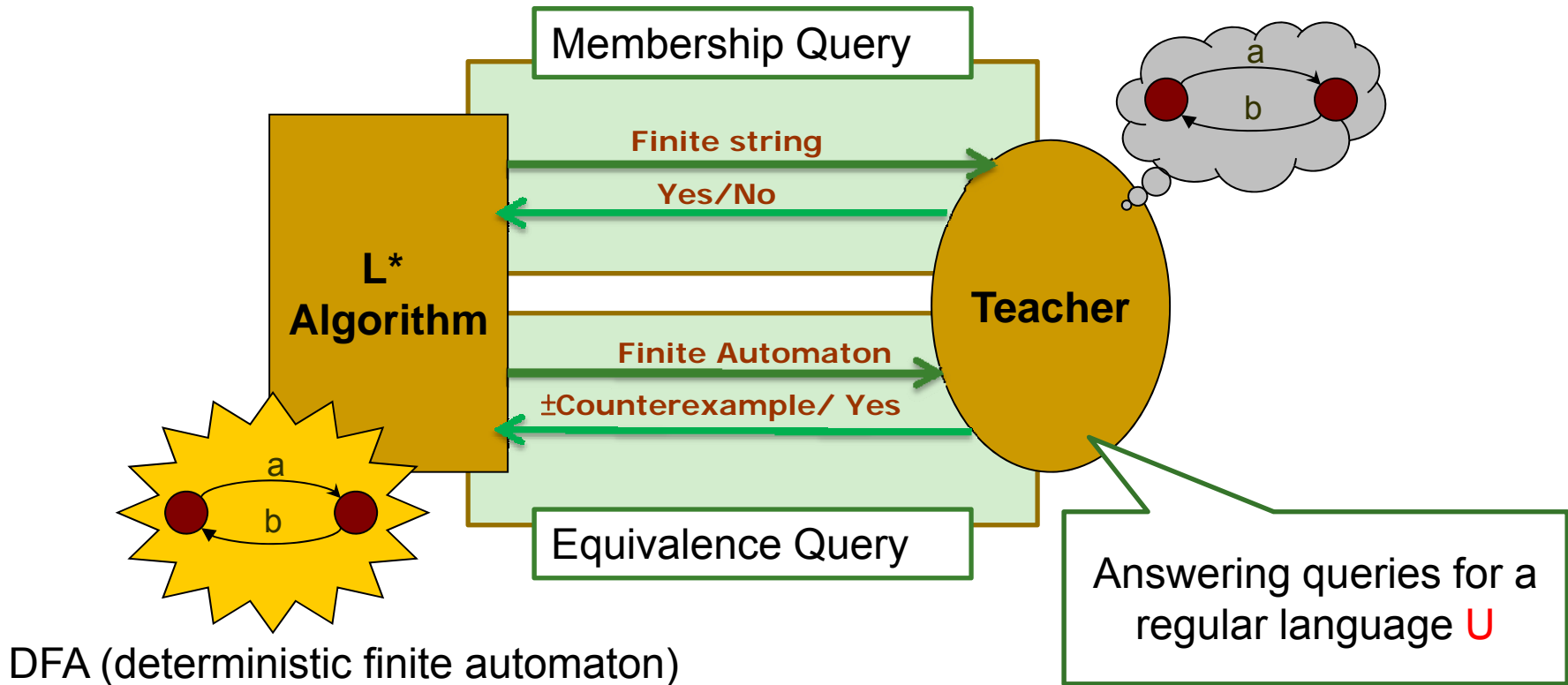


Outline

- Learning-Based Compositional Model Checking:
 - Automation by Learning
 - The L^* Algorithm
 - The Problem of L^* -Based Approaches


- Learning Minimal Separating DFA's:
 - The L^{SEP} Algorithm
 - Comparison with Another Algorithm
 - Adapt L^{SEP} for Compositional Model Checking

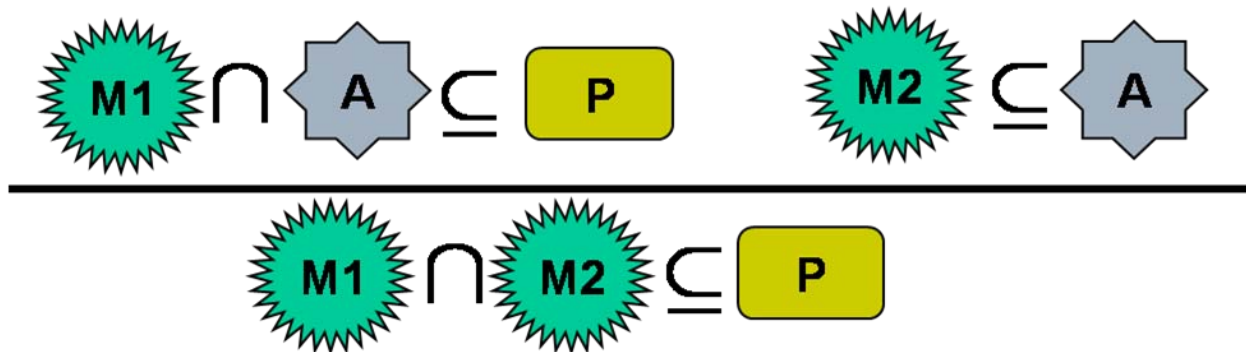
Overview of the L* Algorithm



If such a teacher is provided, L* guarantees to produce a DFA that recognizes U using a polynomial number of queries.

Automation by Learning

- **First developed** by Cobleigh, Giannakopoulou, and Păsăreanu [TACAS 2003]
- Apply the L^* learning algorithm for regular languages to find an  for the A-G rule:



Basic Understanding

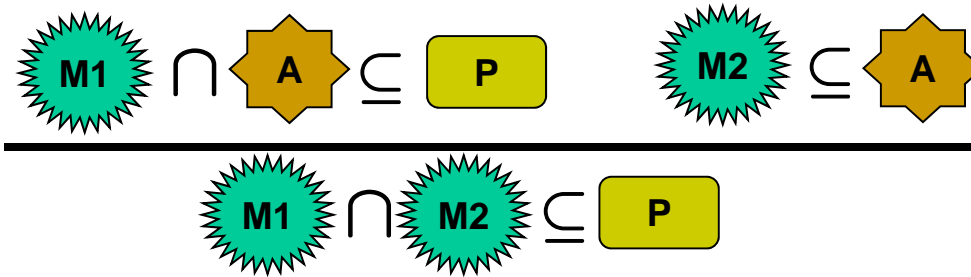
- A closer look at the A-G rule:

