



Vulnerability Detection in Mobile Applications using State Machine Modeling

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Security Protocol Implementations:
Development and Analysis
(SPIDA)

Introduction

1 / 20



WhatsApp Security Flaw Lets Hackers Enter Any Group Unnoticed

Vulnerability in mAadhaar Android app allows anyone to steal your ...

Digital - 11 jan. 2018



Researchers find 147 **vulnerabilities** in 34 SCADA **mobile** applications

SC Magazine - 11 jan. 2018

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

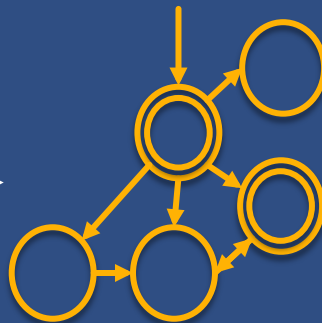


It's Open Season for Hackers: 90% of **Mobile** Cryptocurrency Apps ...
Cryptovest (Cryptocurrency & Blockchain News) - 3 uur geleden

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

3 / 20

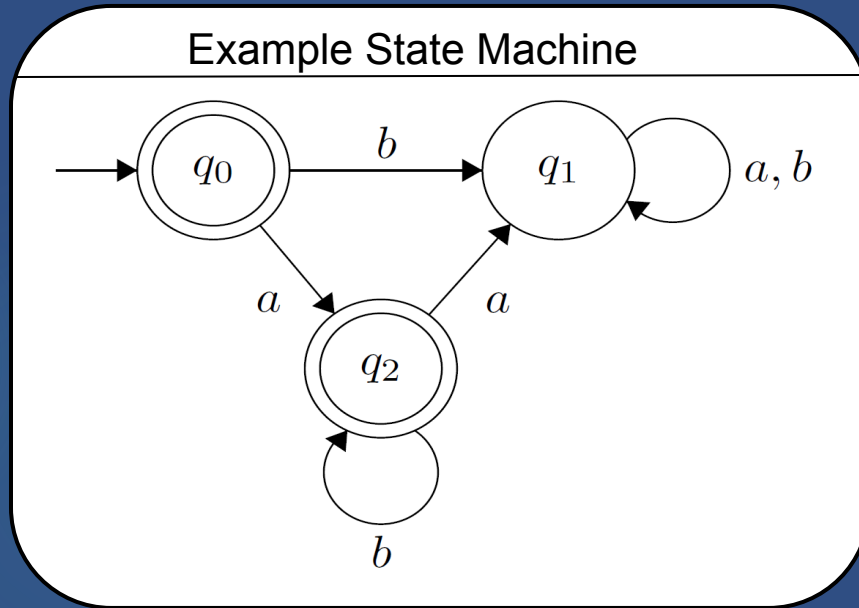


Outline

4 / 20

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

Definition of a State Machine



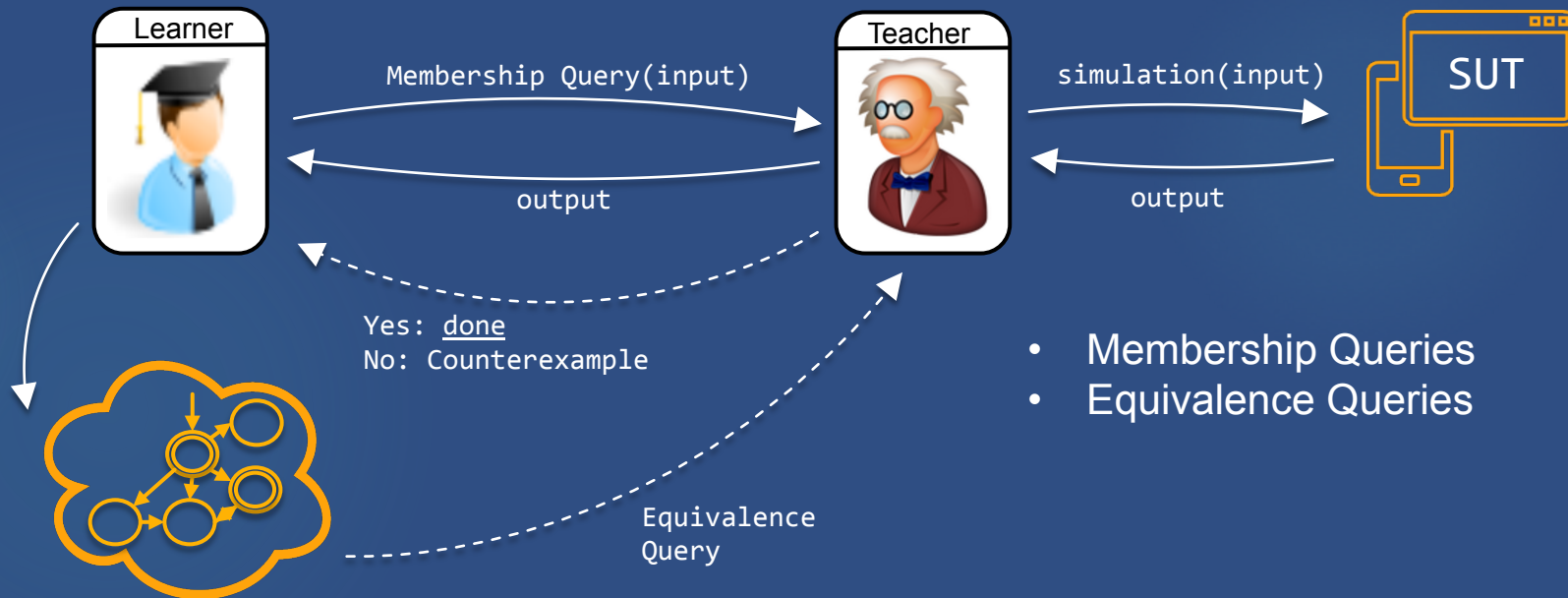
- States = $\{q_0, q_1, q_2\}$
- Start = q_0
- Transitions =

