# **Security Testing of Stateful Systems**

### **Erik Poll**

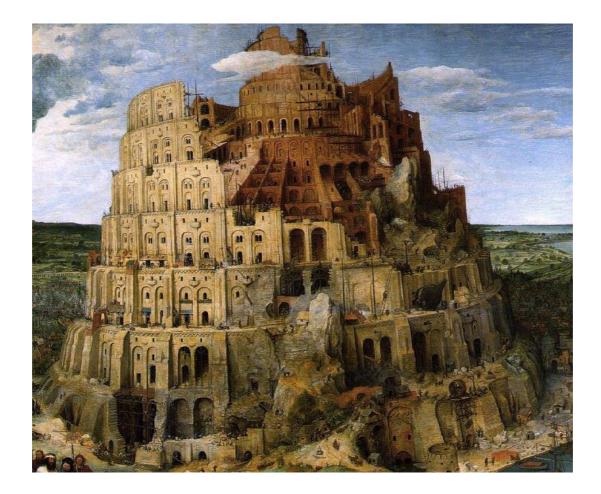
#### joint work with several PhD & MSc students

**Digital Security group** 

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## **Overview**

- 1. Wider context: LangSec
- 2. State machine learning as form of security testing
- 3. More general forms of fuzzing for stateful systems



# **1. LangSec** (Language-Theoretic Security)

## LangSec

Important root cause of security problems:

overly complex, expressive, poorly specified, ambiguous, input languages (aka formats, protocols,...)

Eg PDF, JPEG, Word, Bluetooth, TCP/IP, TLS, 5G, ....



#### Sergey Bratus & Meredith Patterson 'The science of insecurity' CCC 2012

http://www.youtube.com/watch?v=3kEfedtQVOY





## **Typical bug categories**

#### OWASP Top 10 [2017]

#### CWE TOP 25 [2022]

#### 1. Injection

- 2. Broken Authentication
- 3. Sensitive Data Exposure
- 4. XML External Entities (XXE)
- 5. Broken Access Control
- 6. Security Misconfiguration
- 7. Cross-Site Scripting (XSS)
- 8. Insecure Deserialization
- 9. Using Components with Known Vulnerabilities
- 10. Insufficient Logging & Monitoring

1 Out-of-bounds Write
2 Cross-site Scripting
3 SQL Injection
4 Improper Input Validation
5 Out-of-bounds Read
6 OS Command Injection
7 Use After Free
8 Path Traversal
9 Cross-Site Request Forgery (CSRF)
10 Unrestricted Upload of File with Dangerous Type
11 NULL Pointer Dereference
12 Deserialization of Untrusted Data
13 Integer Overflow or Wraparound
14 Improper Authentication
15 Use of Hard-coded Credentials
16 Missing Authorization
17 Command Injection
18 Missing Authentication for Critical Function
19 Improper Restriction of Bounds of Memory Buffer
20 Incorrect Default Permissions
21 Server-Side Request Forgery (SSRF)
22 Race Condition
23 Uncontrolled Resource Consumption
24 Improper Restriction of XML External Entity Reference
25 Code Injection

