

Vulnerability Detection in Mobile Applications using State Machine Modeling

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TUDelft
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Security Protocol Implementations:
Development and Analysis
(SPIDA)

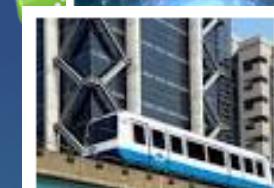
Introduction

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[WhatsApp Security Flaw Lets Hackers Enter Any Group Unnoticed](#)
Vulnerability in mAadhaar Android app allows anyone to steal your ...

Diait - 11 jan. 2018



[Researchers find 147 vulnerabilities in 34 SCADA mobile applications](#)
SC Magazine - 11 jan. 2018

covered how
a French security

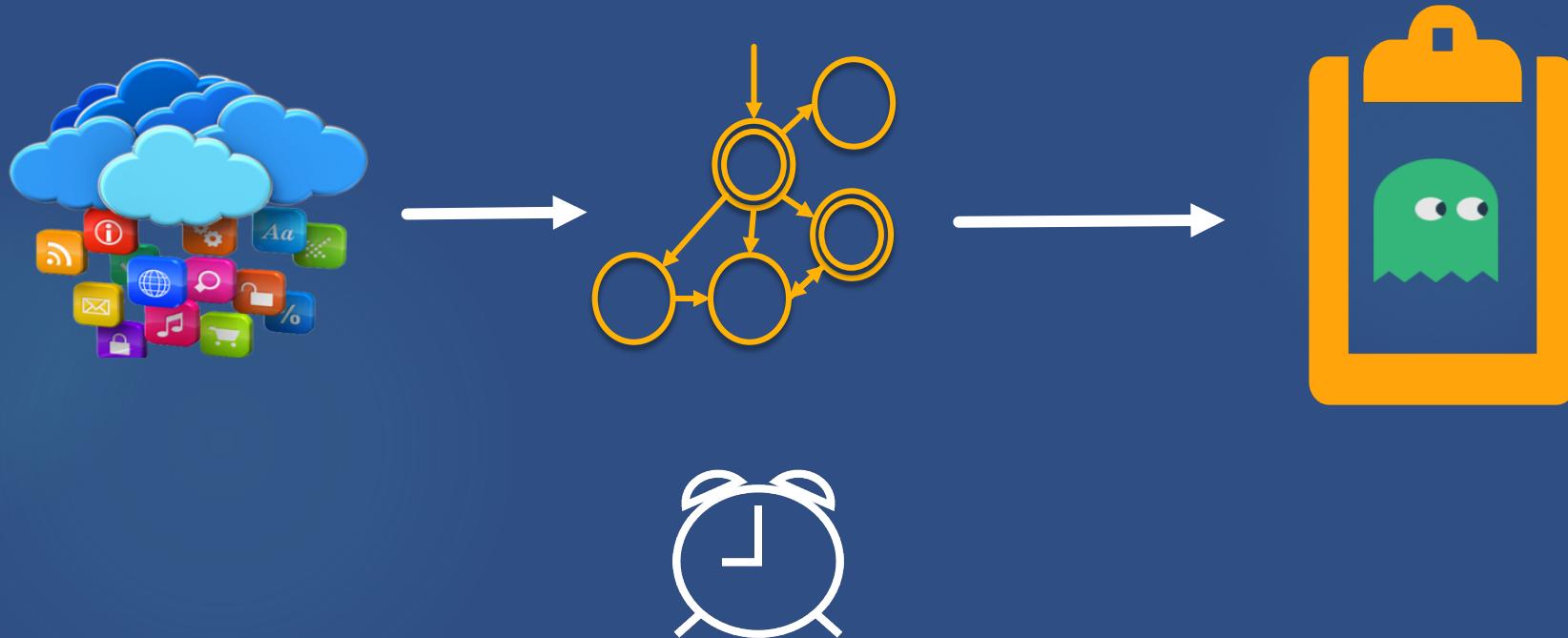
IoActive and Embedi researchers released a whitepaper outlining 147 vulnerabilities in basic
in 34 mobile applications used in tandem with Supervisory Control and Data
Acquisition (SCADA) systems. The vulnerabilities could allow an attacker to ...



[It's Open Season for Hackers: 90% of Mobile Cryptocurrency Apps ...](#)
Cryptovest (Cryptocurrency & Blockchain News) - 3 uur geleden

What's probably more shocking than all of this is that the analysis they performed
found that 77 percent of popular cryptocurrency applications have at least two high-
risk vulnerabilities, making the mobile space a massive party venue for hackers.
At the beginning of this month, the CryptoShuffle virus ...

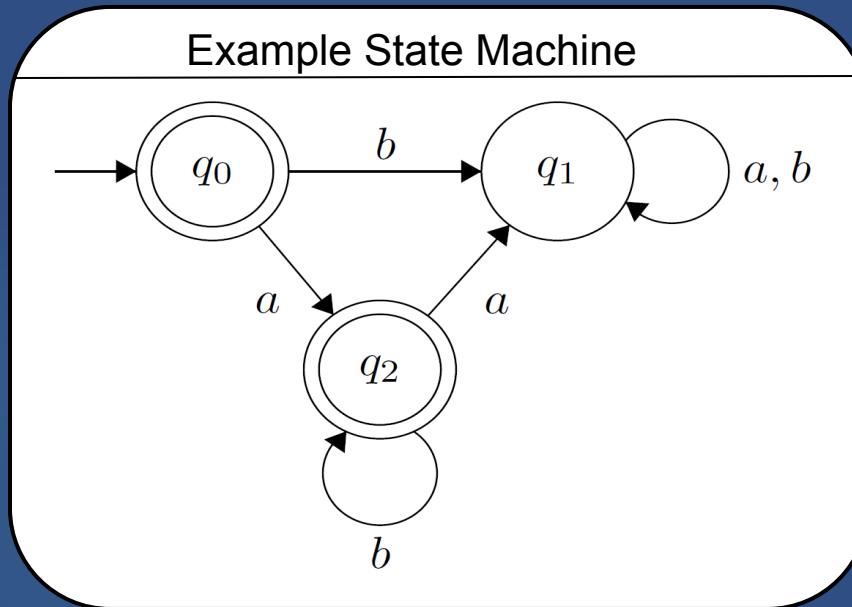
Research Goal



Outline

- ▶ State Machine Learning
- ▶ Mobile State Machine Learning
- ▶ Vulnerability Detection
- ▶ Results
- ▶ Conclusion