- Alphabet = $\{a, b\}$
- Accepting states = $\{q_0, q_2\}$

MAT Framework

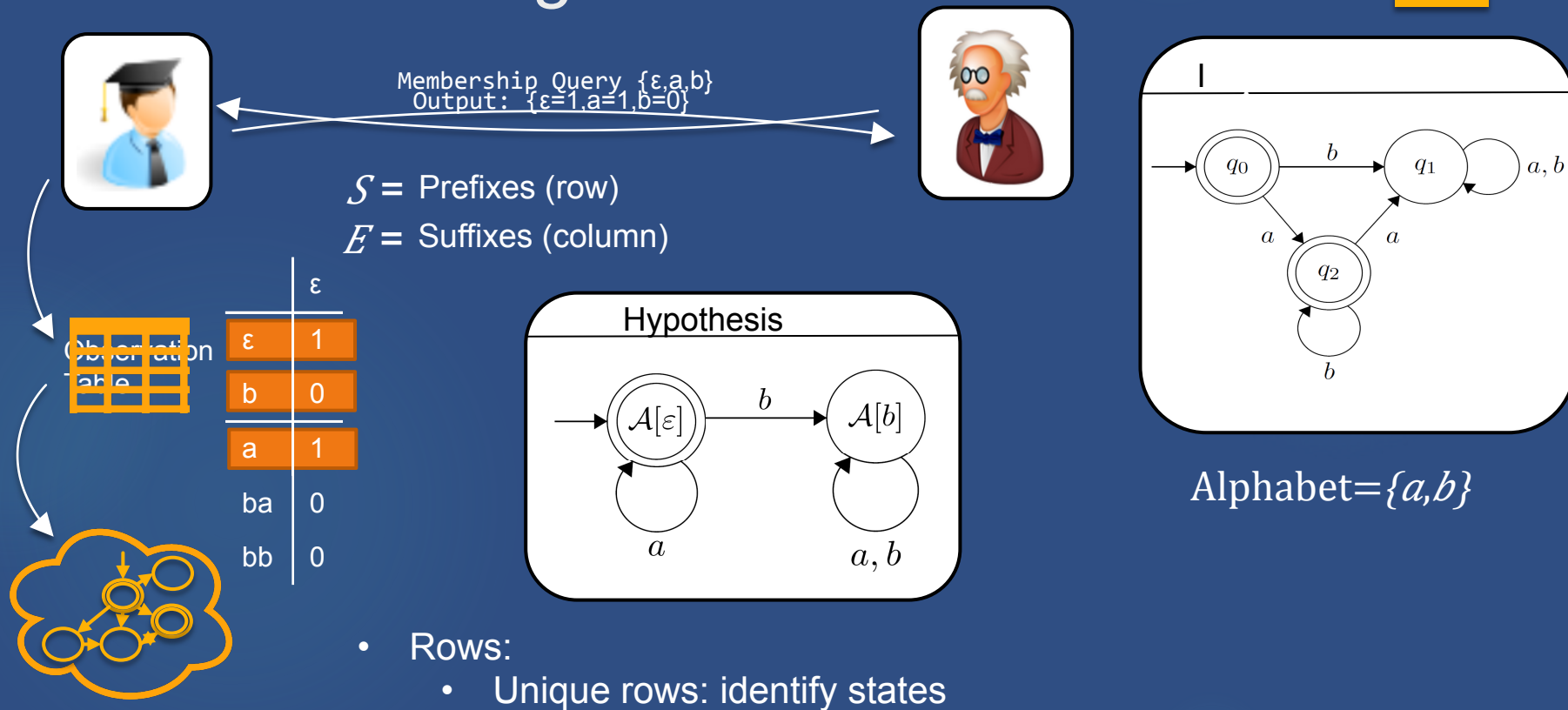
6 / 20

► Minimally Adequate Teacher (Angluin, 1988)



