

Compositional Reasoning

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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., $M_1 || M_2 \models P$.
- M_1 and M_2 may rely on each other to satisfy P.

```
Component M<sub>1</sub>

Out x: Boolean;
In y: Boolean;
Init x = true;
Repeat forever
x:=y;

Component M<sub>2</sub>

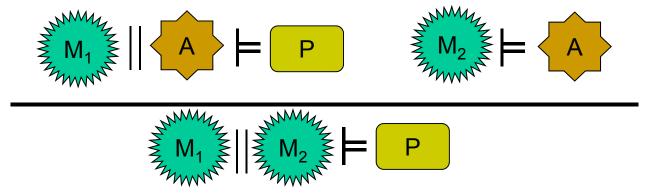
Out y,z: Boolean;
Init y = z = true;
Repeat forever
y, z := true, false;
y, z := true, false;
y, z := true, true;
```

 M_1 alone does not guarantee "always x = true"!

• Can the construction of $M_1 || M_2$ be avoided?

Compositional Reasoning

■ An **Assume-Guarantee** (A-G) rule:



If a small contextual assumption A (an abstraction of M₂) exists, then the overall verification task may become easier.

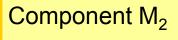


• It is possible when M_1 , M_2 , A, and P are finite automata.

Compositional Reasoning (cont.)

Component M₁

Out x : Boolean; In y,z : Boolean; Init x = true; Repeat forever x := y;



Out y,z : Boolean; Init y = z = true; Repeat forever y, z := true, false; y, z := true, false; y, z := true, true;

___ always x=true

A suitable contextual assumption

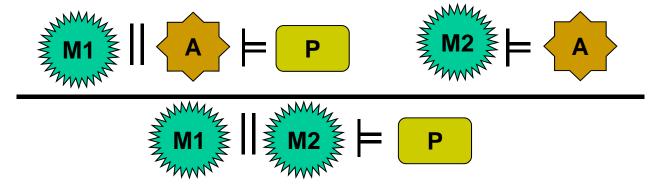


Component A

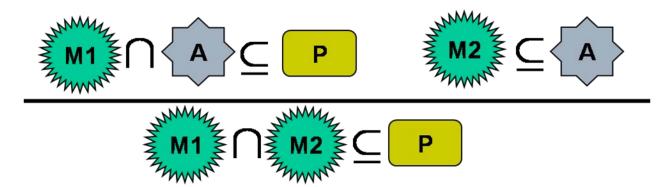
Out y,z : Boolean; Init y = true; Repeat forever y, z := true, ?;

Component A has fewer states (automaton locations) than M₂.

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.

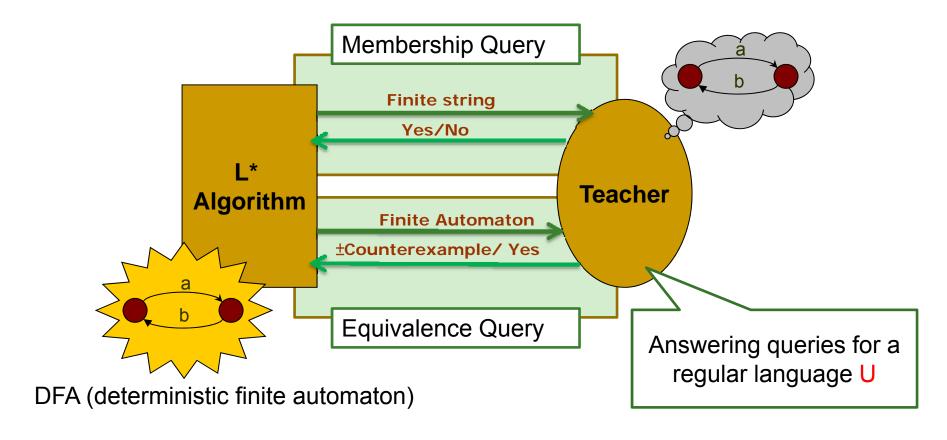


Outline

- 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

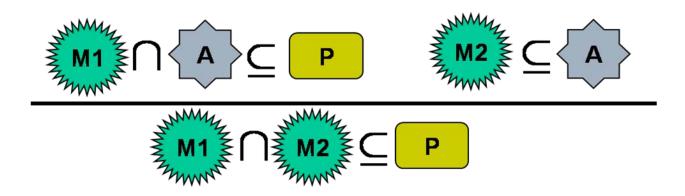


If such a teacher is provided, L* guarantees to produce a DFA that recognizes U using a polynomial number of queries.