#### **CWE TOP 1000**

# ONE main bug category: INPUT handling

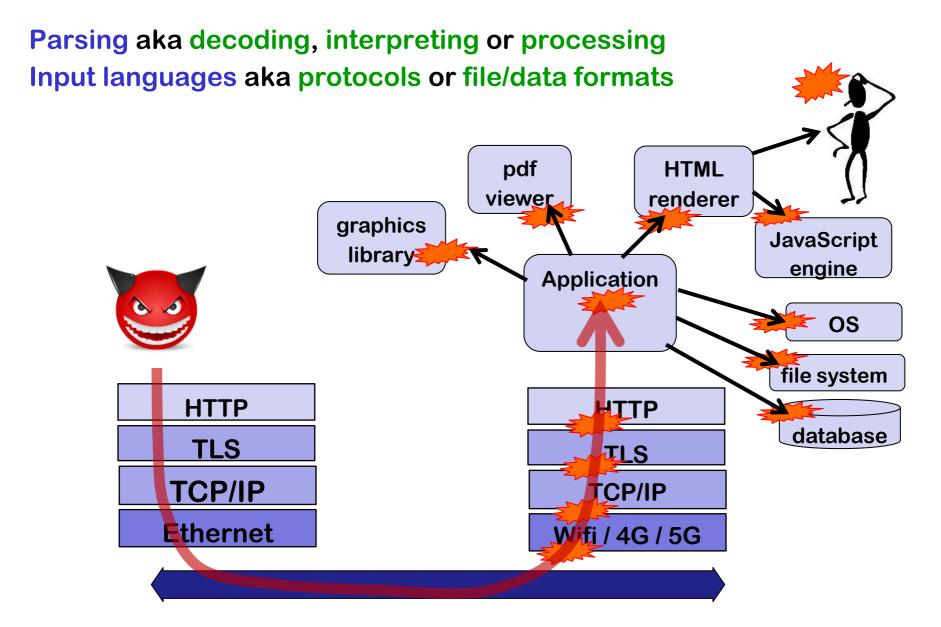
More in particular: **PARSING** of input is dangerous

Garbage In, Garbage Out

becomes Malicious Garbage In, Security Incident Out

or Garbage In, Evil Out

## Parsing of many languages in many places

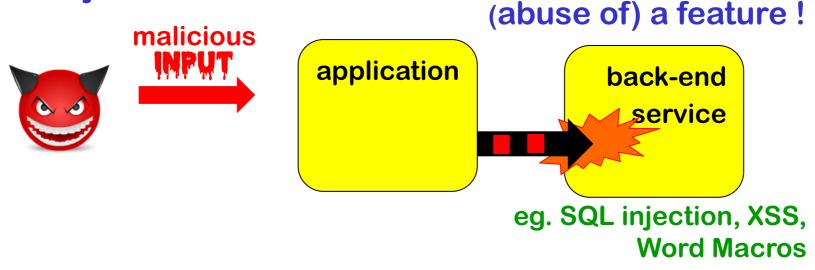


# Two sub-categories: 1) bugs & 2) features

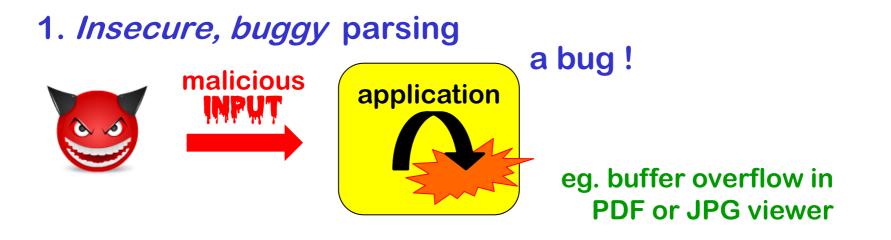
**1. Processing Flaws** 



#### 2. Injection Flaws



# Or 1) insecure parsing and 2) unintended parsing



#### 2. Unintended parsing malicious NPUT application back-end service eg. SQL injection, XSS, Word Macros

# LangSec: tackling buggy parsing

Adding input validation is not the (best) solution,

as we are only adding another parser

More structural 'LangSec' solutions to address root causes

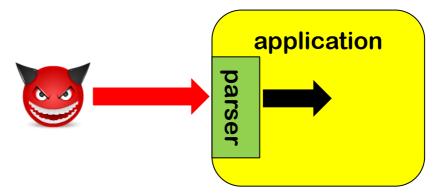
1. Provide clear, formal spec of input language

ideally as regular expression or (E)BNF grammar

2. Generate parser code

using a parser generator tool

3. Separate parsing (of raw bytes and strings into some structured datatype) from subsequent processing (of that structured data)



See langsec.org

# **Tackling unintended parsing**

- Using more specific data types instead of STRINCS
  - different types to distinguish different formats
    - eg URL vs file name
  - to distinguish different trust levels
    - eg compile-time constants and escaped input
       vs raw user input