$$M1 \cap A \subseteq P \Leftrightarrow$$

$$M1 \cap A \cap \bar{P} = \emptyset \Leftrightarrow$$

$$A \cap \overline{(P \cup M1)} = \emptyset \Leftrightarrow$$

$$A \subseteq \overline{P \cup M1}$$



$$\text{When } A = P \cup \bar{M1},$$

$$M2 \subseteq A \Leftrightarrow$$

$$M2 \subseteq \overline{P \cup M1} \Leftrightarrow$$

$$M2 \cap \overline{(P \cup M1)} = \emptyset \Leftrightarrow$$

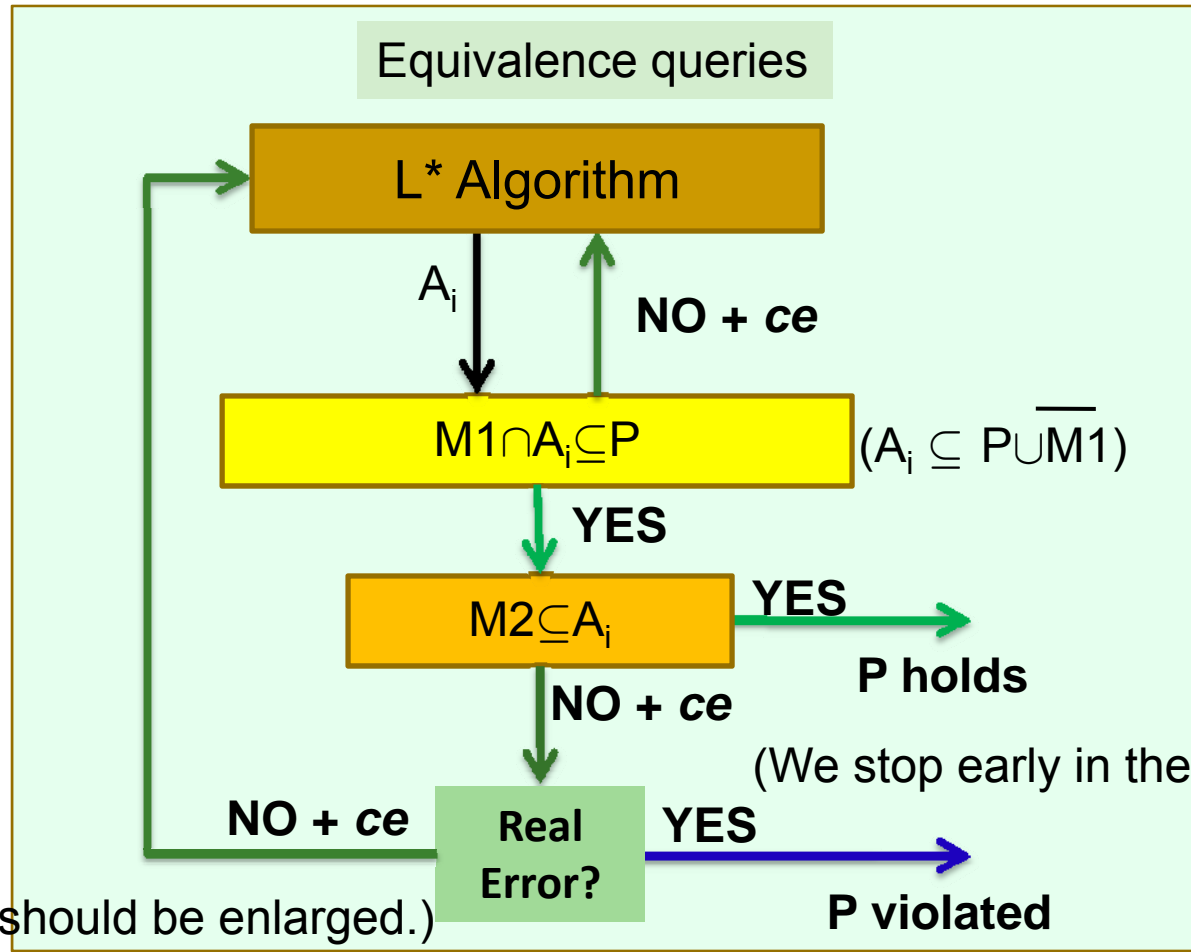
$$M2 \cap \bar{P} \cap M1 = \emptyset \Leftrightarrow$$

$$M1 \cap M2 \subseteq P$$

- Conceptually, the target language is $\overline{P \cup M1}$, the *weakest assumption* for the premise $M1 \cap A \subseteq P$.
- Actually reaching the target would be even worse than checking $M1 \cap M2 \subseteq P$ directly.
- It really pays off when we can stop earlier ...

The Algorithm of Cobleigh *et al.*

Target: $\overline{P \cup M1}$



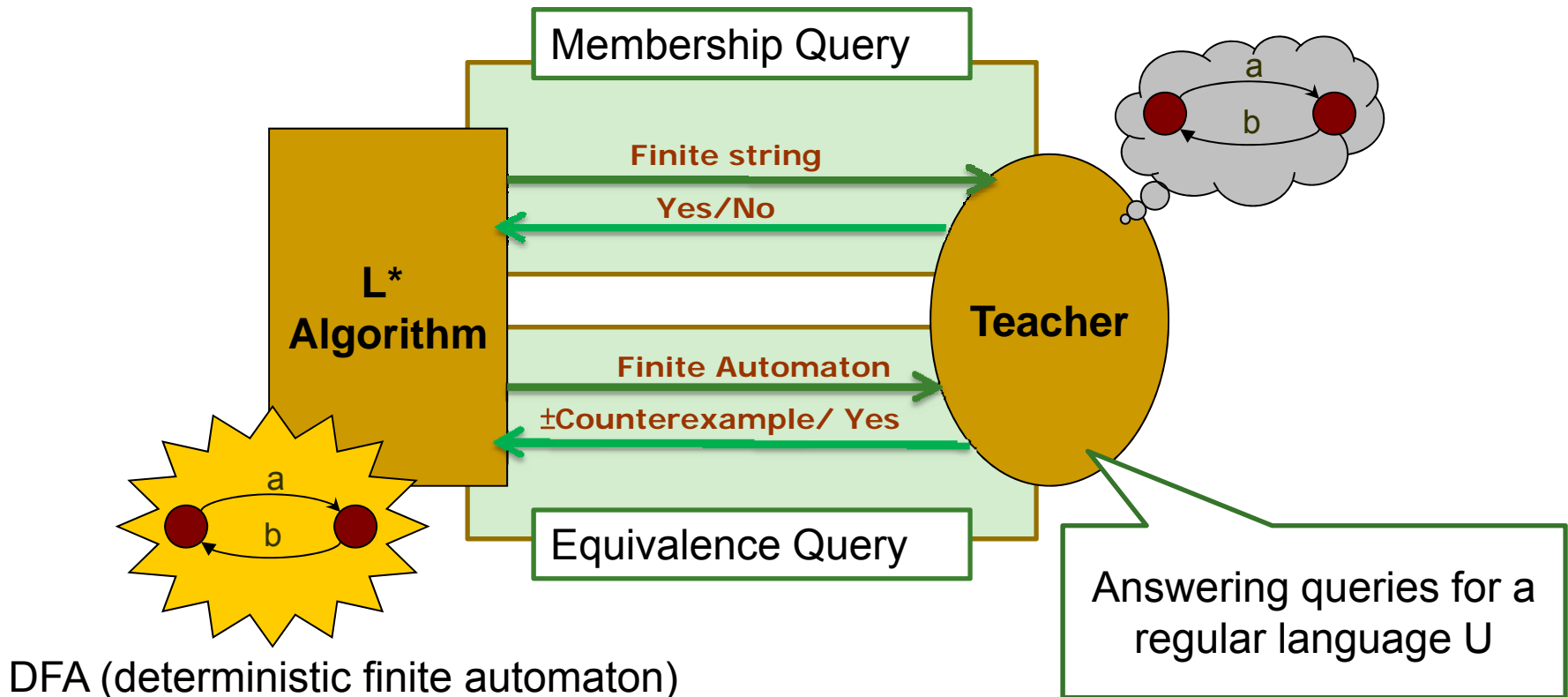
(ce tells how A_i should be enlarged.)