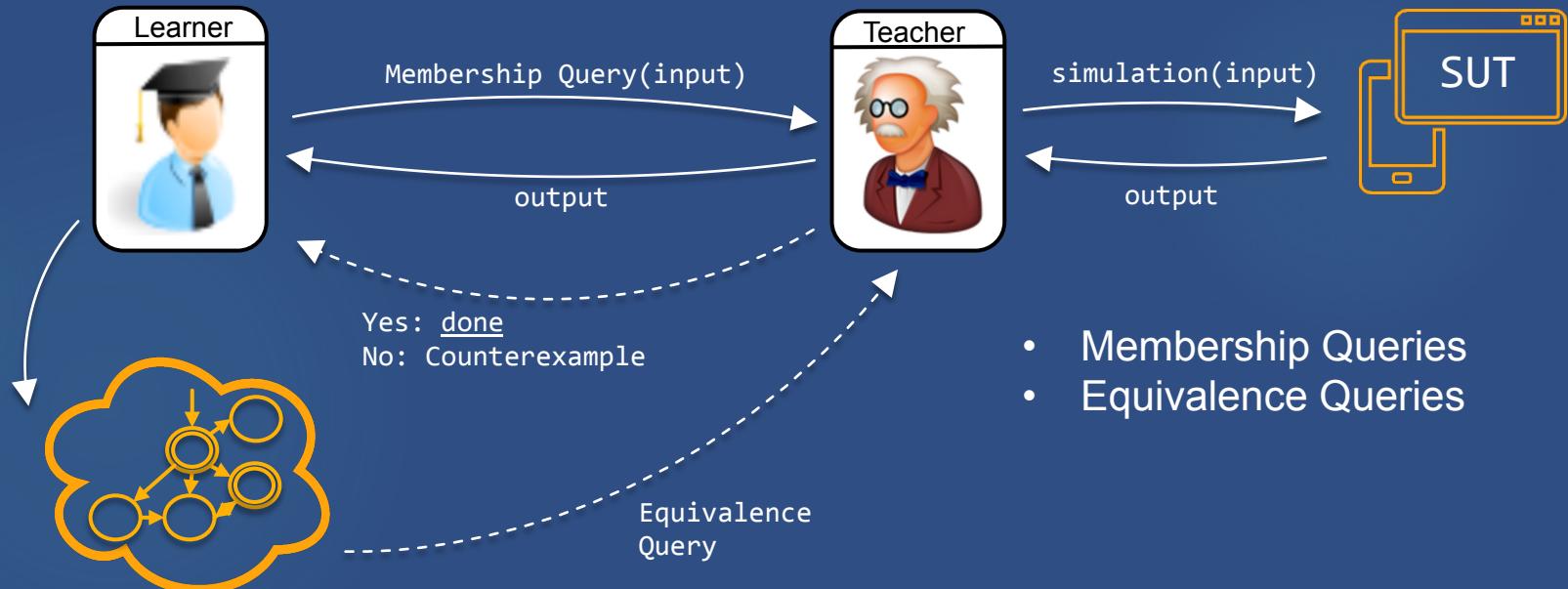
Definition of a State Machine



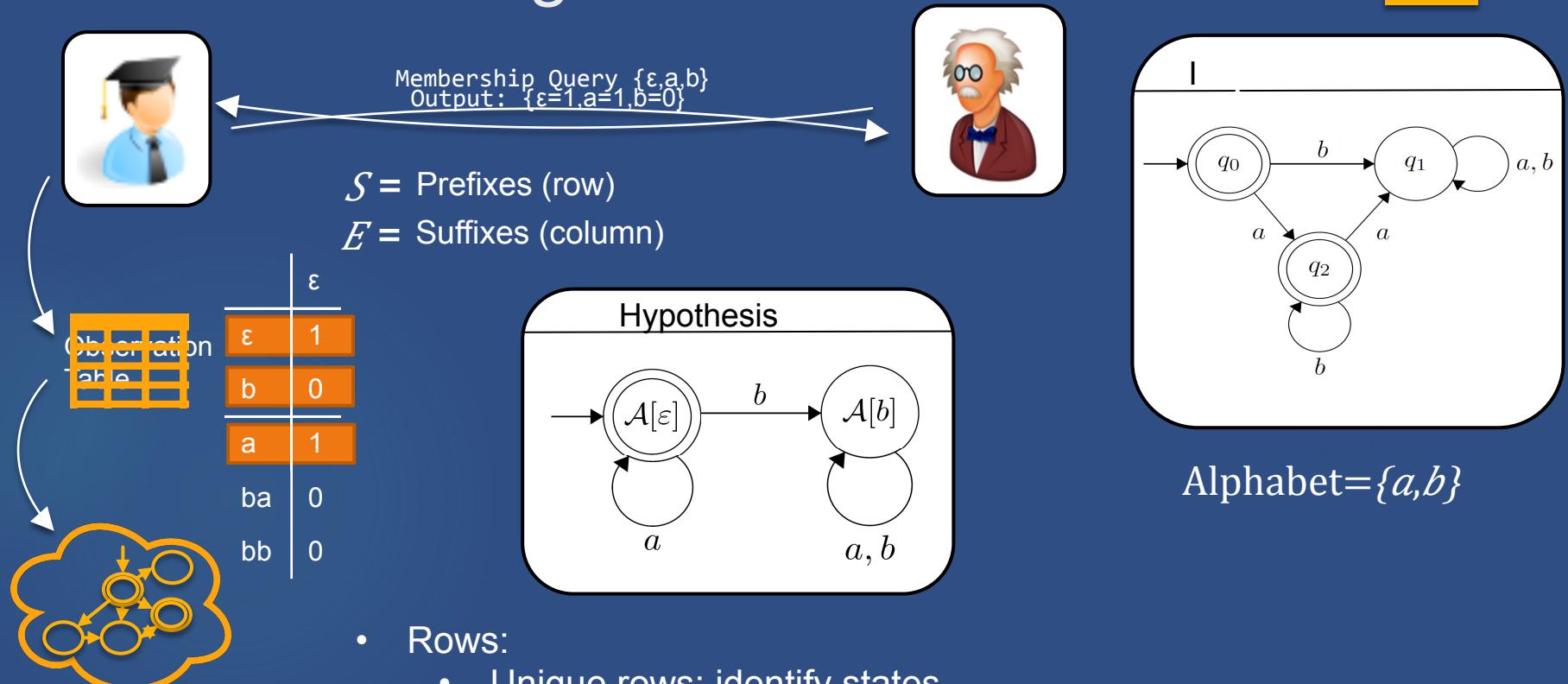
- States= $\{q_0, q_1, q_2\}$
- Start= q_0
- Transitions = 
- Alphabet= $\{a, b\}$
- Accepting states= $\{q_0, q_2\}$