Exemplified by Google's Trusted Type API



[Erik Poll, *Strings considered harmful*, USENIX :login; 2019]

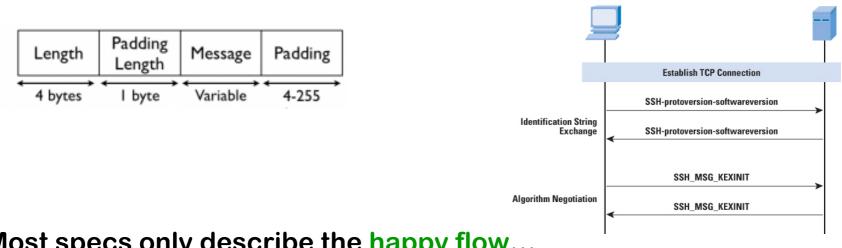
# 2. State machine learning for security

## Sessions, i.e. sequences of inputs

Client

Many protocols not only involves a language of input messages

but also a notion of session, ie. *sequence* of messages



- Most specs only describe the happy flow... ۲
- For security, getting unhappy flows correct is crucial ٠
- Fortunately, we can extract state machines from ۲ implementations using state machine inference aka active learning using just black box testing

Server

## Active Learning aka State Machine Inference

Just try out many sequences of inputs, and observe outputs

Eg. suppose input A results in output X ( A X )

- If second input A results in *different* output Y
- If second input A results in the *same* output X

Now try more sequences of inputs with A, B, C, ...

to e.g. infer

The inferred Mealy machine is an under-approximation of real system

L\* algorithm [Angluin 1987], implemented (in improved form) in e.g. LearnLib

A/Y

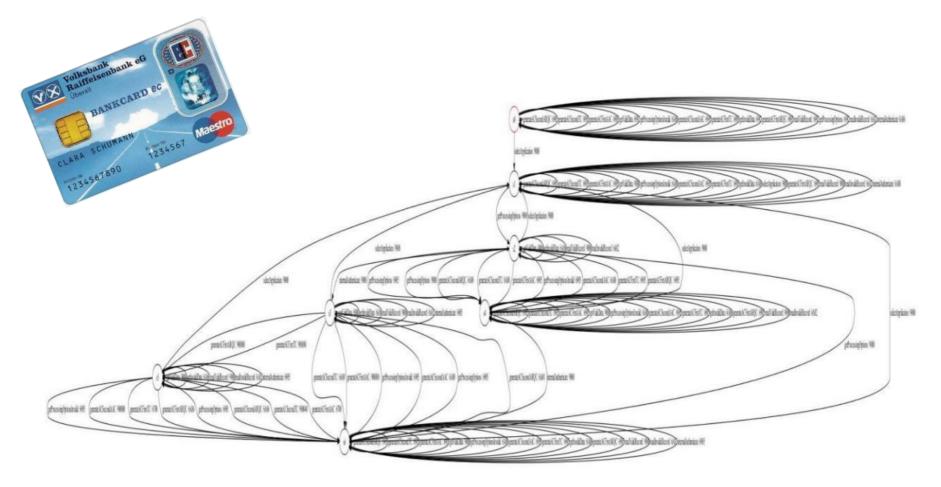
## Case study: EMV

- Most banking smartcards implement a variant of EMV
  - EMV = Europay-Mastercard-Visa
- Specification in 4 books totalling > 700 pages
- Contactless payments: another 7 books with > 2000 pages