Automation by Learning

 First developed by Cobleigh, Giannakopoulou, and Păsăreanu [TACAS 2003]

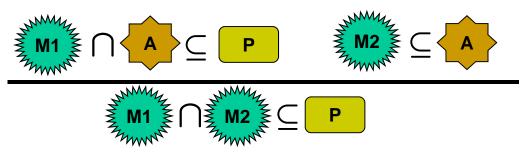
Apply the L* learning algorithm for regular languages to find an for the A-G rule:



Basic Understanding

☐ A closer look at the A-G rule:

```
M1∩ A⊆P \Leftrightarrow
M1∩ A∩\overline{P}=\emptyset \Leftrightarrow
A∩(\overline{P}\cup\overline{M1})=\emptyset \Leftrightarrow
A⊆ \overline{P}\cup\overline{M1}
```

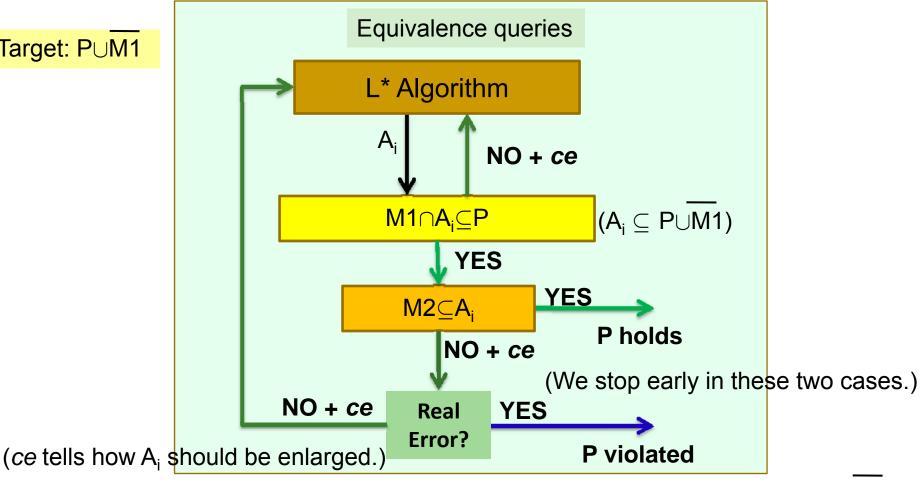


When $A=P\cup\overline{M1}$, $M2\subseteq A\Leftrightarrow$ $M2\subseteq P\cup\overline{M1}\Leftrightarrow$ $M2\cap (P\cup\overline{M1})=\emptyset\Leftrightarrow$ $M2\cap\overline{P}\cap M1=\emptyset\Leftrightarrow$ $M1\cap M2\subseteq P$

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

The Algorithm of Cobleigh et al.