MAT Framework

► Minimally Adequate Teacher (Angluin, 1988)



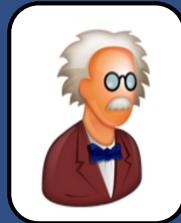
Active Learning with L*



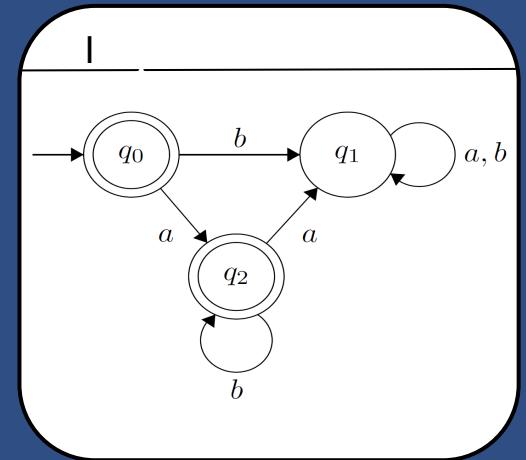
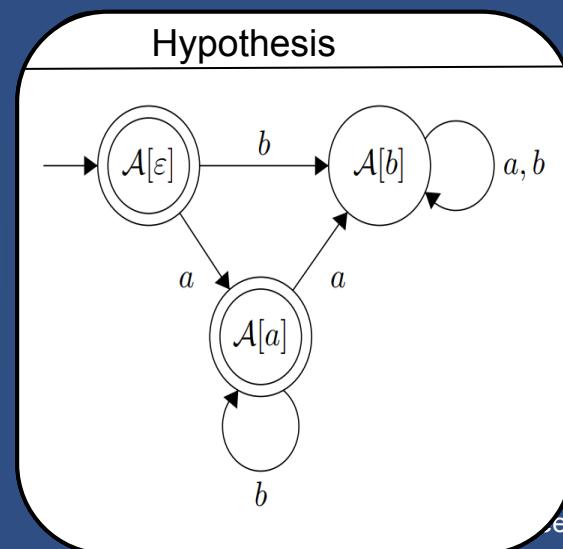
Active Learning with L*



NO YES Found equivalence: $pb \Downarrow$

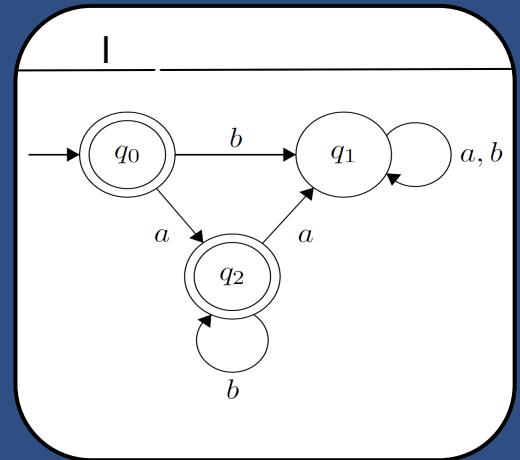
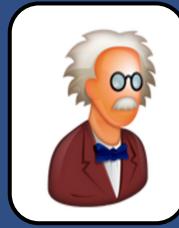


	ε	b
ε	1	0
b	0	
a	1	1
ba	0	
bb	0	
aa	0	
ab	1	1



Equivalence Queries

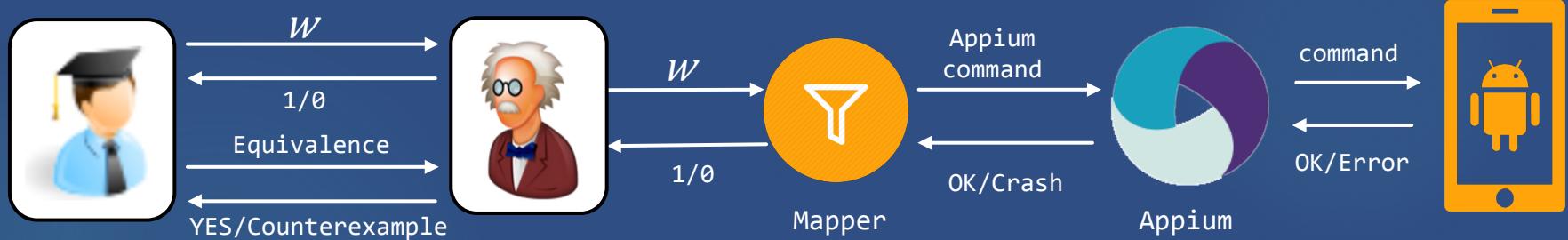
- ▶ Conformance Testing
 - ▶ Random Walk
 - ▶ HappyFlow
- ▶ W-Method
 - ▶ $P \times \Sigma^* n \times W$ (Chow, 1978)
 - ▶ $P \times \Sigma^* n \times W'$ (Smetsers et al., 2016)



Chow, T.S. "Testing software design modeled by finite-state machines." *IEEE transactions on software engineering* 3 (1978): 178-187.