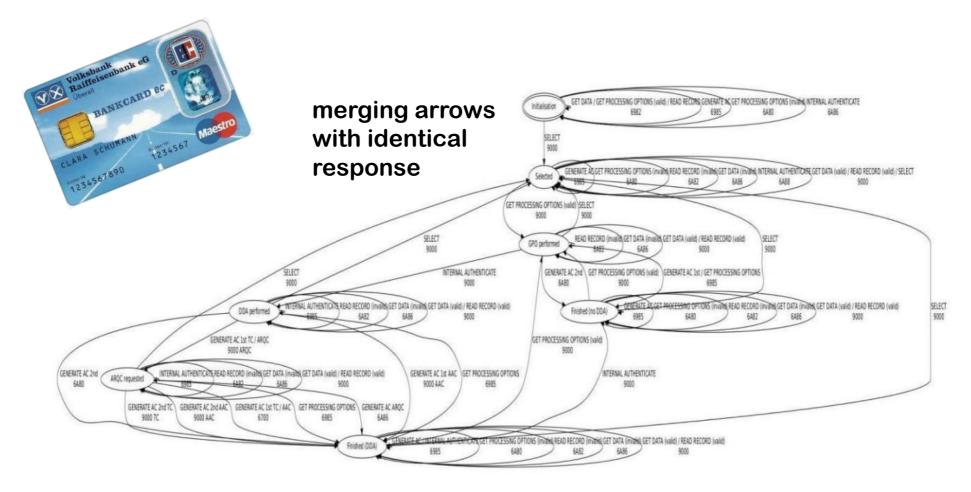




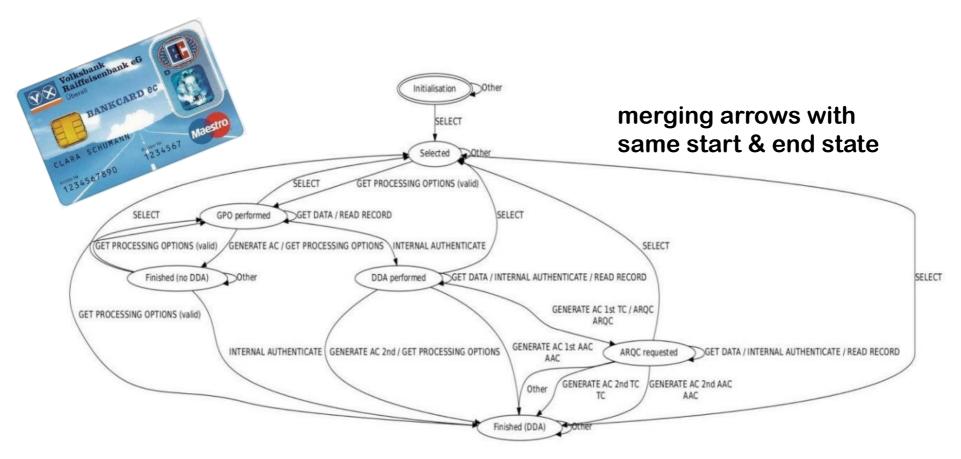








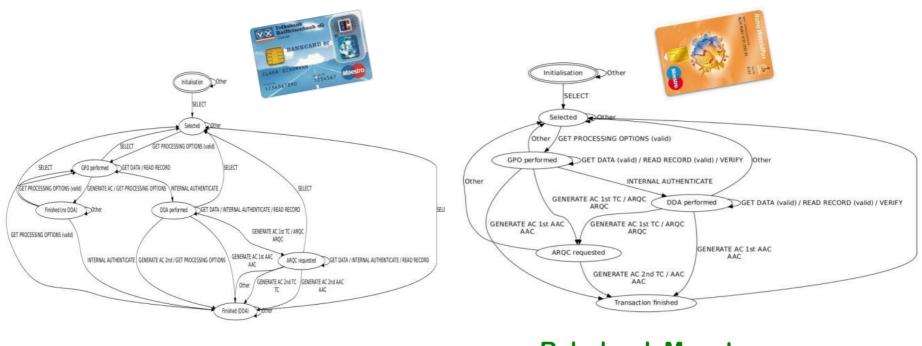




We found no bugs, but lots of variety between cards.