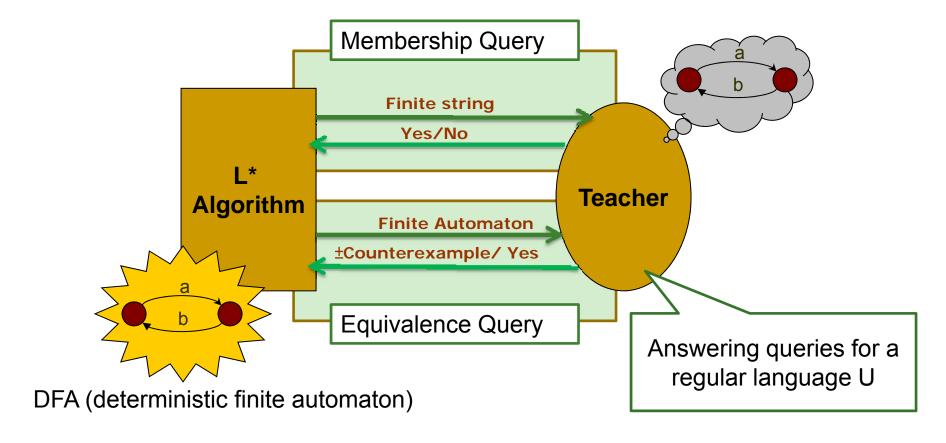
Target: P∪M1



(*ce* is a real error<u>if</u>*ce* is in M2, but not in P∪M1, implying M2 \nsubseteq P \cup M1, i.e., M1 \cap M2 \nsubseteq P.)

The L* Learning Algorithm

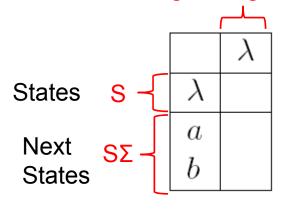
 Proposed by D. Angluin [Info.&Comp. 1987] and improved by Rivest and Schapire [Info.&Comp. 1993]

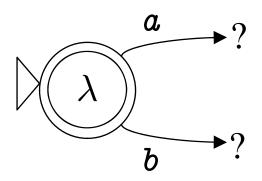


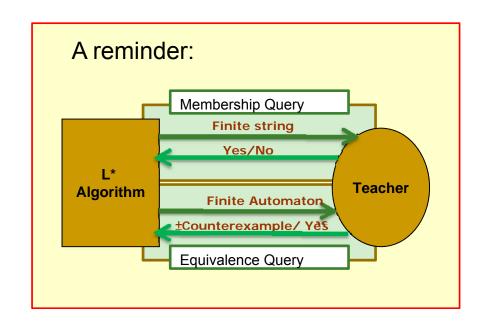


L*: Initial Setting

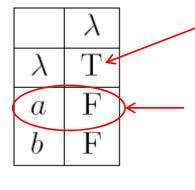
E: Distinguishing Experiments





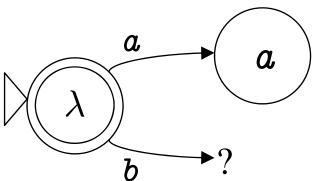


L*: Fill Up the Table by Membership Queries



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.

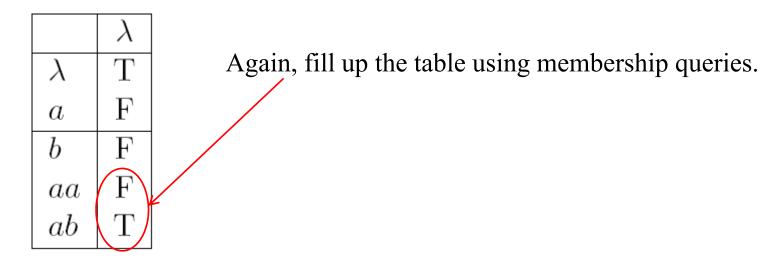


L*: Table Expansion

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

	λ
λ	Τ
a	F
b	F
aa	
ab	

L*: A Closed Table

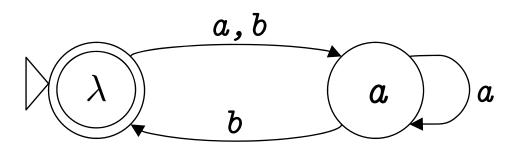


We say that the table is **closed** because every row in the $S\Sigma$ set appears somewhere in the S set.

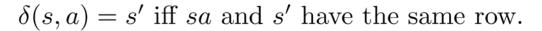
L*: Making a Conjecture

	λ
λ	Τ
a	F
b	F
aa	\mathbf{F}
ab	Τ

Construct a DFA from the learned equivalence classes.