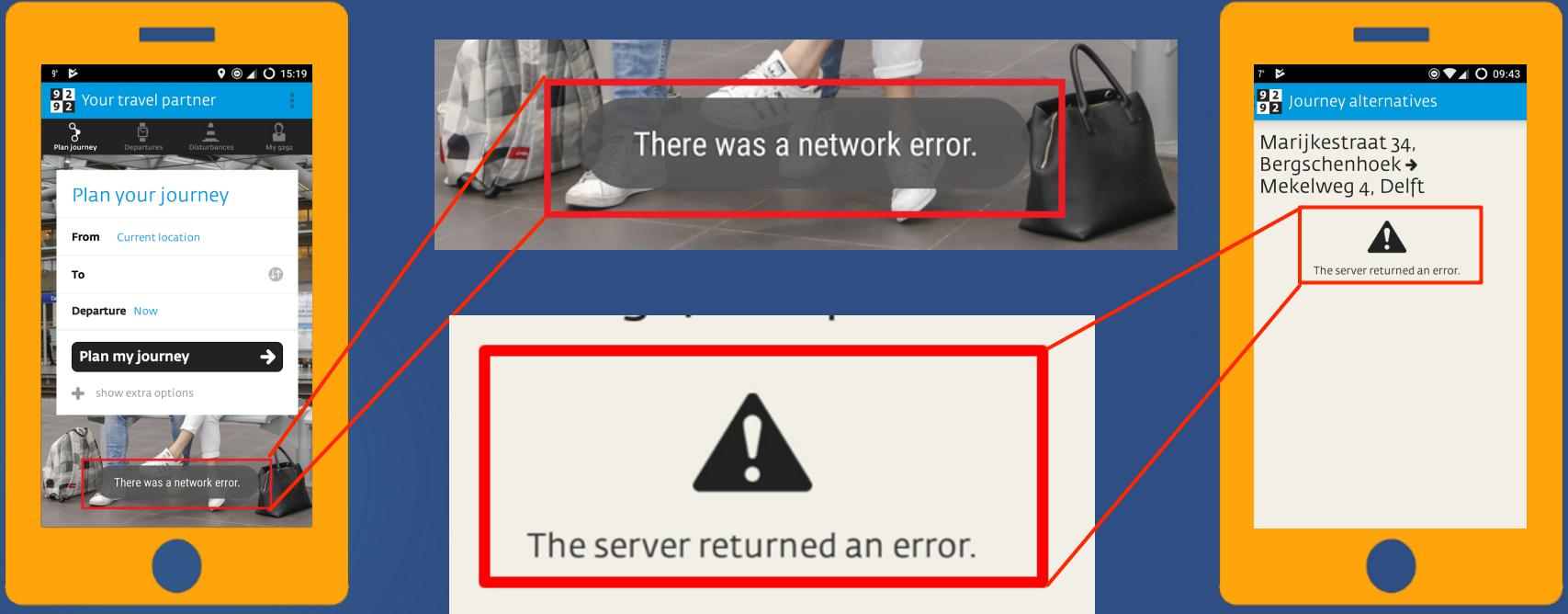
Smetsers, R. et al. "Minimal separating sequences for all pairs of states." *International Conference on Language and Automata Theory and Applications*. Springer International Publishing, 2016.

Mobile State Machine Learning



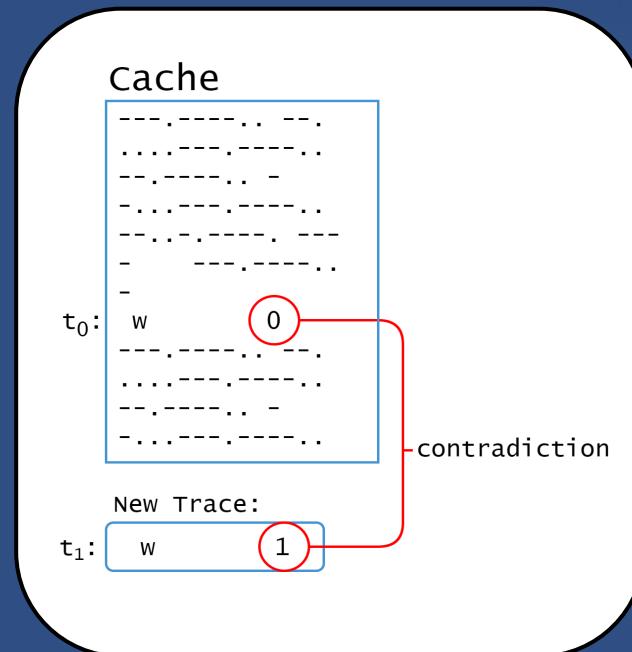
Problems when learning

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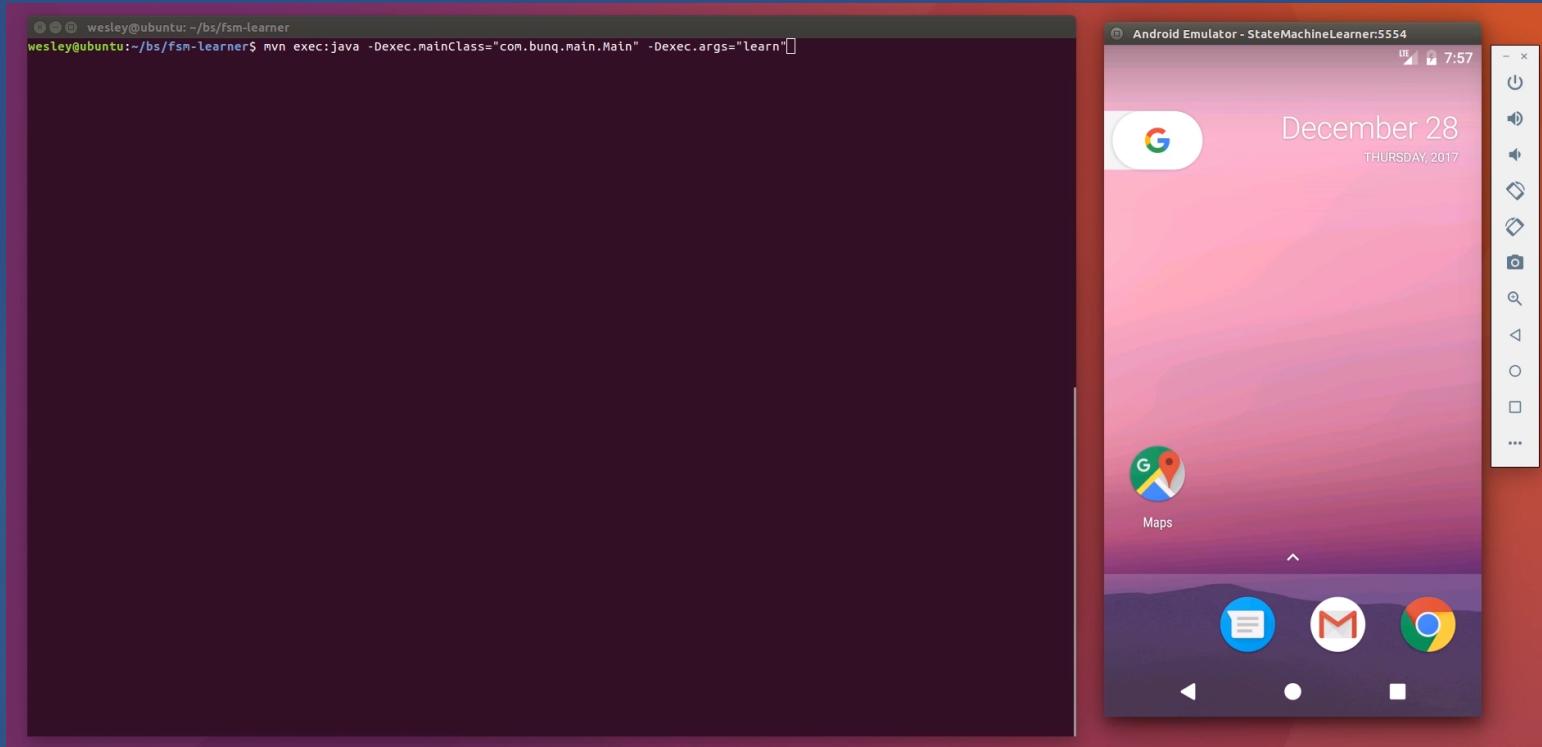
Problems when learning