Active Learning with L^*

7 / 20



Active Learning with L^*

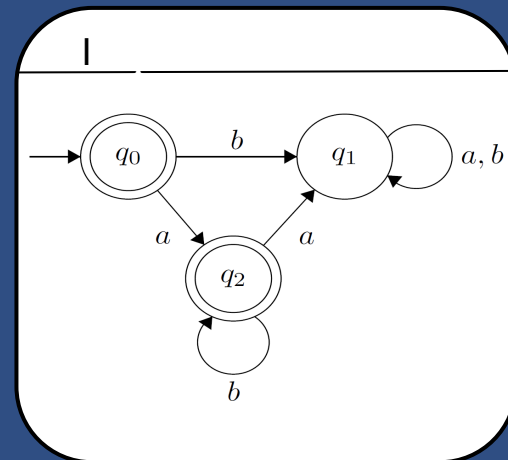
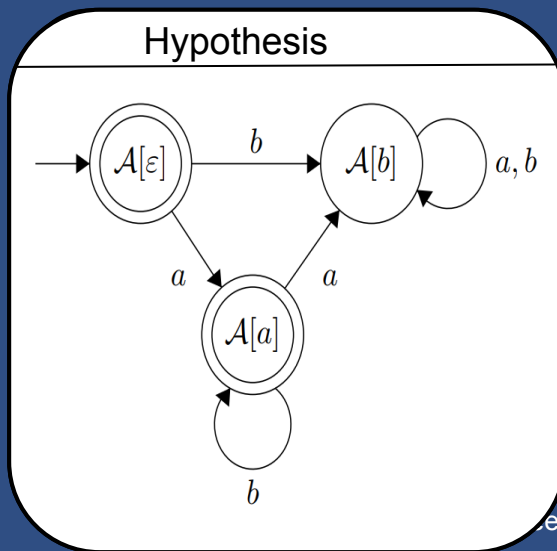
8 / 20



NO YES Surrogate: q_1
Equivalence Query $A \neq$

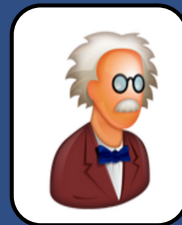


	ϵ	b
ϵ	1	0
b	0	
a	1	1
ba	0	
bb	0	
aa	0	
ab	1	1



Equivalence Queries

9 / 20



► Conformance Testing

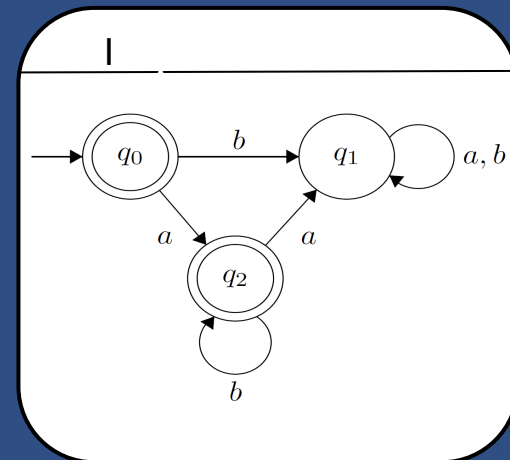
► Random Walk

► HappyFlow

► W-Method

► $P \times \Sigma^n \times W$ (Chow, 1978)

► $P \times \Sigma^n \times W'$ (Smetsters et al., 2016)

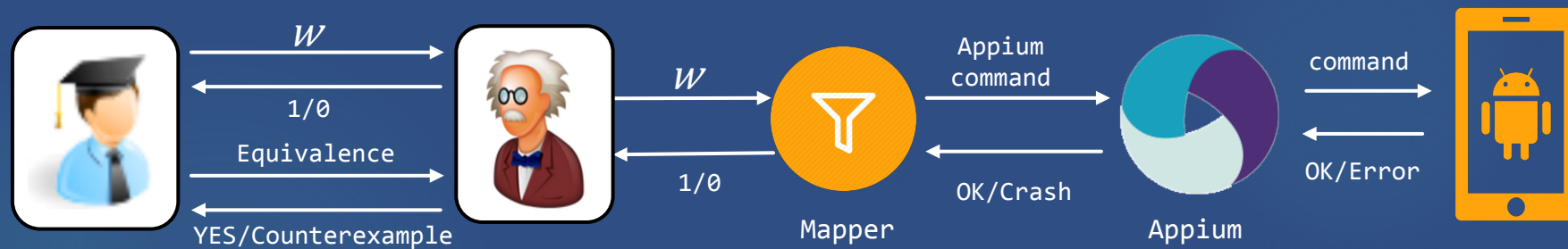


Chow, T.S. "Testing software design modeled by finite-state machines." IEEE transactions on software engineering 3 (1978): 178-187.