Counterexample: bb



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



Target: $(ab+aab)^*$

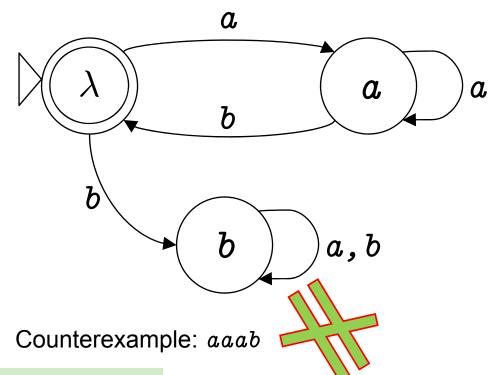
Theorem:

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	Τ	F
a	\mathbf{F}	T
b	F	F
aa	F	Τ
ab	T	F
ba	\mathbf{F}	F
bb	F	F

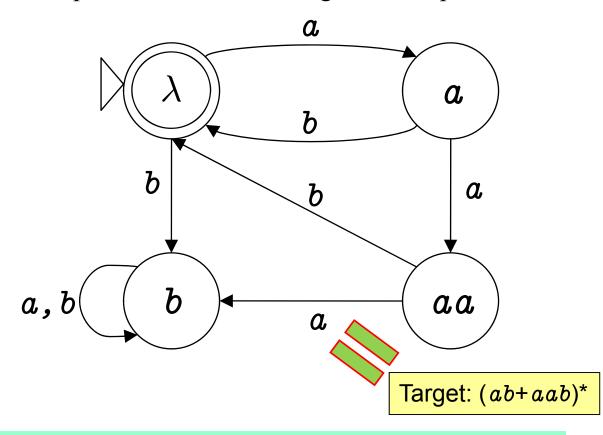


A suffix *ab* is extracted from *aaab* as a valid distinguishing experiment.

L*: 3rd Iteration (Completed)

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

	λ	b	ab
λ	Τ	F	Τ
a	\mathbf{F}	T	Τ
b	\mathbf{F}	\mathbf{F}	F
aa	\mathbf{F}	Τ	F
ab	Τ	F	Τ
ba	F	\mathbf{F}	F
bb	\mathbf{F}	\mathbf{F}	F
aaa	\mathbf{F}	\mathbf{F}	F
aab	Τ	F	T



Theorem:

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

L*: Complexity

Complexity:

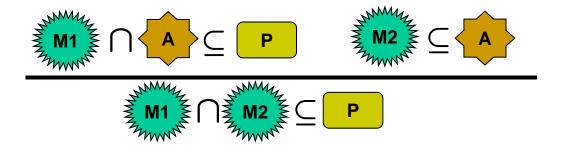
- Equivalence query: at most n
- □ Membership query: $O(|\Sigma|n^2 + n \log m)$

	λ	b	ab
λ	Τ	F	Τ
a	\mathbf{F}	T	Τ
b	\mathbf{F}	F	F
aa	F	Τ	F
ab	Τ	F	Τ
ba	F	\mathbf{F}	F
bb	\mathbf{F}	\mathbf{F}	F
aaa	\mathbf{F}	F	F
aab	Τ	F	Τ

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

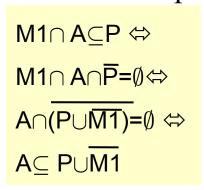
□ The L*-based approaches cannot guarantee finding the minimal assumption (in size), even if there exists one.

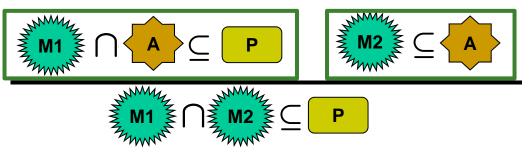


- The smaller the size of is, the easier it is to check the correctness of the two premises.
- □ L* targets a single language, however, there exists a range of languages that satisfy the premises of an A-G rule.

Finding a Minimal Assumption

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



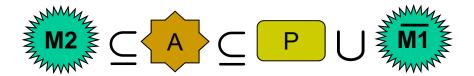


■ The two premises can be rewritten as follows:



Finding a Minimal Assumption (cont.)

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



- 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 $P \cup M1$.