- ▶ Non-deterministic behavior
- ▶ Inconsistent cache
- ▶ Cache Roll-Back



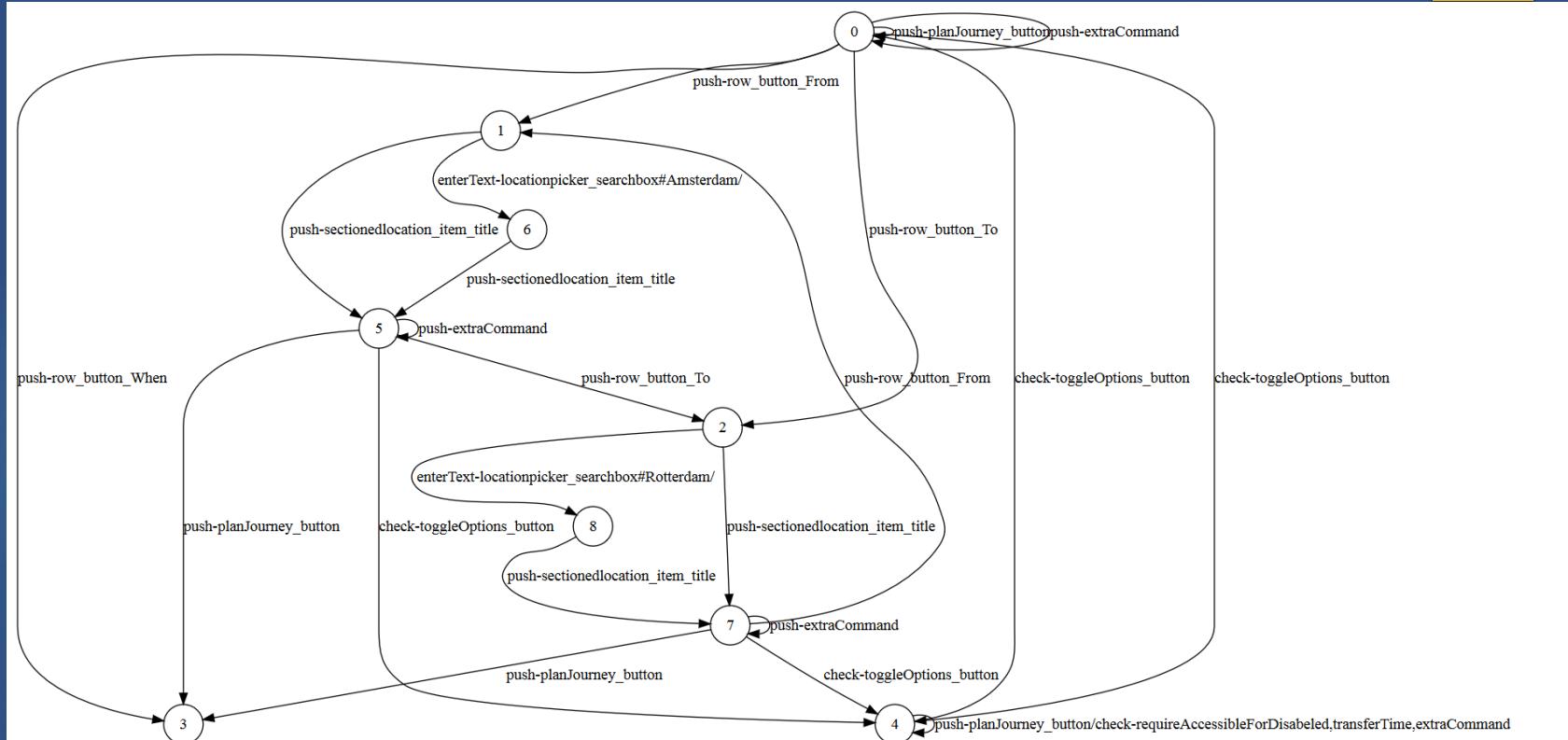
Video

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9292 Model

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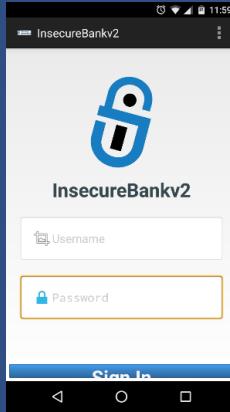
Vulnerability Detection

- ▶ Enrich Models
- ▶ Define Vulnerability Algorithms
 - ▶ OWASP Top10 Mobile
 - ▶ i.e. Insecure Authentication:

Does there exist a path to a node after the login state without traversing the login state?