(We stop early in these two cases.)

(ce is a real error if ce is in $M2$, but not in $\overline{P \cup M1}$, implying $M2 \not\subseteq \overline{P \cup M1}$, i.e., $M1 \cap M2 \not\subseteq P$.)

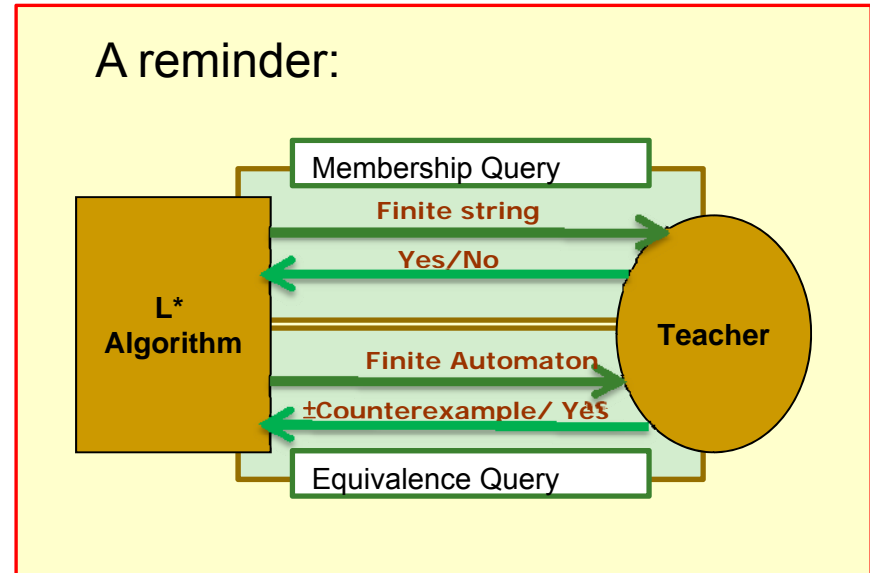
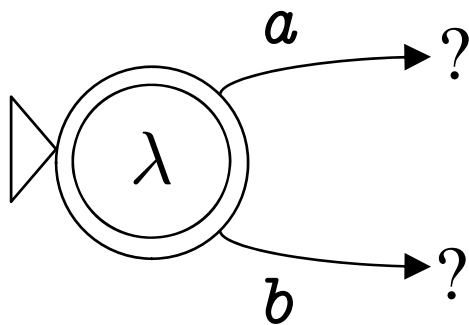
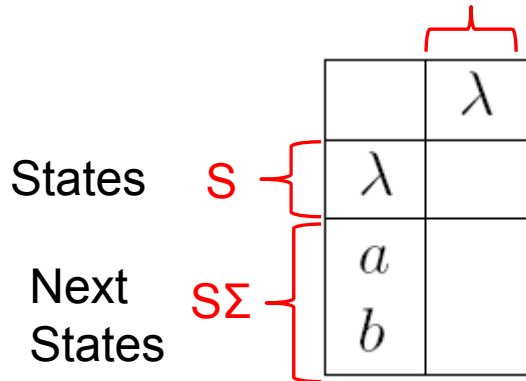
The L* Learning Algorithm

- Proposed by D. Angluin [Info.&Comp. 1987] and improved by Rivest and Schapire [Info.&Comp. 1993]



L*: Initial Setting

E: Distinguishing Experiments



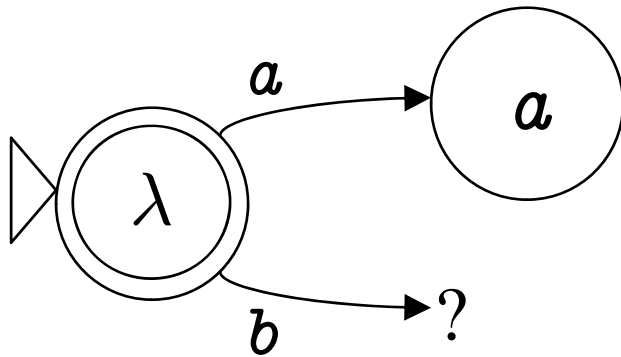
Target: $(ab+aab)^*$

L*: Fill Up the Table by Membership Queries

	λ
λ	T
a	F
b	F

Fill up the table using **membership queries**.

a represents a new equivalence class, because its **row** is different from all of those in the current S set.



Target: $(ab+aab)^*$

L*: Table Expansion

Move a to the S set and expand the table with elements aa and ab .

	λ
λ	T
a	F
b	F
aa	
ab	

Target: $(ab+aab)^*$

L^* : A Closed Table

	λ
λ	T
a	F
b	F
aa	F
ab	T

Again, fill up the table using membership queries.