[Fides Aarts et al., Formal models of bank cards for free, SECTEST 2013]

## Using state machines for comparison

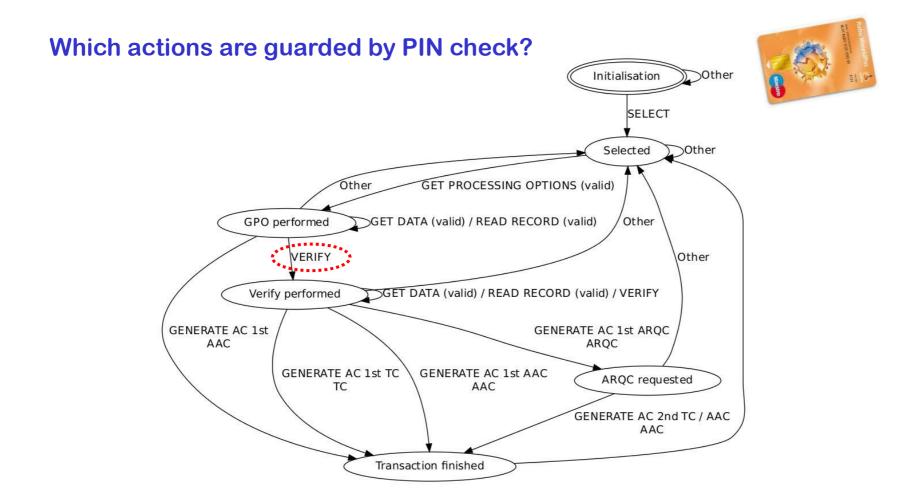


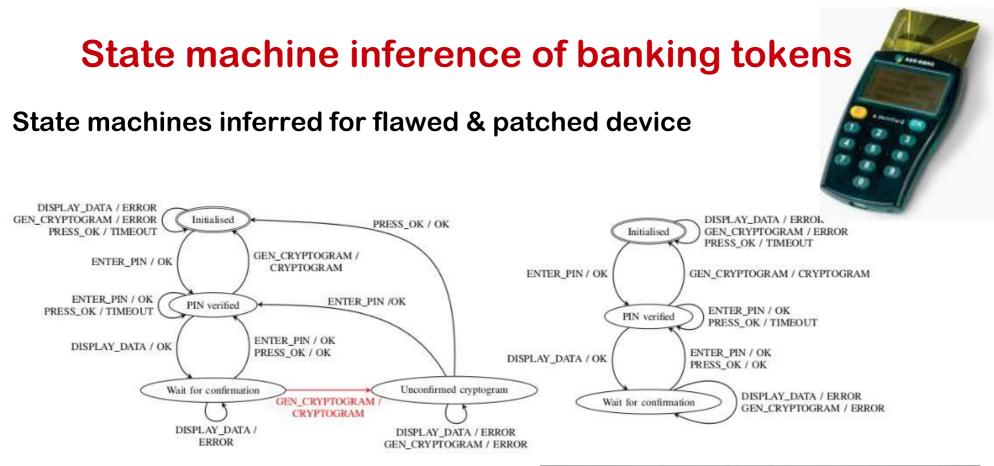
Volksbank Maestro implementation

Rabobank Maestro implementation

Are both implementations correct & secure? Or compatible?