Results

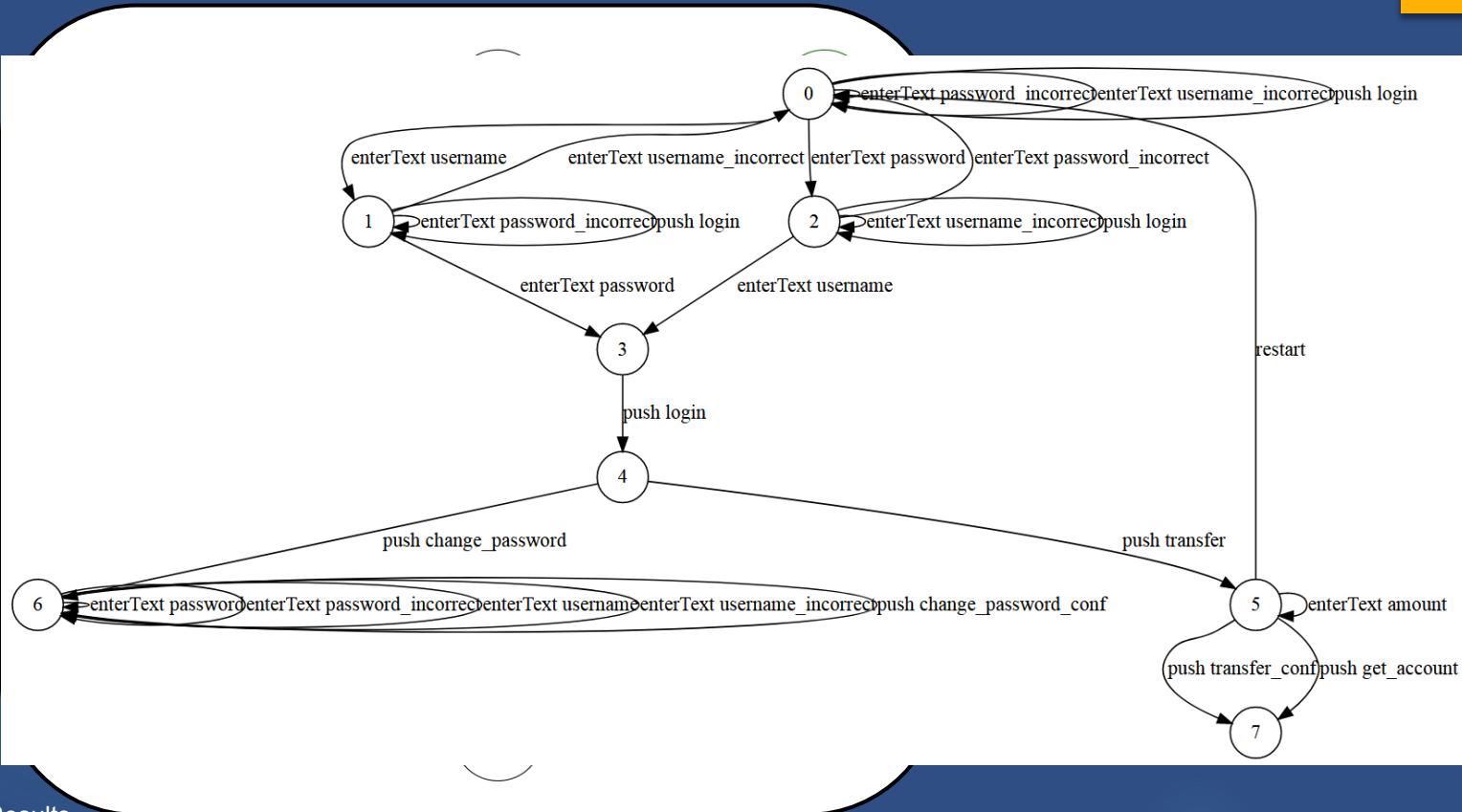
- ▶ InsecureBankV2
 - ▶ Known Vulnerable App
- ▶ Fake WhatsApp
 - ▶ Adware
 - ▶ Malware Spreading



InsecureBankV2

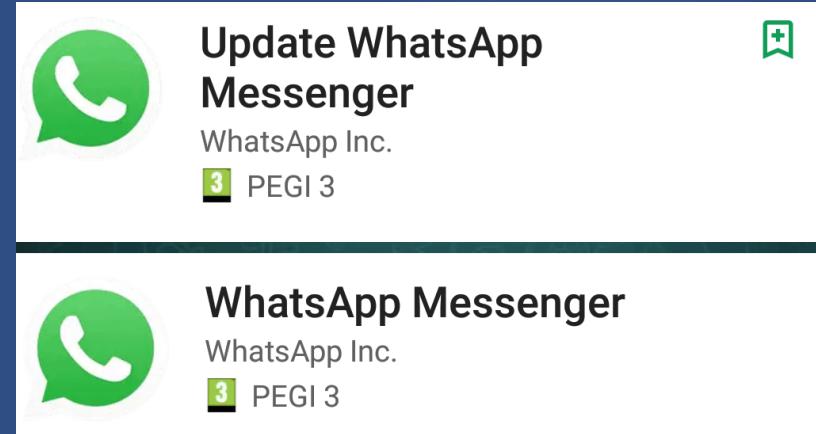
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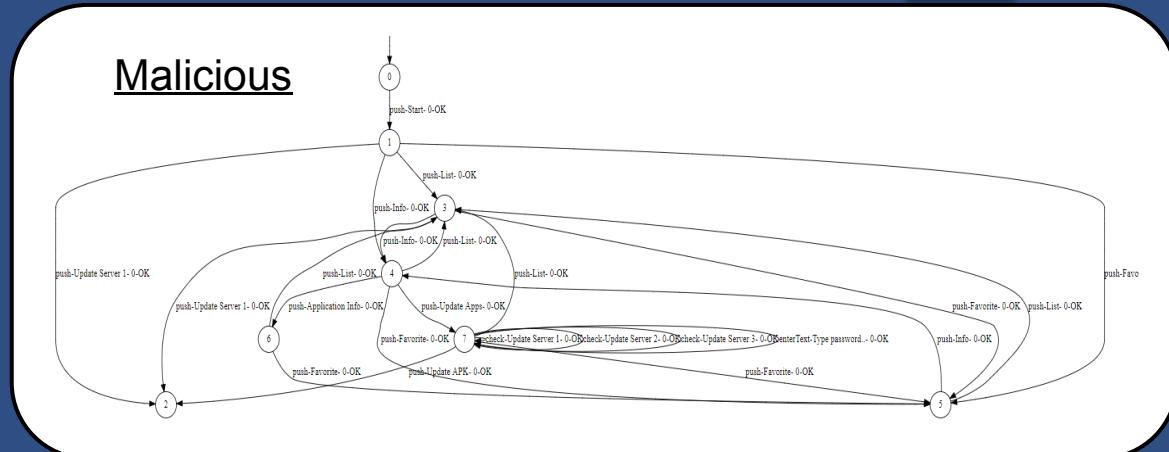
Fake WhatsApp

Background: Unicode trick allows malicious developers to impersonate WhatsApp developers
> 1M downloads, now deleted from Play Store

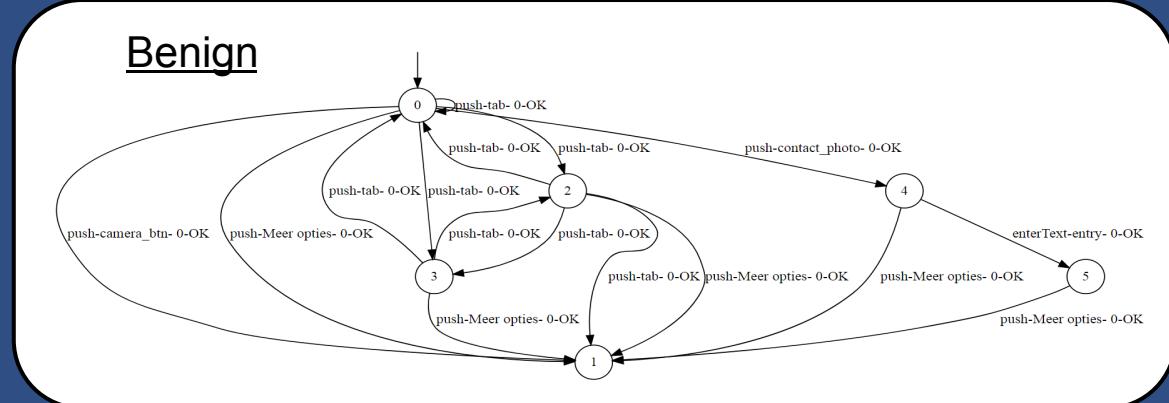


Fake WhatsApp

- Extraneous Functionality
 - SUT > REF: 100%
 - REF > SUT: 100%



■ Insecure Communication



Run statistics

Learning Algorithm	L*
Equivalence Oracle	RandomWalk
Membership Queries	1778
Equivalence Queries	1
States	9
Transitions	25
Learning Time	26:06 (hh : mm)