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

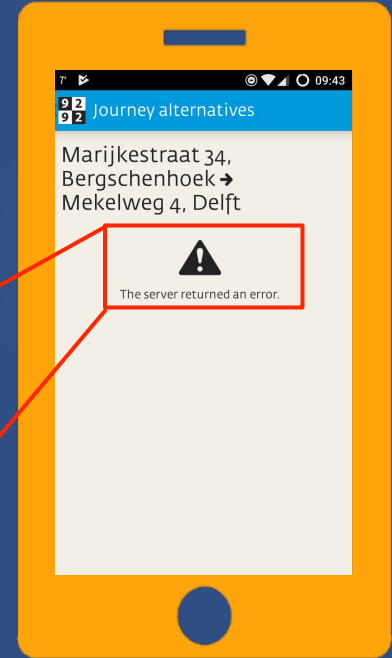
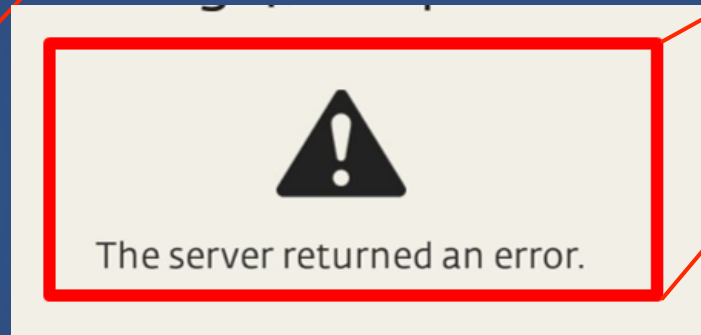
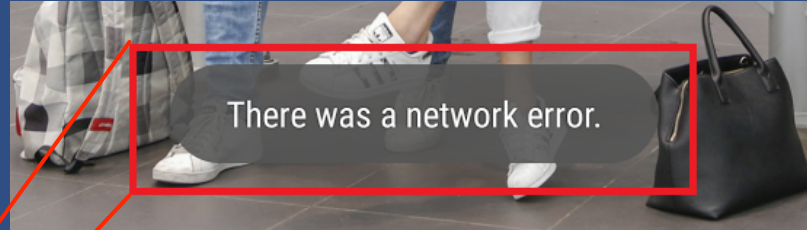
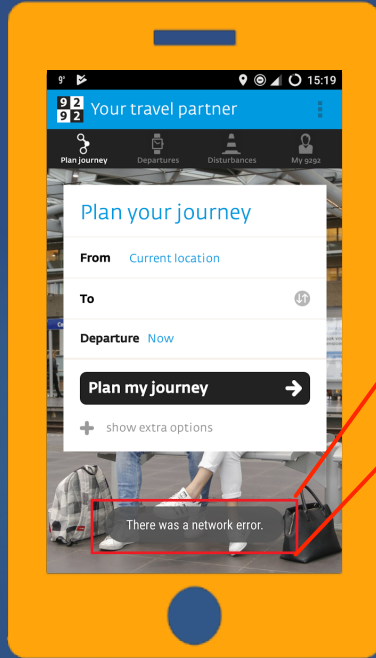
Mobile State Machine Learning