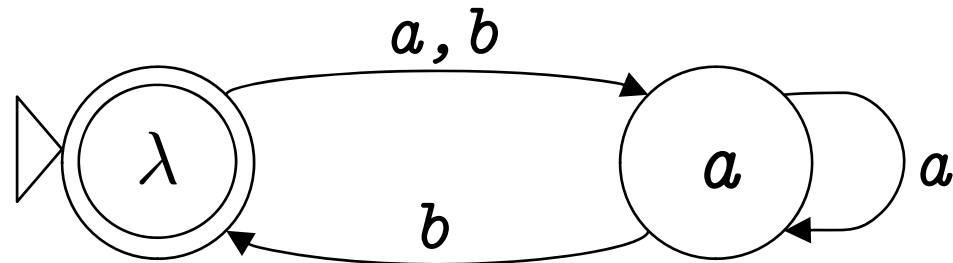
We say that the table is **closed** because every row in the $S\Sigma$ set appears somewhere in the S set.

Target: $(ab+aab)^*$

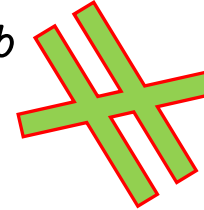
L*: Making a Conjecture

	λ
λ	T
a	F
b	F
aa	F
ab	T

Construct a DFA from the learned equivalence classes.



Counterexample: bb



$\delta(s, a) = s'$ iff sa and s' have the same row.

A suffix b is extracted from bb as a valid distinguishing experiment

Target: $(ab+aab)^*$

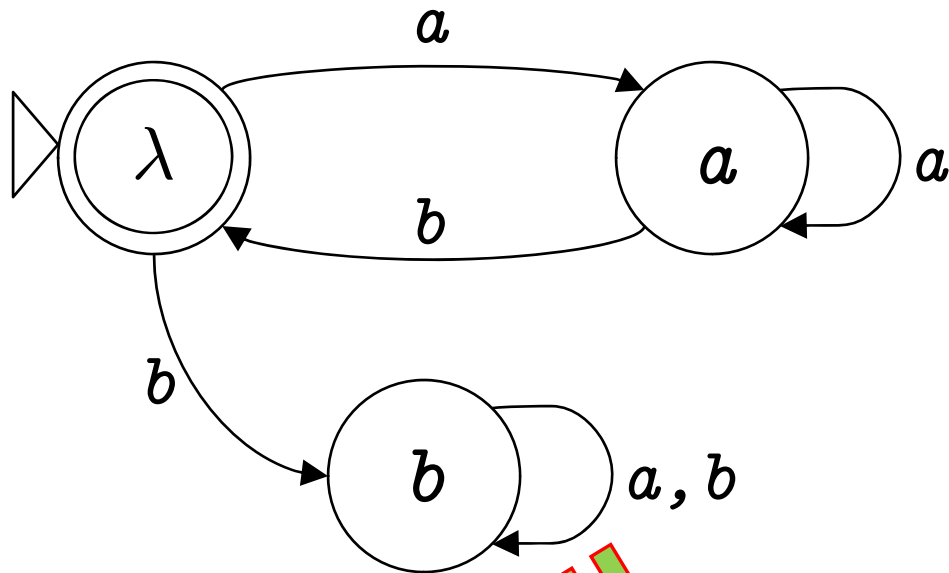
Theorem:

At least one suffix of the counterexample is a valid distinguishing experiment.

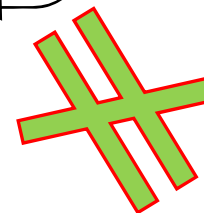
L*: 2nd Iteration

Add b to the E set, fill up and expand the table following the same procedure.

	λ	b
λ	T	F
a	F	T
b	F	F
aa	F	T
ab	T	F
ba	F	F
bb	F	F



Counterexample: $aaab$



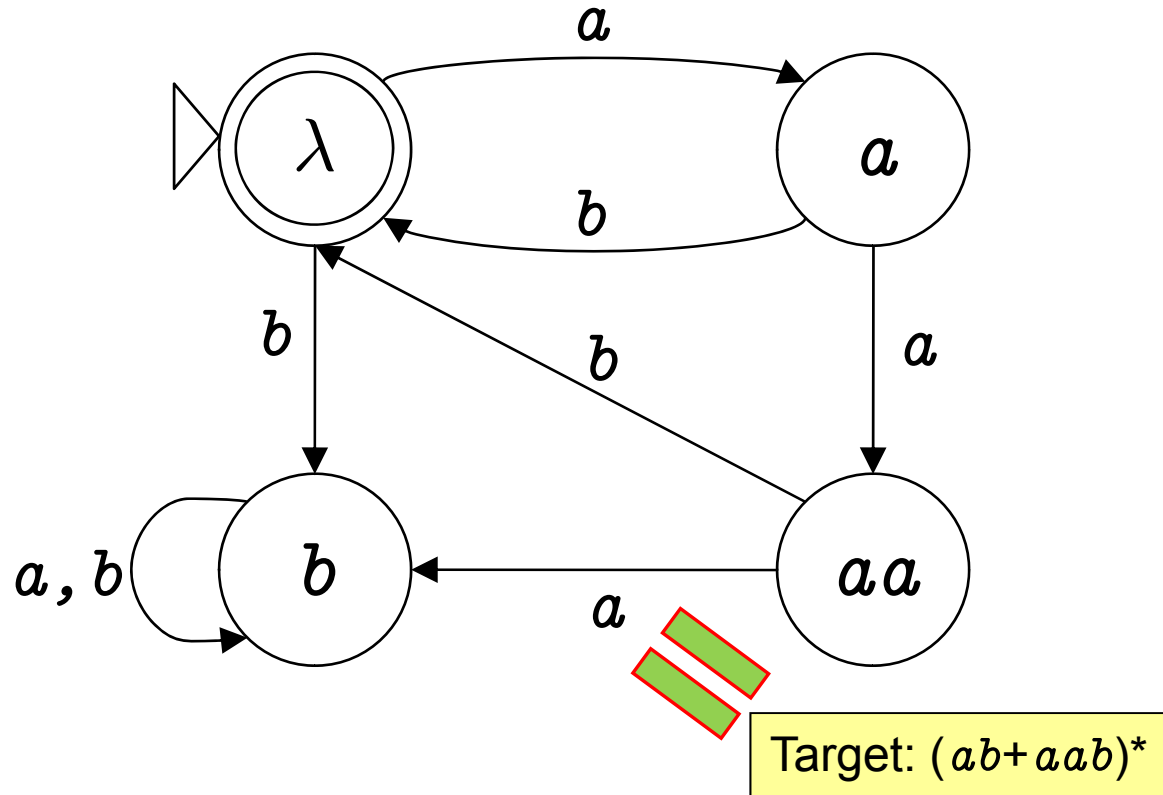
A suffix ab is extracted from $aaab$ as a valid distinguishing experiment.

Target: $(ab+aab)^*$

L*: 3rd Iteration (Completed)

Add ab to the E set, fill up and expand the table following the same procedure.