Tab. 4.1: Statistics for active learning the inferred state machine for the 9292 application using L* and RandomWalk

Learning Algorithm	TTT
Equivalence Oracle	RandomWalk
Membership Queries	55
Equivalence Queries	2
States	2
Transitions	4
Learning Time	00:21 (hh : mm)

Tab. 4.2: Statistics for active learning the inferred machine for the 9292 application using TTT and RandomWalk

Learning Algorithm	TTT	L*
Equivalence Oracle	WMethod-Minimal	WMethod-Minimal
Membership Queries	15619	16966
Equivalence Queries	10	2
States	10	10
Transitions	30	30
Learning Time	23:14 (hh : mm)	27:37 (hh : mm)

Tab. 4.3: Statistics for Active Learning the Inferred Machine for the 9292 application using TTT, L* and WMethod-Minimal

Run statistics

InsecureBankv2

InsecureBankv2	
Learning Time	3:13
Learning Algorithm	TTT
Equivalence Oracle	WMethod-Minimal
Equivalence Queries	7
Membership Queries	11627
States	6
Alphabet Size	6
Cache Hit	49%
Fast Forwarded	42%

Tab. 6.1: Learning statistics for InsecureBankv2

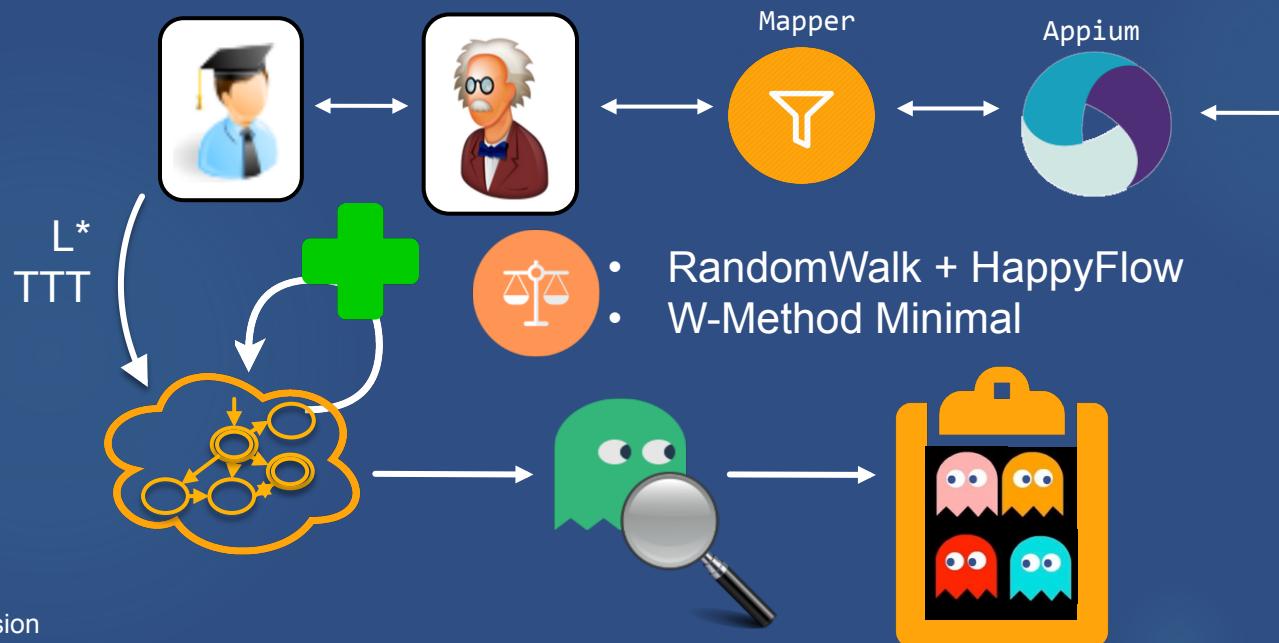
Fake WhatsApp Real WhatsApp

	Fake WhatsApp	Real WhatsApp
Learning Time	7:22	5:47
Learning Algorithm	TTT	TTT
Equivalence Oracle	WMethod-Minimal	WMethod-Minimal
Equivalence Queries	7	6
Membership Queries	29.751	25.083
States	22	11
Alphabet Size	12	11
Cache Hit	71%	69%
Fast Forwarded	67%	62%

Tab. 6.3: Learning statistics for both WhatsApp versions

Conclusion

How can one identify weaknesses in mobile Android applications through feasible behavioral state machine learning?



Future work

- ▶ It is feasible, but slow, possible speedups:
 - ▶ Obtaining the possible actions without testing
 - ▶ Using other information, e.g. the screen content, during learning
 - ▶ ...
- ▶ Performing the security analysis using a model checker
- ▶ Solving inconsistencies caused by popups or the keyboard not closing