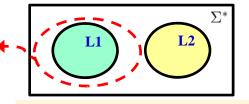
First observed by Gupta, McMillan, and Fu



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

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



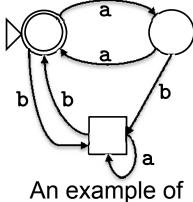
We say that A is a separating

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: $?\subseteq L1$, $?\subseteq L1$, $?\subseteq L2$, and $?\supseteq L2$

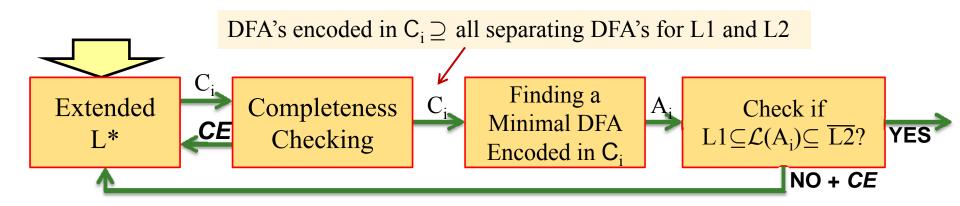
3-Value DFA (3DFA)

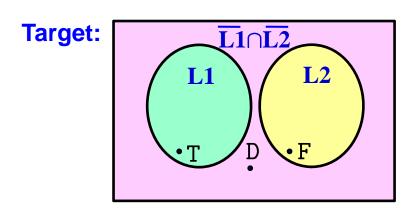
- A 3DFA is a tuple $C = (\Sigma, S, s_0, \delta, Acc, Rej, Dont)$.
- \blacksquare A DFA A is encoded in a 3DFA C iff A
 - accepts all strings that C accepts and
 - \blacksquare rejects all strings that C rejects.
 - \blacksquare A don't care string in C can be either accepted or rejected by A.



an example of a 3DFA

The L^{Sep} Algorithm: Overview





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

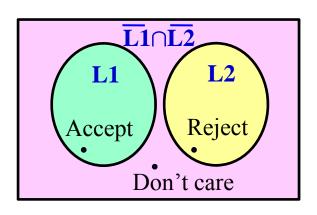
Extend the L*

algorithm to allow don't care values.

The Target 3DFA

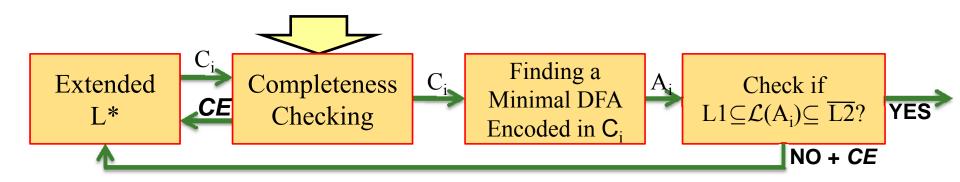
- The target 3DFA *C*
 - **accepts** every string in L1, and
- DFA's encoded in C = all separating DFA's for L1 and L2

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

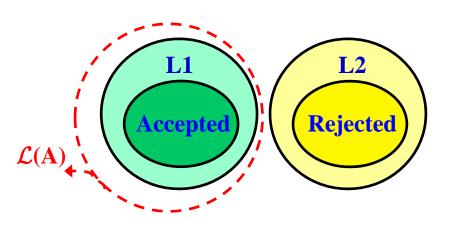


Definition:

- **A DFA** A is **encoded in** a **3DFA** C iff A
 - accepts all strings that *C* accepts and
 - rejects all strings that *C* rejects.
- \blacksquare 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**.