	λ	b	ab
λ	T	F	T
a	F	T	T
b	F	F	F
aa	F	T	F
ab	T	F	T
ba	F	F	F
bb	F	F	F
aaa	F	F	F
aab	T	F	T



Theorem:

The DFA produced by L* is the minimal DFA that recognizes that target language.

L*: Complexity

■ Complexity:

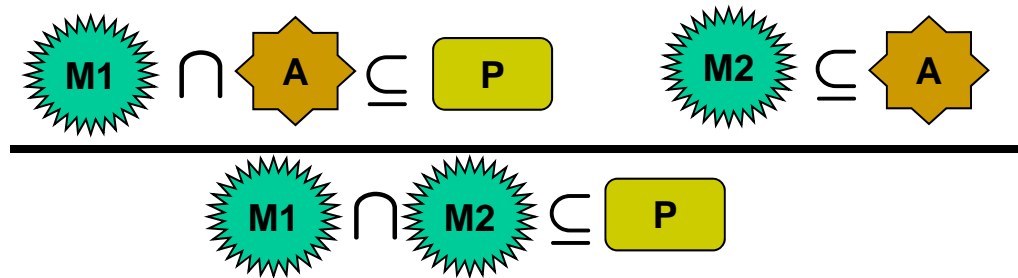
- Equivalence query: at most n
- Membership query: $O(|\Sigma|n^2 + n \log m)$

	λ	b	ab
λ	T	F	T
a	F	T	T
b	F	F	F
aa	F	T	F
ab	T	F	T
ba	F	F	F
bb	F	F	F
aaa	F	F	F
aab	T	F	T

Note: n is the size of the minimal DFA that recognizes U , m is the length of the longest counterexample returned from the teacher.

The Problem

- The L^* -based approaches cannot guarantee finding the **minimal assumption** (in size), even if there exists one.



- The smaller the size of A is, the easier it is to check the correctness of the two premises.
- L^* targets a single language, however, there exists a range of languages that satisfy the premises of an A-G rule.

Finding a Minimal Assumption

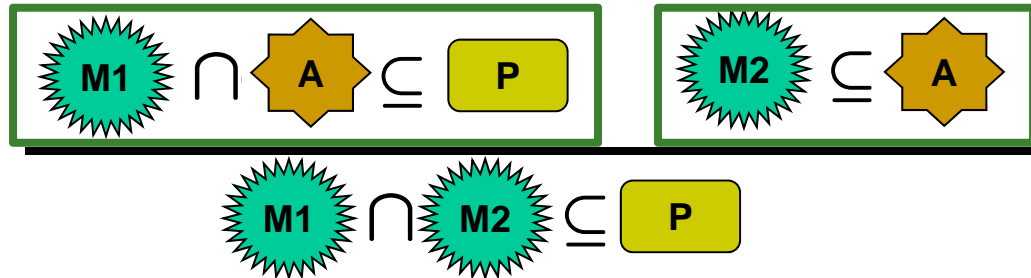
- **A reminder:** we use the following Assume-Guarantee rule for decomposition.

$$M1 \cap A \subseteq P \Leftrightarrow$$

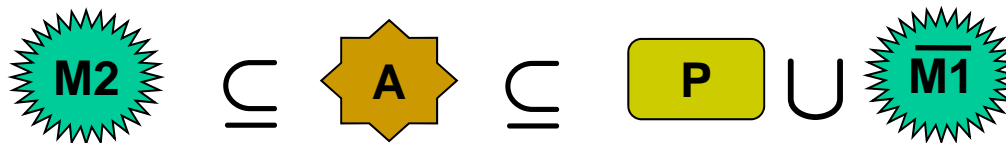
$$M1 \cap A \cap \bar{P} = \emptyset \Leftrightarrow$$

$$A \cap \overline{(P \cup \bar{M1})} = \emptyset \Leftrightarrow$$


$$A \subseteq P \cup \bar{M1}$$



- The two **premises** can be rewritten as follows:



Finding a Minimal Assumption (cont.)

- To apply the A-G rule is to find an  satisfying the following constraint:

$$\text{M2} \subseteq \text{A} \subseteq \text{P} \cup \overline{\text{M1}}$$

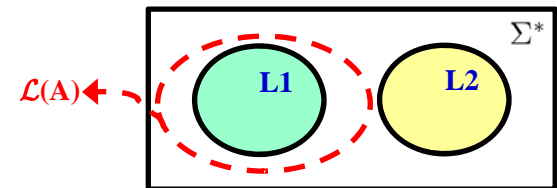
- So, the problem of finding a minimal assumption for the A-G rule reduces to **finding a minimal separating DFA** that
 - **accepts** every string in **M2** and
 - **rejects** every string not in **P** \cup **$\overline{\text{M1}}$** .

First observed by Gupta, McMillan, and Fu

Learning a Minimal Separating DFA

- **Contribution of [Chen et al. TACAS 2009]:** a **polynomial-query** learning algorithm, L^{Sep} , for minimal separating DFA's.
- **Problem:** given two disjoint regular languages **L1** and **L2**, we want to find a minimal DFA **A** that satisfies

$$L1 \subseteq \mathcal{L}(A) \subseteq \overline{L2}$$

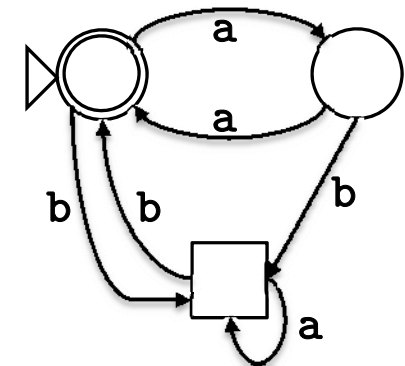


We say that **A** is a **separating DFA** for L1 and L2

- **Assumption:** a teacher for L1 and L2:
 - Membership query: if a string **s** is in L1 (resp. L2)
 - Containment query: $? \subseteq L1$, $? \supseteq L1$, $? \subseteq L2$, and $? \supseteq L2$

3-Value DFA (3DFA)