Thank You

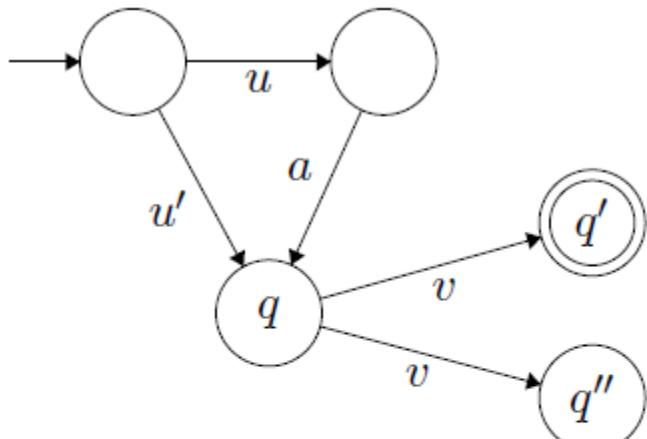




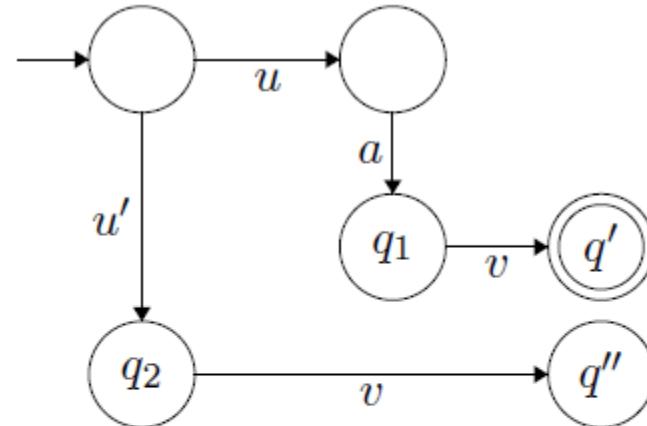
```

1   $S = E = \{\lambda\}$ 
2   $(S, E, T) \leftarrow MQ(\lambda) \cup \forall a \in A : MQ(a) \# (S, E, T)$  is the observation table
3
4  While  $M$  is incorrect: #  $M$  is the conjecture
5    While  $(S, E, T)$  is not consistent or not closed
6      if  $(S, E, T)$  is not consistent:
7         $\exists (s_1, s_2) \in S, a \in A, e \in E : \text{row}(s_1) = \text{row}(s_2)$  and  $T(s_1 \cdot a \cdot e) \neq T(s_2 \cdot a \cdot e)$ 
8         $E \leftarrow a \cdot e$ 
9        extend  $T$  to  $(S \cup S \cdot A) \cdot E$  using  $MQ$ 
10     if  $(S, E, T)$  is not closed:
11        $\exists s_1 \in S, a \in A : \forall \text{row}(s_1 \cdot a) \neq \text{row}(s)$ 
12        $S \leftarrow s_1 \cdot a$ 
13       extend  $T$  to  $(S \cup S \cdot A) \cdot E$  using  $MQ$ 
14    $M = M(S, E, T)$  #  $(S, E, T)$  is closed and consistent
15   if Teacher replies with a counter-example  $t$ :
16     add  $t$  and all its prefixes to  $S$ 
17     extend  $T$  to  $(S \cup S \cdot A) \cdot E$  using  $MQ$ 
18 Halt and output  $M$ 

```



(a)



(b)

Fig. 2.4: Formal progression of an incorrect conjecture: (a) inconsistent model for distinguishing suffix v from state q , (b) consistent model after splitting q into new states q_1 and q_2 .

	Improper Platform Usage	Insecure Communication	Insecure Authentication	Extraneous Functionality
InsecureBank	+/-	+/-	+/-	-/-
FreeCola	+/-	-/-	-/+	-/-
9292	-/-	-/-	-/-	-/-
Benign WhatsApp	-/-	-/-	-/-	-/-
Malicious WhatsApp	+/-	+/-	-/-	+/-

Tab. 7.2: The expected results versus the expected results for all applications under test.

Algorithm 7 Improper Platform Usage Identification

Input: Inferred State Machine $M = \{Q, \Sigma, \delta, q_0, F\}$

Output: Returns a set R of activities that induce supplementary behavior in machine M . There possibly exists a vulnerability if R is non-empty.

```
1:  $R \leftarrow \emptyset$ 
2:  $A \leftarrow$  activities from  $M$ 
3: for all  $a$  in  $A$  do
4:   if  $a$  is a callable activity then
5:     add  $a$  to  $R$ 
6:   end if
7: end for                                 $\triangleright R$  contains all callable activities
8: for all  $q$  in  $Q$  and  $R$  is not empty do
9:   remove  $q$ 's activity from  $R$ 
10: end for
11: return  $R$ 
```

Algorithm 8 Insecure Communication Identification

Input: Inferred State Machine $M = \{Q, \Sigma, \delta, q_0, F\}$

Output: Returns a set R of requests that made by actions in machine M that do not adhere to common network security standards. There possibly exists a vulnerability if R is non-empty.

```
1:  $R \leftarrow \emptyset$ 
2: for all  $t$  in  $\delta$  do
3:    $r \leftarrow$  request_insecure( $t.\text{request}$ )
4:   if  $r$  then
5:     add  $r$  to  $R$ 
6:   end if
7: end for
8: return  $R$ 
```

Algorithm 9 Insecure Authentication Identification

Input: Inferred State Machine $M = \{Q, \Sigma, \delta, q_0, F\}$
Output: Returns a set R of authentication bypass techniques in machine M .
There possibly exists a vulnerability if R is non-empty.

```
1:  $a \leftarrow$  authentication state of  $M$                                  $\triangleright a \in Q$ 
2: if  $a$  is null then       $\triangleright$  no authentication  $\rightarrow$  no authentication bypass
3:   return  $R$ 
4: end if
5:  $Marks \leftarrow$  subset of nodes possible to reach after  $a$ 
6:  $M' \leftarrow M - a$            $\triangleright$  the machine without the authentication state
7: for all  $m$  in  $Marks$  do
8:   if a path from the  $q_0$  to  $m$  exists in  $M'$  then
9:     add path to  $R$ 
10:  end if
11: end for
12:  $Q'' \leftarrow Q - Marks$ 
13:  $A \leftarrow$  callable activities
14: for all  $q$  in  $Q''$  and  $A$  is not empty do
15:   remove  $q$ 's activity from  $A$ 
16: end for
17: add  $A$  to  $R$ 
18: return  $R$ 
```

Algorithm 10 Code Tampering Identification

Input: Inferred machine $M = \{Q, \Sigma, \delta, q_0, F\}$ and reference machine $M' = \{Q', \Sigma', \delta', q'_0, F'\}$
Output: Finds difference in M and M' Returns a set R of sequences that yield a different output for the two machines M and M' . The sequences are divided into sets R_1 and R_2 which depict what machine models the sequence. There possibly exists a vulnerability if R is non-empty.

```
1:  $R_1, R_2 \leftarrow \emptyset$ 
2:  $TCS_1 \leftarrow TCS(M)$ 
3:  $TCS_2 \leftarrow TCS(M')$ 
4: for all  $w \in TCS_1$  do
5:   if  $\lambda^M(w) \neq \lambda^{M'}(w)$  then
6:      $R_1 \leftarrow w$ 
7:   end if
8: end for
9: for all  $w \in TCS_2$  do
10:  if  $\lambda^M(w) \neq \lambda^{M'}(w)$  and  $w \notin R$  then
11:     $R_2 \leftarrow w$ 
12:  end if
13: end for
14:  $R \leftarrow R_1, R_2$ 
15: return  $R$ 
```