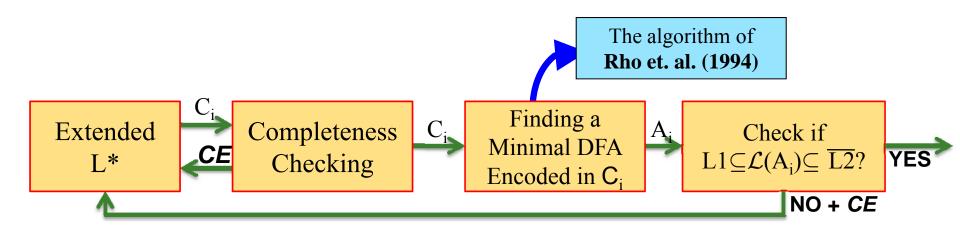


Check if all of the **separating** DFA's of L1 and L2 are **encoded** in C_i , which can be done by checking the following conditions:



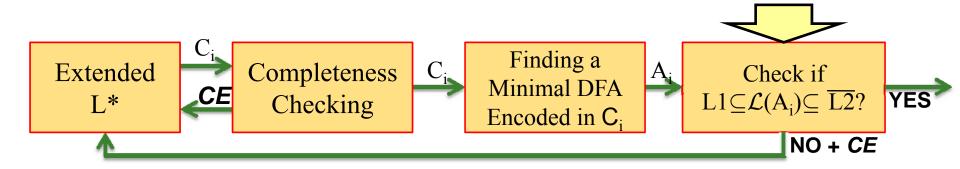
Definition:

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



LEMMA:

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



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

A_i is a minimal separating DFA.

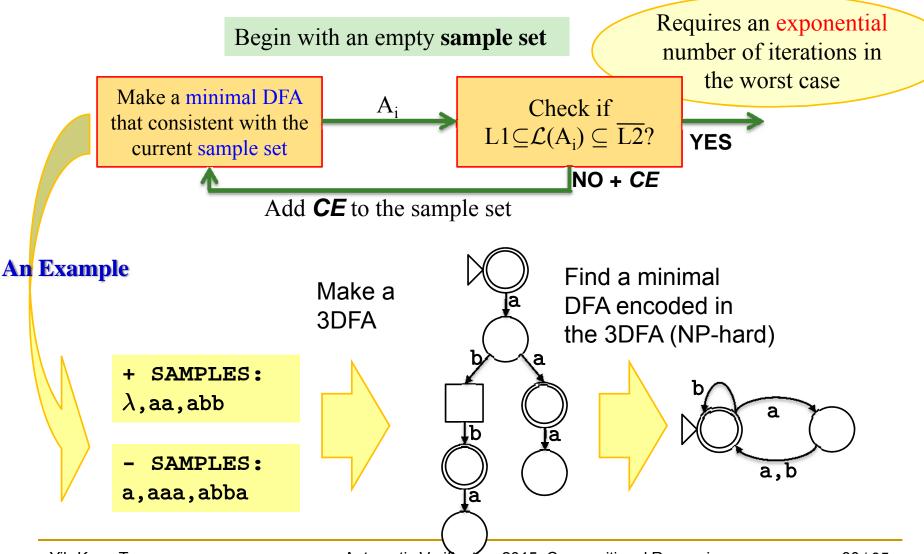
If L1
$$\nsubseteq \mathcal{L}(A_i)$$
 or $\mathcal{L}(A_i) \nsubseteq \overline{L2}$:

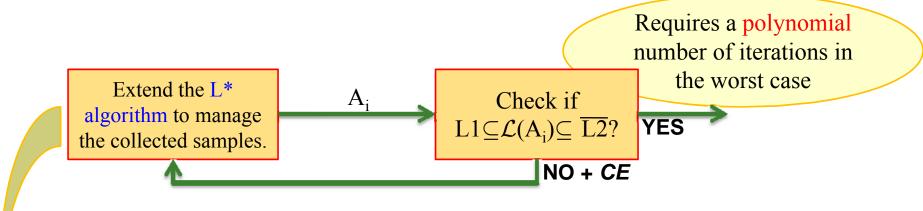
Counterexample CE is a witness for C_i not being the target 3DFA.

LEMMA:

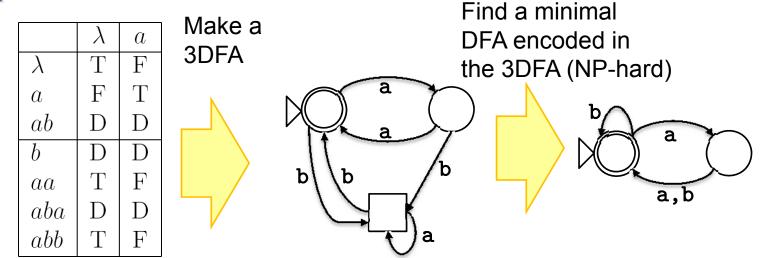
The size of **minimal separating DFA** of L1 and L2 \geq $|A_i|$, the size of the **minimal DFA encoded in C**i.