10 / 20



Problems when learning

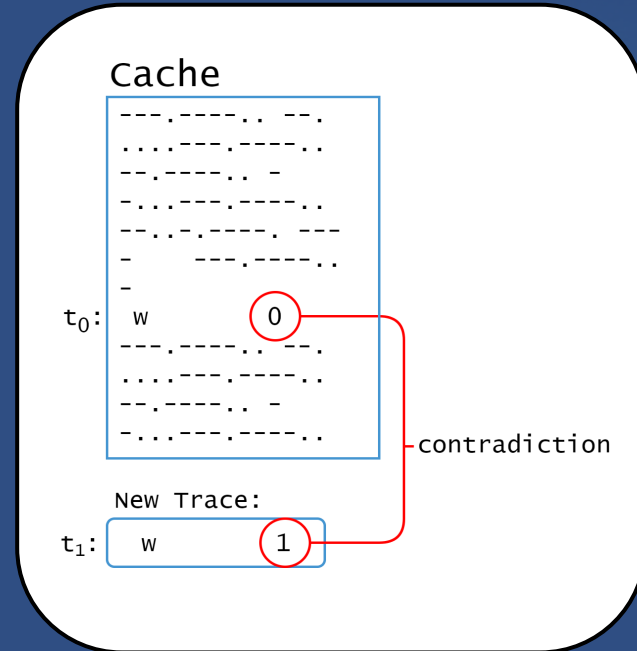
11 / 20



Problems when learning

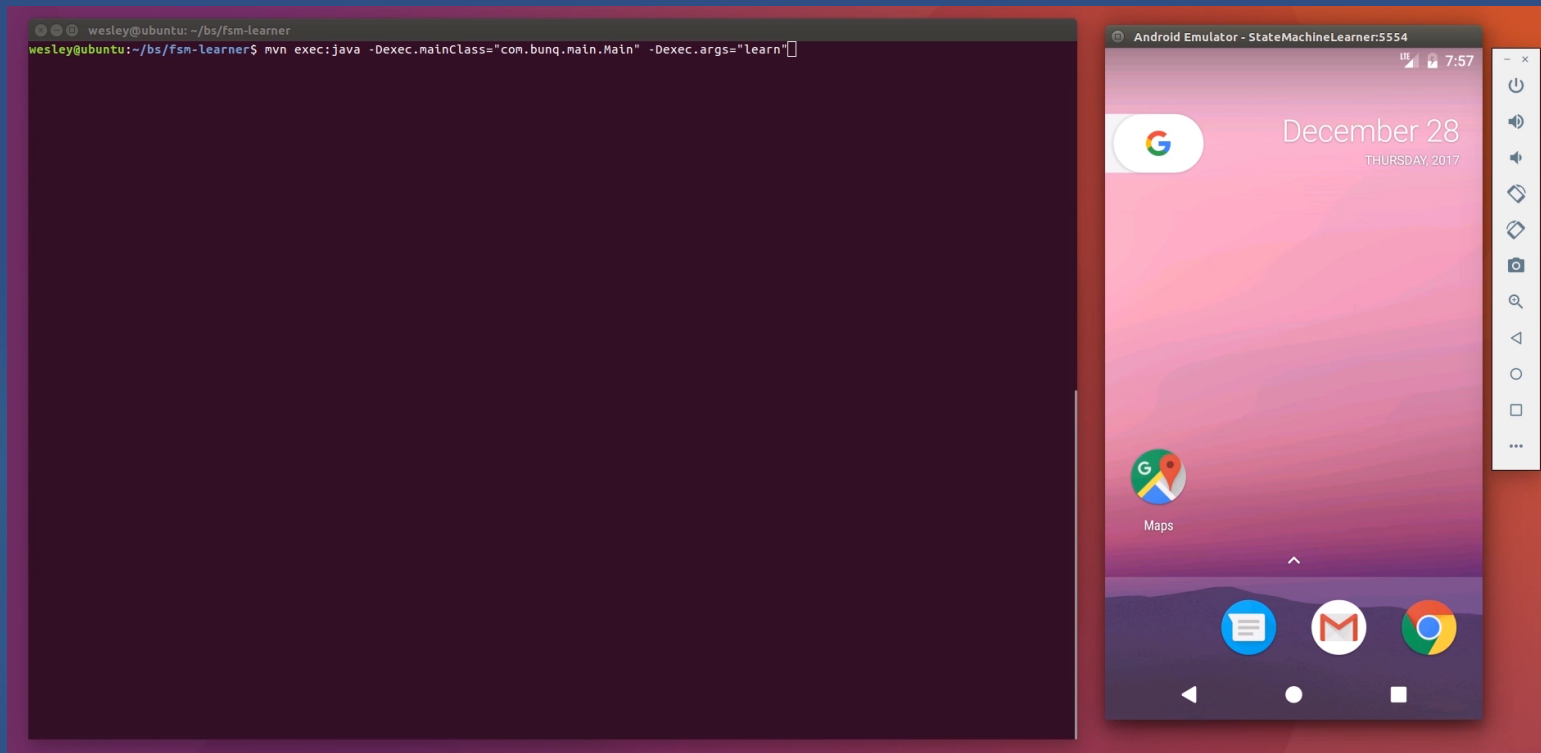
12 / 20

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



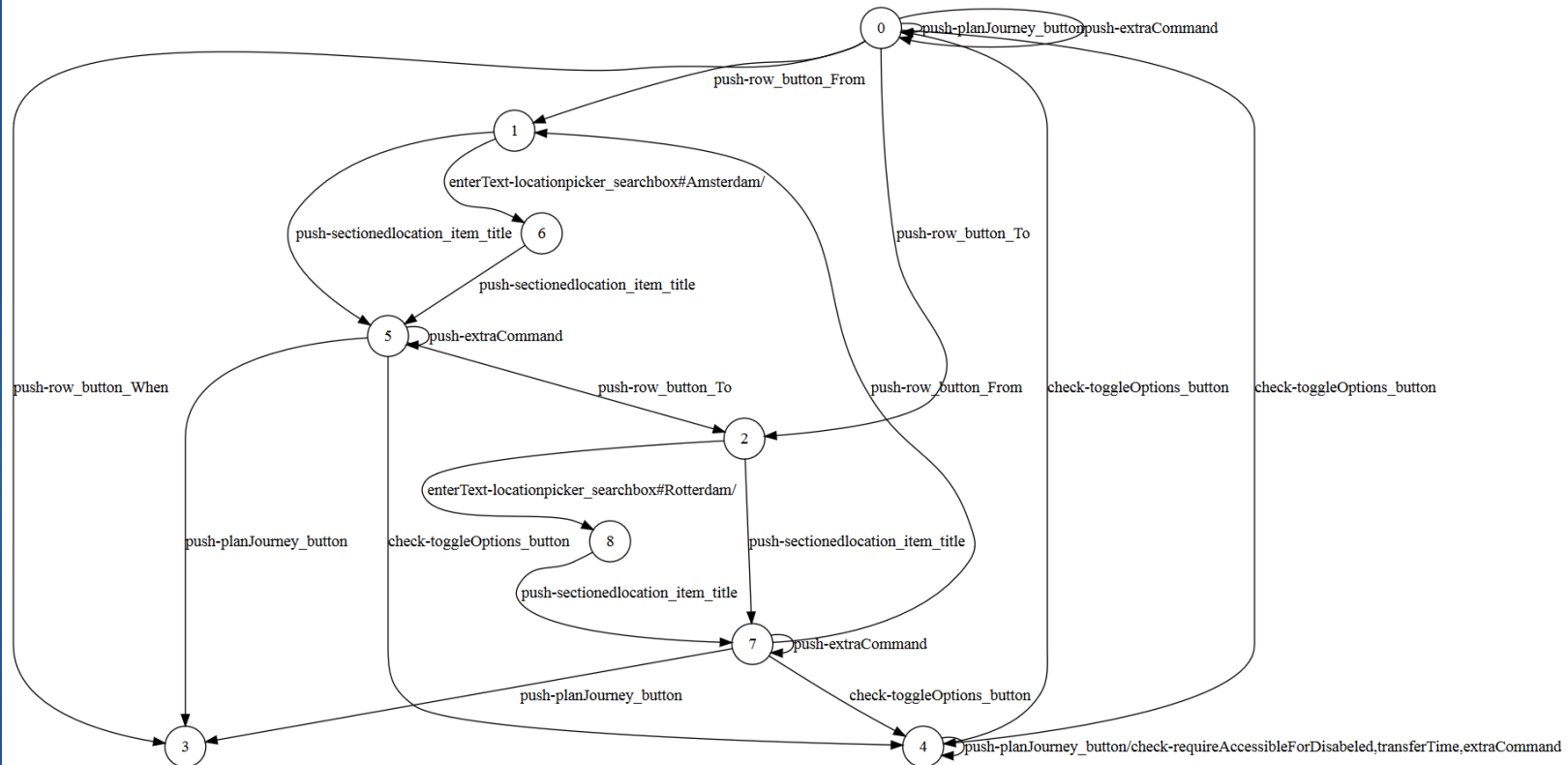
Video

13 / 20



9292 Model

14 / 20



Vulnerability Detection

15 / 20

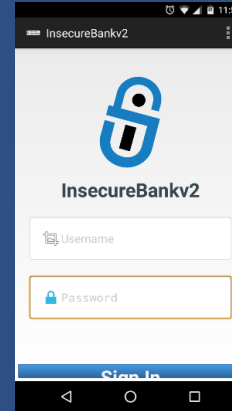
- ▶ Enrich Models
- ▶ Define Vulnerability Algorithms
 - ▶ OWASP Top10 Mobile
 - ▶ i.e. Insecure Authentication:

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

Results

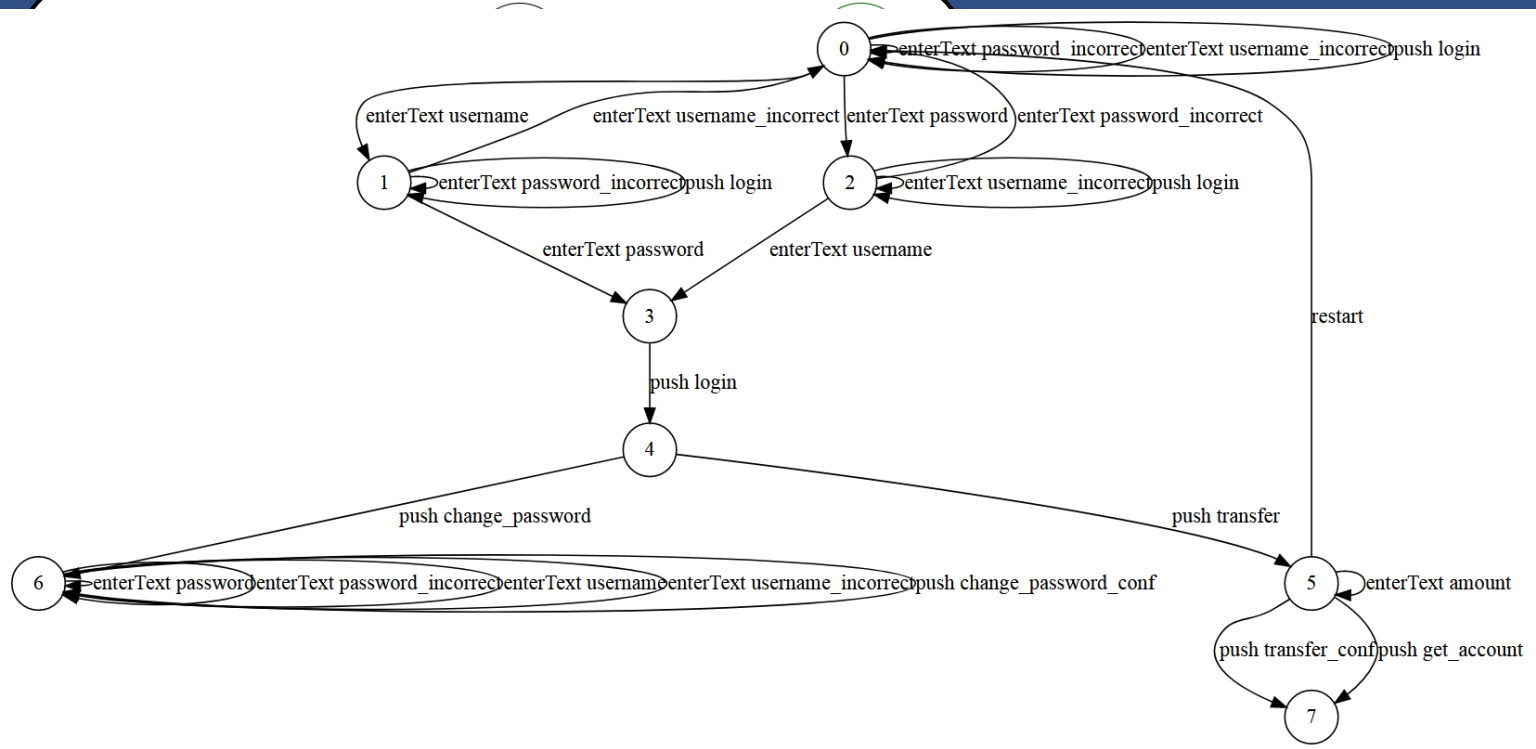
16 / 20

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



InsecureBankV2

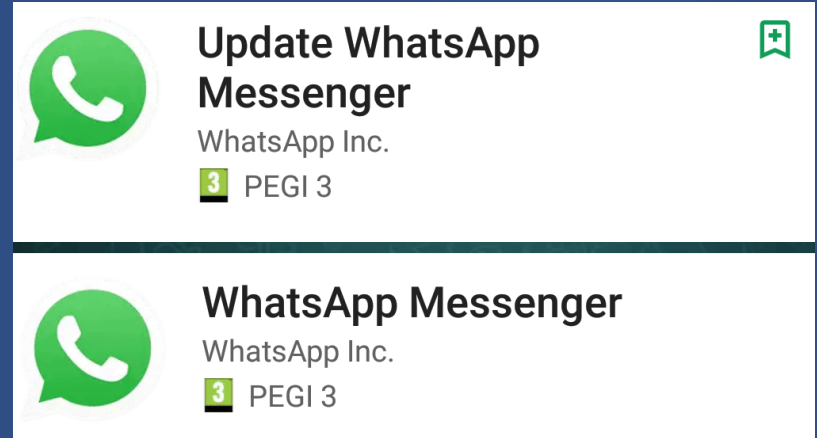
17 / 20



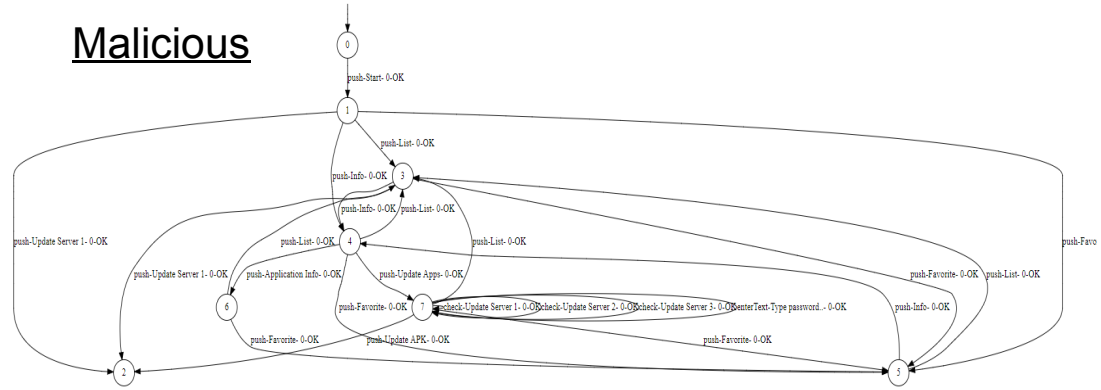
Fake WhatsApp

18 / 20

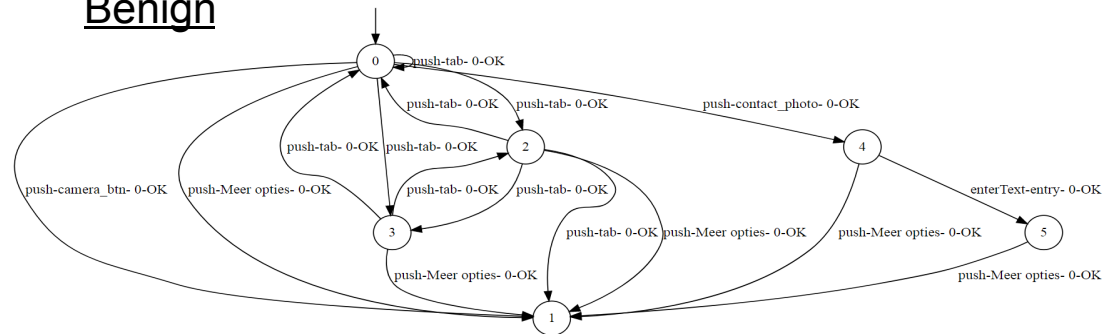
Background: Unicode trick allows malicious developers to impersonate WhatsApp developers
> 1M downloads, now deleted from Play Store



Malicious



Benign



■ Extraneous Functionality

- SUT > REF: 100%
- REF > SUT: 100%

■ Insecure Communication

```
GET http://req.startappserv
ice.com/1.4/..
```

Run statistics

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

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 (<i>hh : mm</i>)

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 (<i>hh : mm</i>)	27:37 (<i>hh : mm</i>)