## Using state machine for security analysis

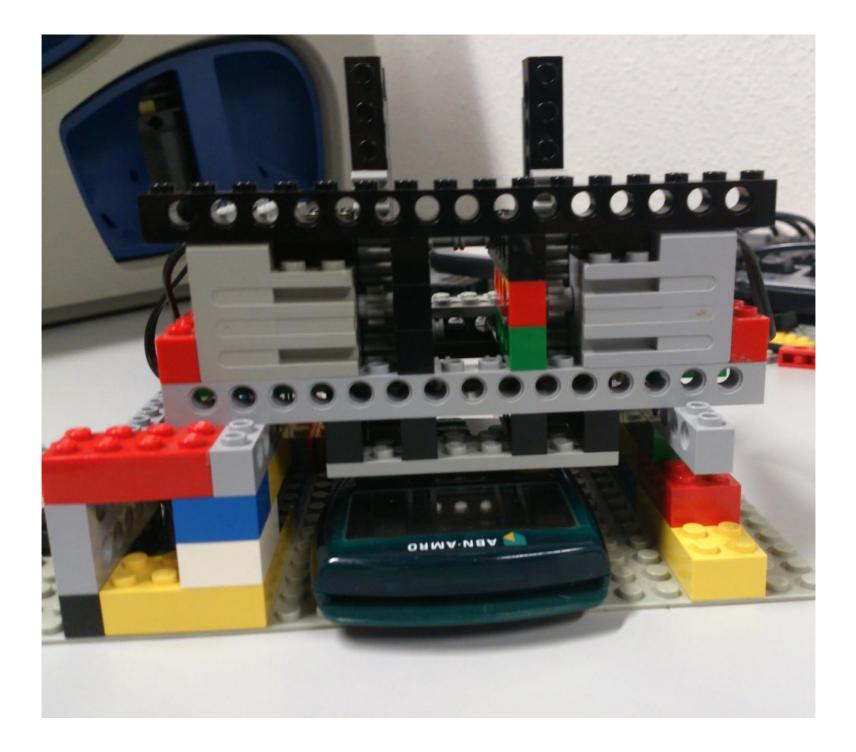


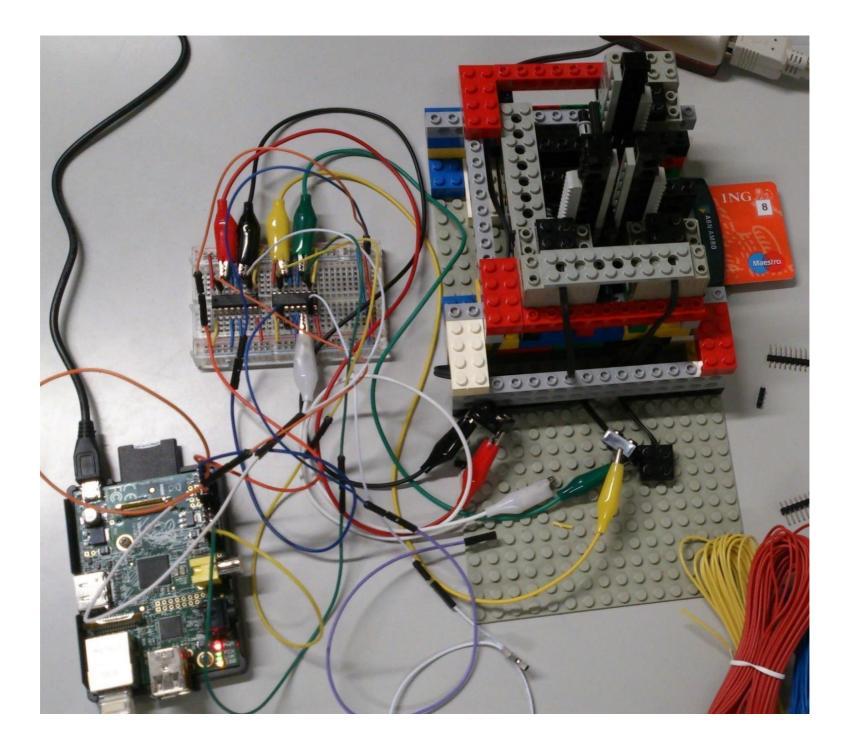


[Georg Chalupar, Stefan Peherstorfer, Erik Poll, Joeri de Ruiter, *Automated reverse engineering using Lego,* WOOT 2014]