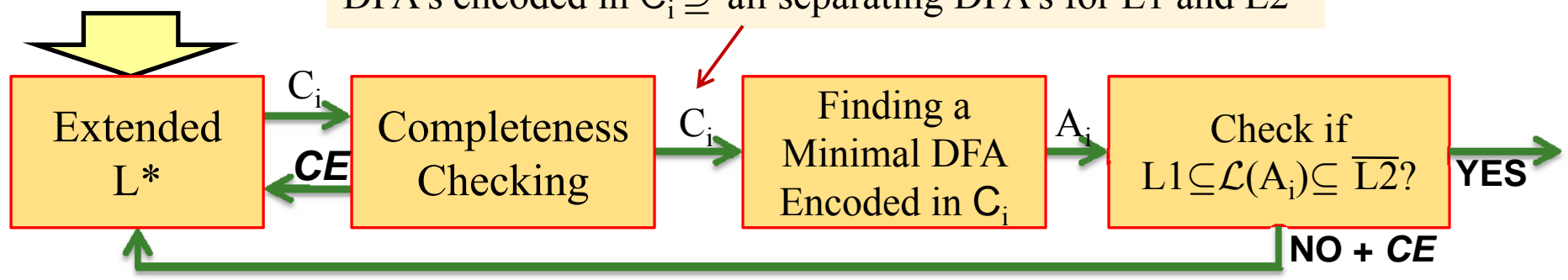
- A 3DFA is a tuple $C = (\Sigma, S, s_0, \delta, Acc, Rej, Dont)$.
- A **DFA** A is **encoded in** a **3DFA** C iff A
 - accepts all strings that C accepts and
 - rejects all strings that C rejects.
 - A don't care string in C can be either accepted or rejected by A .



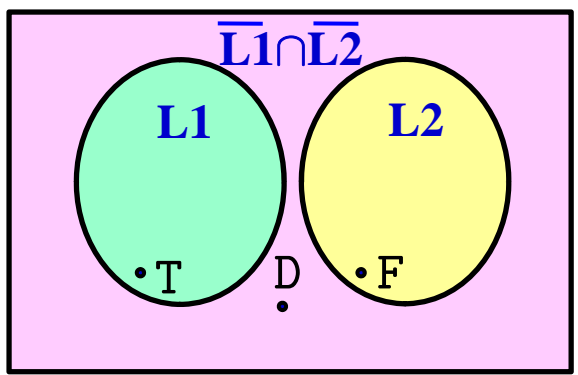
An example of a 3DFA

The L^{Sep} Algorithm: Overview

DFA's encoded in $C_i \supseteq$ all separating DFA's for L1 and L2



Target:



	λ	a
λ	T	F
a	F	T
ab	D	D
b	D	D
aa	T	F
aba	D	D
abb	T	F

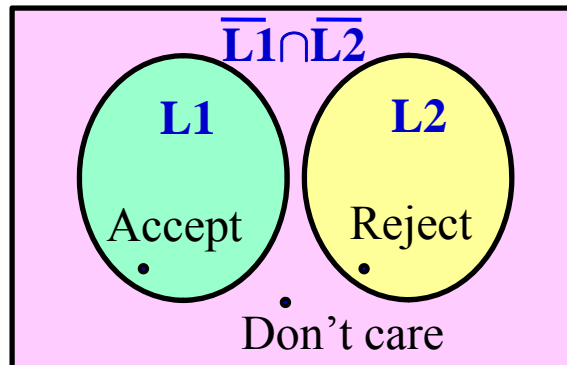
Extend the **L*** algorithm to allow **don't care** values.

The Target 3DFA

■ The target 3DFA C

- **accepts** every string in $L1$, and
- **rejects** every string in $L2$.
- Strings in $\overline{L1} \cap \overline{L2}$ are **don't care** strings.

DFA's encoded in C =
all separating DFA's for $L1$ and $L2$

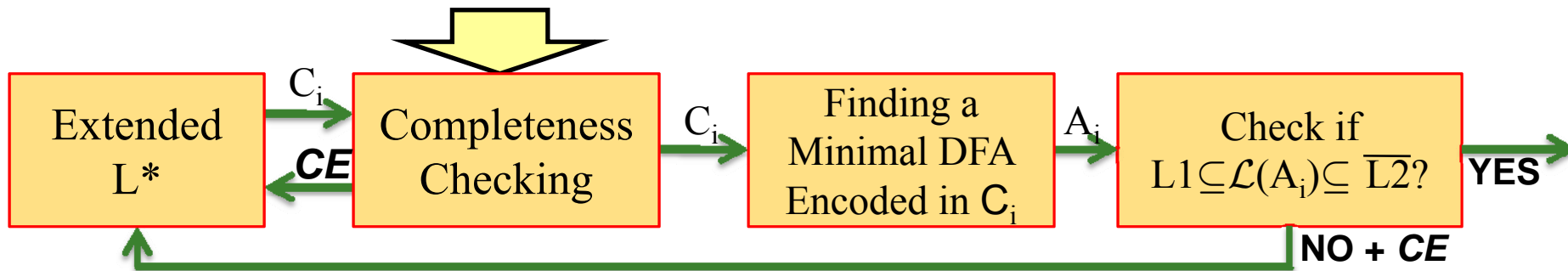


Definition:

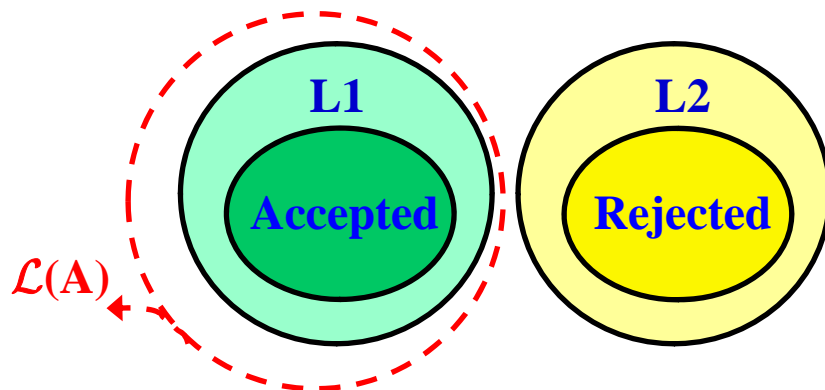
- A DFA A is **encoded in** a 3DFA C iff A
 - accepts all strings that C accepts and
 - rejects all strings that C rejects.
- A DFA A **separates** $L1$ and $L2$ iff A
 - accepts all strings in $L1$ and
 - rejects all strings in $L2$.

- A minimal DFA encoded in C is a minimal separating DFA of $L1$ and $L2$.