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

Run statistics

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
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

20 / 20

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

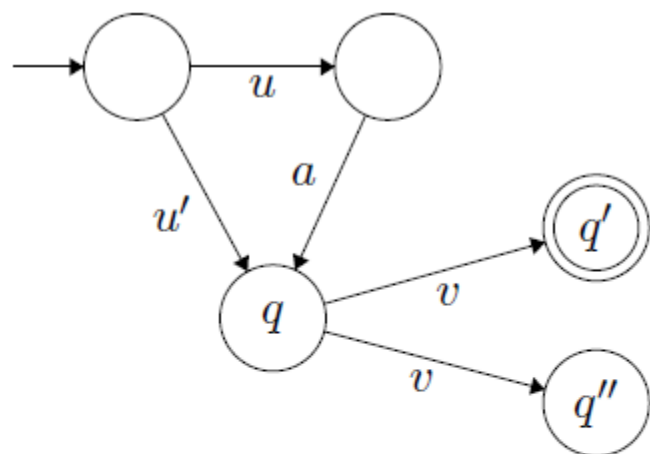




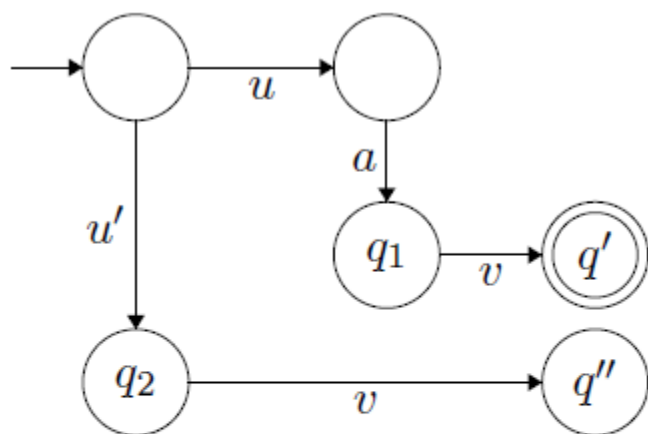
```

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        $\exists (s_1, s_2) \in S, a \in A, e \in E : row(s_1) = row(s_2) \text{ and } T(s_1 \cdot a \cdot e) \neq T(s_2 \cdot a \cdot e)$ 
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       $\exists s_1 \in S, a \in A : \forall row(s_1 \cdot a) \neq row(s)$ 
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 ▷  $R$  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:    $r \leftarrow \text{request\_insecure}(t.\text{request})$ 
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  $R$ 
```