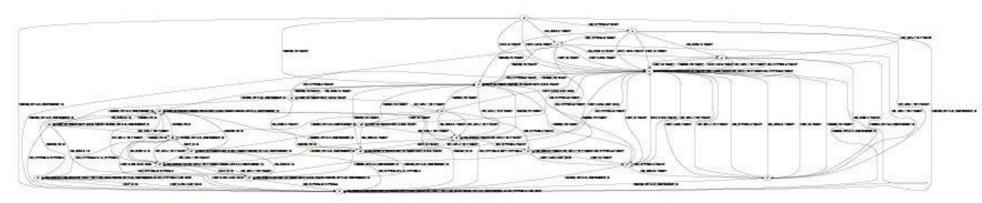
Movie at http://tinyurl/legolearn







#### State machine of Gemalto internet banking device



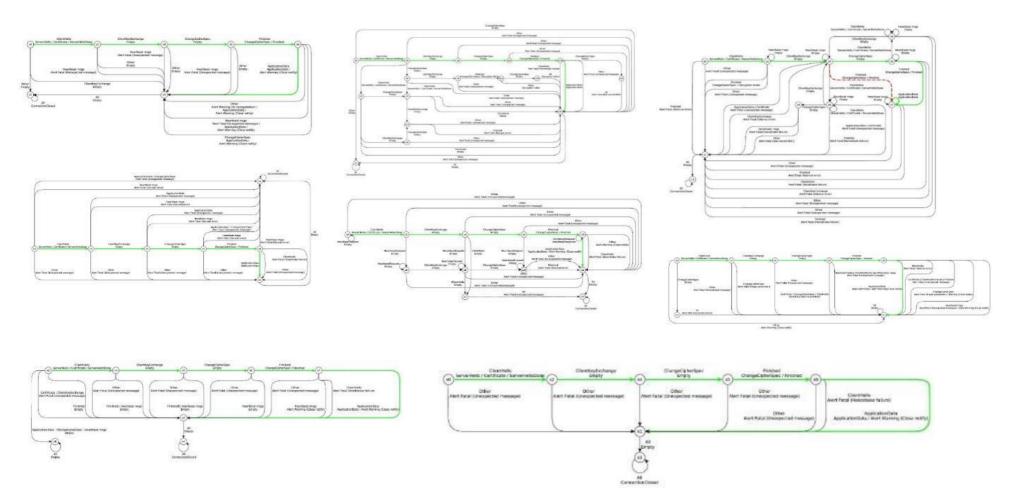
Complete inferred state machine

#### Would you trust this to be secure?

Did designers of Gemalto SWYS (Sign What You See) really intend all this complexity?



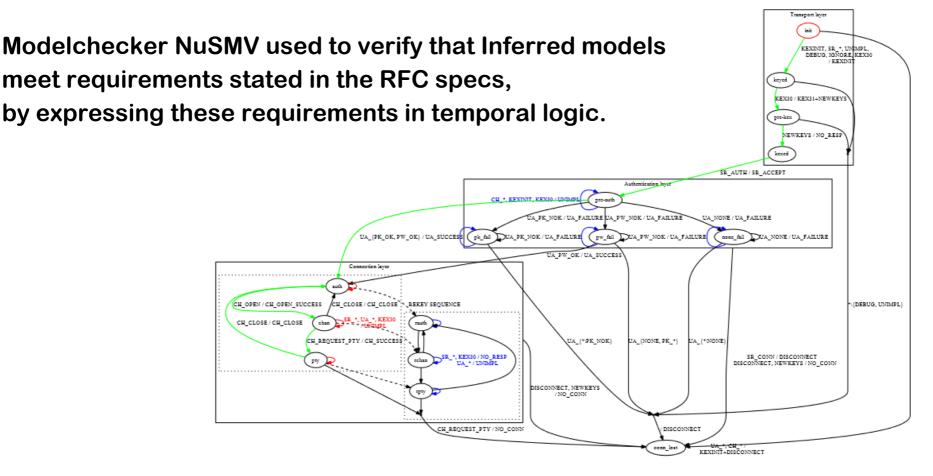
## State machine inference of TLS implementation



#### All implementations we analysed are different! Why doesn't the TLS spec include a state machine?

[Joeri de Ruiter and Erik Poll, *Protocol state fuzzing of TLS implementations*, Usenix Security 2015]

# State machine inference of SSH implementations



[Paul Fiterau-Brostean, Toon Lenaerts, Erik Poll, Joeri de Ruiter, Frits Vaandrager and Patrick Verleg, *Model Learning and Model Checking of SSH Implementations*, SPIN 2017]