The Algorithm of Gupta et al.

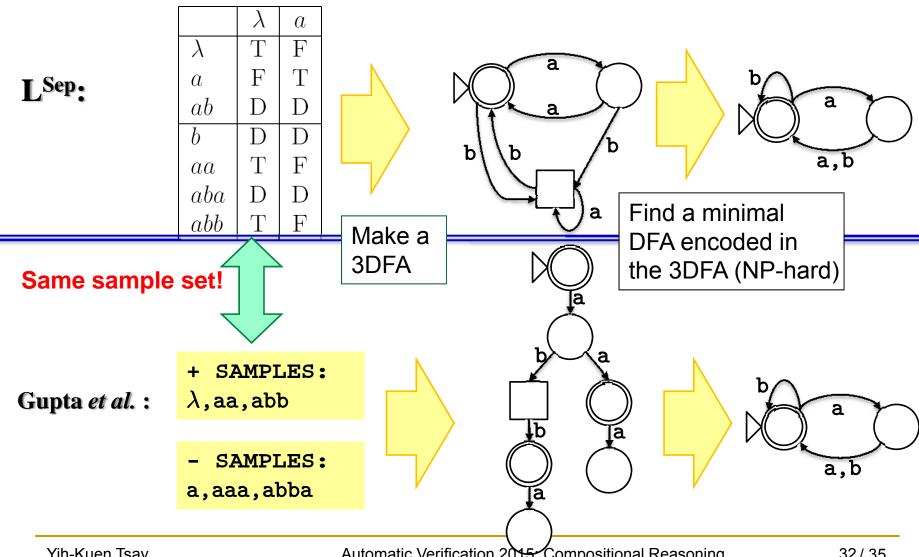




An Example



Comparing the Two Algorithms





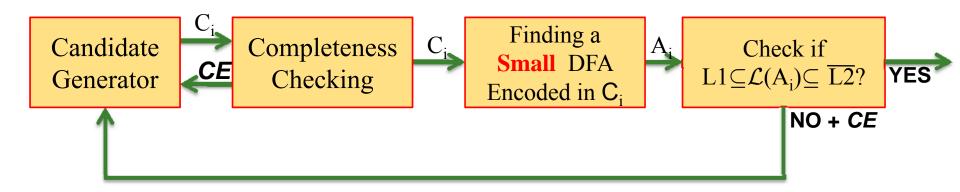
Adapt L^{Sep} for Compositional Verification

Let L1 = M2 and $\overline{L2} = P \cup M1$, use L^{Sep} to find a separating DFA for L1 and L2.

■ When $M2 \nsubseteq P \cup M1$ (i.e., $M1 \cap M2 \nsubseteq P$), L^{Sep} can be modified to guarantee finding a string in M2, but not in $P \cup \overline{M1}$ (i.e., $M1 \cap M2 \setminus P$).

Adapt L^{Sep} for Compositional Verification

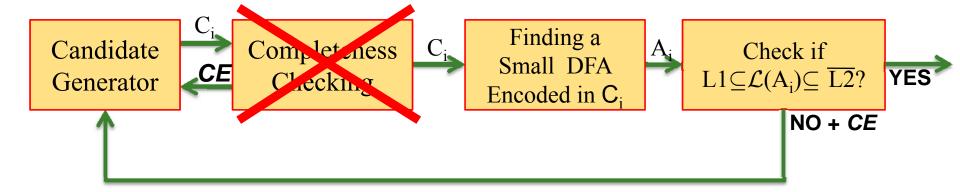
Use heuristics to find a small consistent DFA:



Minimality is no longer guaranteed!

Adapt L^{Sep} for Compositional Verification

Skip completeness checking:



Minimality is no longer guaranteed!