The L^{Sep} Algorithm



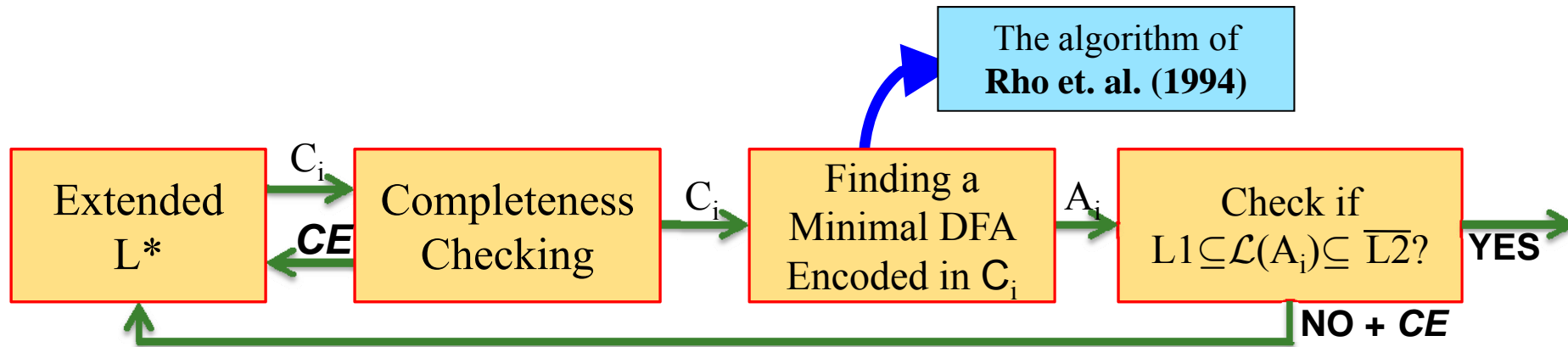
Check if all of the **separating** DFA's of $L1$ and $L2$ are **encoded** in C_i , which can be done by checking the following conditions:



Definition:

- A **DFA A** is **encoded in** a **3DFA C** iff **A**
 - accepts all strings that **C** accepts and
 - rejects all strings that **C** rejects.
- A **DFA A** **separates** **$L1$** and **$L2$** iff **A**
 - accepts all strings in **$L1$** and
 - rejects all strings in **$L2$** .

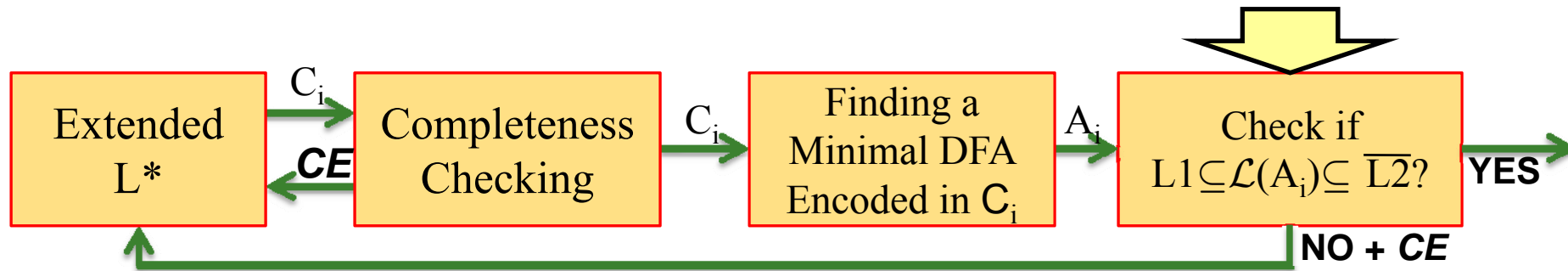
The L^{Sep} Algorithm



LEMMA:

The size of **minimal separating DFA** of L_1 and $L_2 \geq |A_i|$, the size of the **minimal DFA encoded in C_i** .

The L^{Sep} Algorithm



If $L1 \subseteq \mathcal{L}(A_i) \subseteq \overline{L2}$:

A_i is a minimal separating DFA.

If $L1 \not\subseteq \mathcal{L}(A_i)$ or $\mathcal{L}(A_i) \not\subseteq \overline{L2}$:

Counterexample CE is a witness for C_i not being the target 3DFA.

LEMMA:

The size of **minimal separating DFA** of $L1$ and $L2 \geq |A_i|$, the size of the **minimal DFA encoded in C_i** .

The Algorithm of Gupta *et al.*

Begin with an empty **sample set**

Requires an **exponential** number of iterations in the worst case

Make a **minimal DFA** that consistent with the current **sample set**

Check if $L1 \subseteq \mathcal{L}(A_i) \subseteq \overline{L2}$?

YES

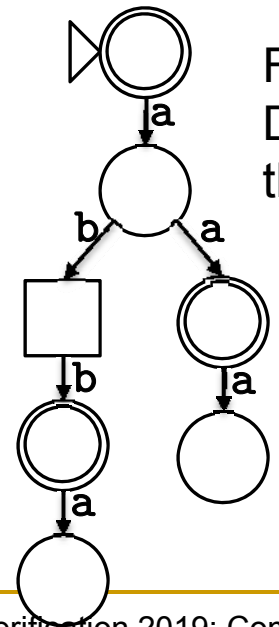
NO + CE
Add **CE** to the sample set

An Example

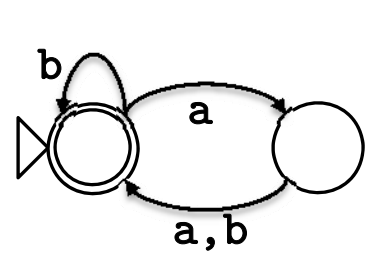
+ SAMPLES :
 λ, aa, abb

- SAMPLES :
 $a, aaa, abba$

Make a 3DFA



Find a minimal DFA encoded in the 3DFA (NP-hard)



The L^{Sep} Algorithm

Extend the L^* algorithm to manage the collected samples.

A_i

Check if $L1 \subseteq \mathcal{L}(A_i) \subseteq \overline{L2}$?