## Typical prose specifications: SSH ⊗

Quote from the RFCs defining SSH:

"Once a party has sent a SSH\_MSG\_KEXINIT message for key exchange or re-exchange, until it has sent a SSH\_MSG\_NEWKEYS message, it MUST NOT send any messages other than:

- Transport layer generic messages (1 to 19) (but SSH\_MSG\_ SERVICE\_REQUEST and SSH\_MSG\_SERVICE\_ACCEPT MUST NOT be sent);
- Algorithm negotiation messages (20 to 29) (but further SSH\_MSG KEXINIT messages MUST NOT be sent);
- Specific key exchange method messages (30 to 49).

The provisions of Section 11 apply to unrecognised messages"

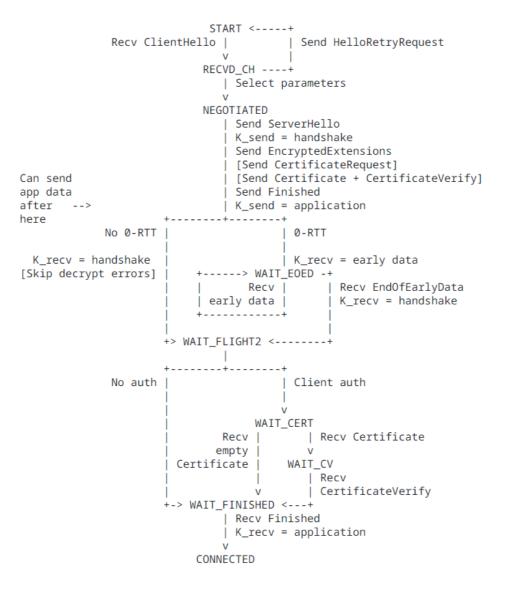
#### Understanding state machines from prose is hard!

This is another instance of the issues raised by LangSec: poor, informal specification of input formats

[Erik Poll, Joeri de Ruiter and Aleksy Schubert, *Protocol state machines and session languages*, LangSec 2015]



## There is some progress: TLS 1.3 [RFC 8446]



# 3. Fuzzing of stateful systems (work in progress)

## **Stateful fuzzing**

• State machine inference is a limited form of fuzzing:

it does not use strange/malformed input messages, but only strange *sequences* of *normal* input messages

- Given the success of fuzzing, how can we combine this with fuzzing for fuzzing stateful systems?
  - Note: fuzzing is successful because of the problems signalled by LangSec
- Not only for cryptographic protocols!
  - In fact, fuzzing cryptographic protocols is hard, as it requires a custom test harness that is labour-intensive to make

#### Survey of fuzzers for stateful systems [ArXiv:2301.02490]

There are many fuzzers around, but not that many for stateful systems

7 categories of stateful fuzzers

- Using different combinations of white box, grey box, black box
- Some infer a state machine, using active or passive learning

EVOLUTIONARY RAMMAR LEARNER EVOLUTIONARY RAMMAR RAMMAR CRAMMAR CRAMMAR CRAMMAR

**Open questions:** 

- What are the optimal combinations?
- Should implementations be made "fuzzer-friendly"?

[Cristian Daniele, Seyed Andarzian, Erik Poll, *Fuzzers for stateful systems: Survey and Research Directions*, ArXiv:2301.02490, 2023]

## Conclusions

- Most security flaws are INPUT processing flaws
- These flaws arise in **PARSING** of many input languages / formats
  1) buggy/insecure parsing or 2) unintended parsing
- LangSec identifies root causes and points to structural solutions
- Particular case of input format: *sequence* of messages in a protocol
- We can automatically extract state machine of such behaviour using state machine learning (aka active learning)
- State machine learning is limited form of fuzzing. *How can we best fuzz stateful systems?*

Should protocols be implemented in a fuzzer-friendly way?

*Also use passive learning using FlexFringe?* [Verwer & Hammerschidt, 2022 arXiv2203.16331]