YES

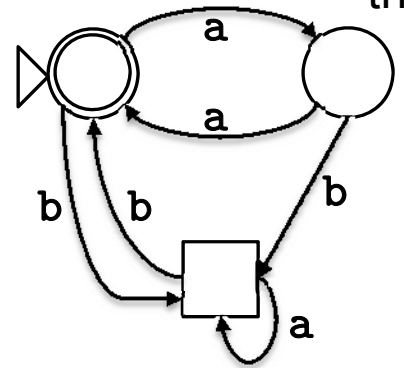
NO + CE

Requires a **polynomial** number of iterations in the worst case

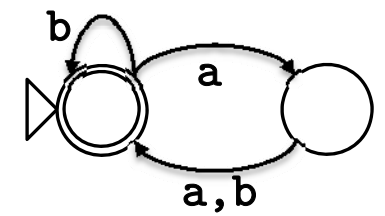
An Example

	λ	a
λ	T	F
a	F	T
ab	D	D
b	D	D
aa	T	F
aba	D	D
abb	T	F

Make a 3DFA



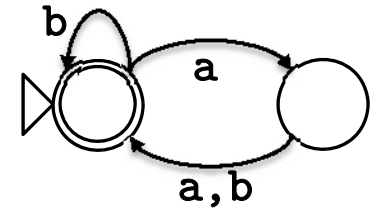
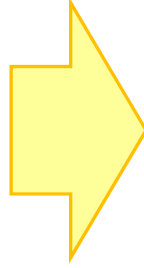
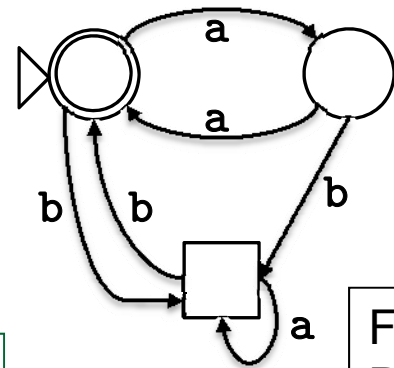
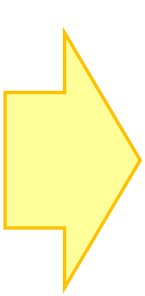
Find a minimal DFA encoded in the 3DFA (NP-hard)



Comparing the Two Algorithms

L_{Sep}:

	λ	a
λ	T	F
a	F	T
ab	D	D
b	D	D
aa	T	F
aba	D	D
abb	T	F



Make a 3DFA

Find a minimal DFA encoded in the 3DFA (NP-hard)

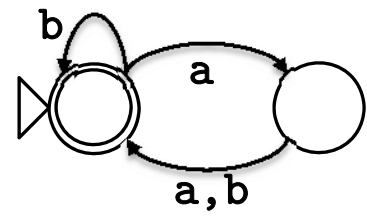
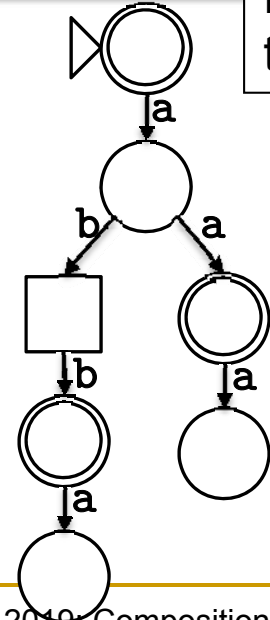
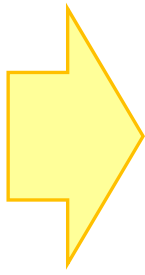
Same sample set!



+ SAMPLES:
 λ, aa, abb

- SAMPLES:
 $a, aaa, abba$

Gupta et al. :

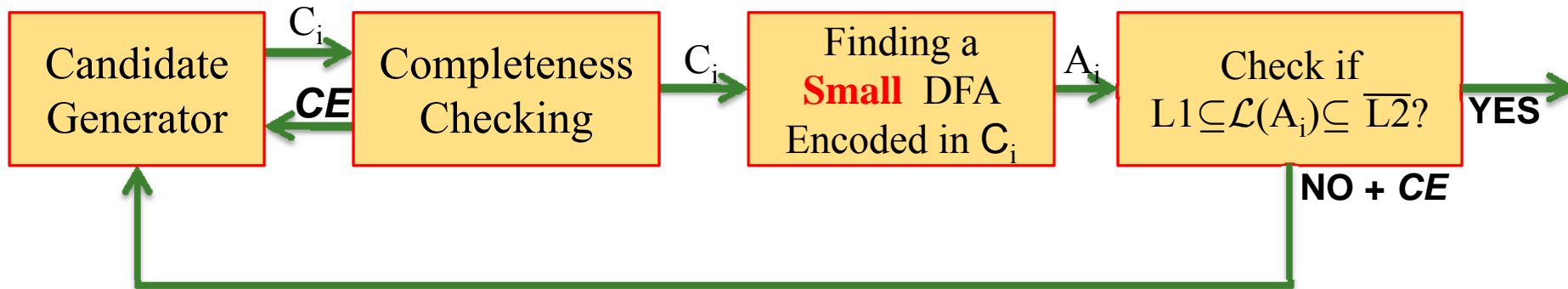


Adapt L^{Sep} for Compositional Verification

- Let $L1 = M2$ and $\bar{L}2 = P \cup \bar{M}1$, use L^{Sep} to find a separating DFA for $L1$ and $L2$.
- When $M2 \not\subseteq P \cup \bar{M}1$ (i.e., $M1 \cap M2 \not\subseteq P$), L^{Sep} can be modified to guarantee finding a string in $M2$, but not in $P \cup \bar{M}1$ (i.e., $M1 \cap M2 \setminus P$).

Adapt L^{Sep} for Compositional Verification

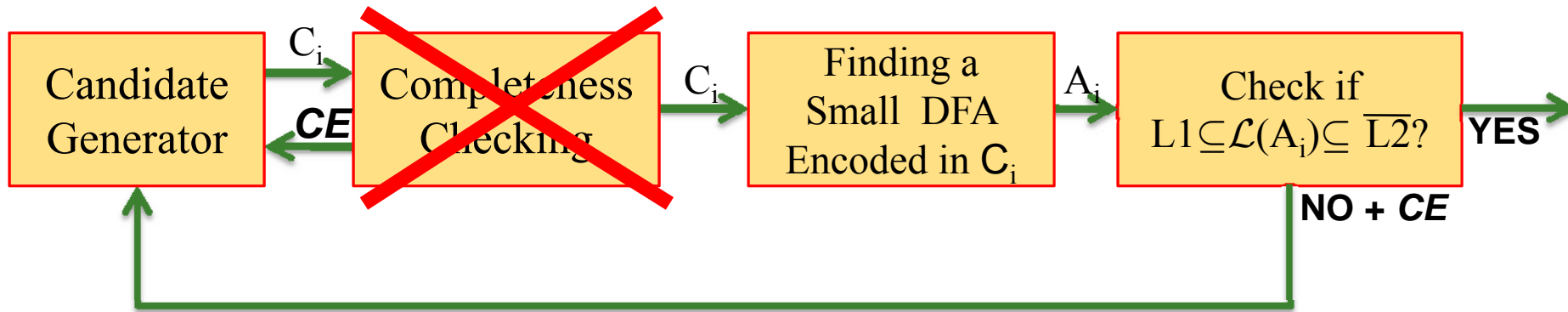
- Use heuristics to find a small consistent DFA:



Minimality is no longer guaranteed!

Adapt L^{Sep} for Compositional Verification

- Skip completeness checking:



Minimality is no longer guaranteed!