

SQL Injections and Reinforcement Learning:

An Empirical Evaluation of the Role of Action
Structure

Nordsec 2021



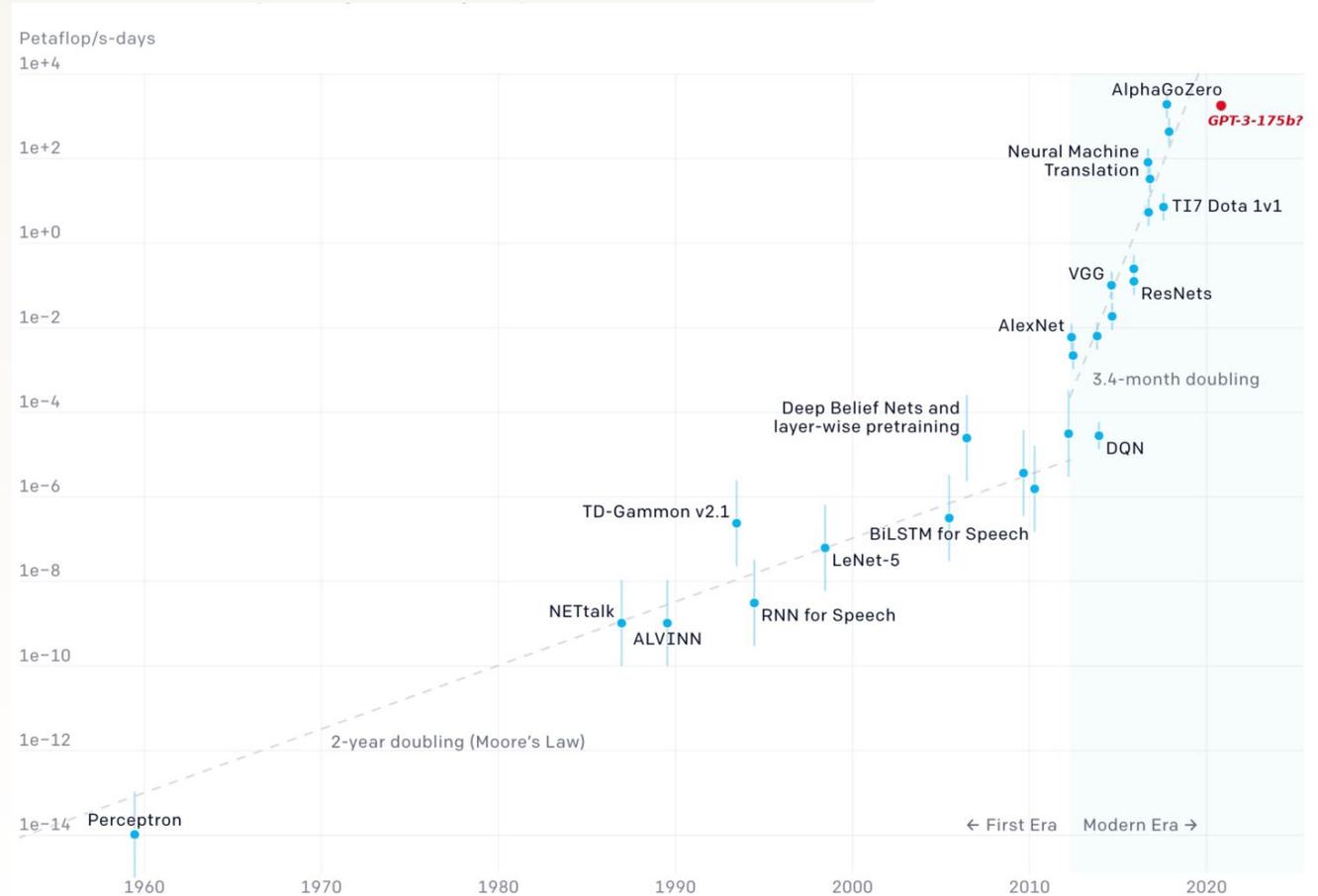
UiO : **University of Oslo**



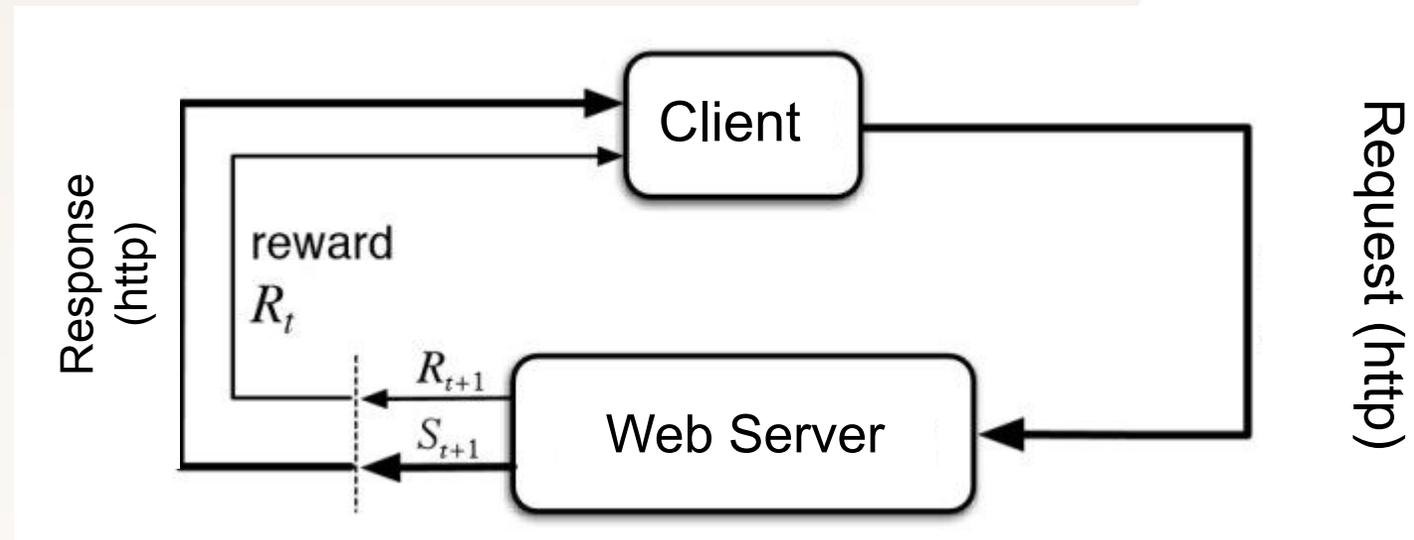
Mila

Deep Learning

- Computational power is cheap.
- The exponential growth is holding.
- **GPT3** 175 billion parameters.
- **OpenAI Codex** Code generation.
- **AlphaGo** 1,920 CPUs 280 GPUs.



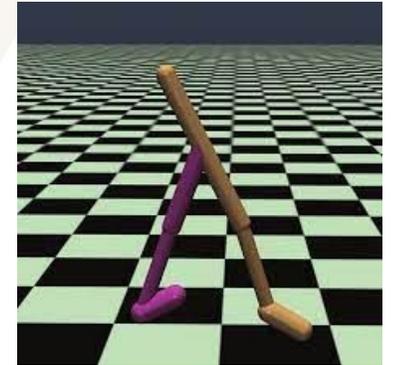
Reinforcement Learning for Web Fuzzing



- **Action space**
 - How does the agent communicate to the environment
 - GET /<payload> HTTP/1.1
 - payload can be: hardcoded, composed of a vocabulary, characters, bits.
- **Observation space**
 - How does the environment communicate with the agent
 - HTTP status code, HTML

Deep Reinforcement Learning

- Reinforcement Learning with Deep Neural Networks
- The classical problem settings:
 - Robotics, physics simulators (OpenAI gym MuJoCo-ant)
 - ~1k steps/second
 - ~10k steps/second with specialized hardware (2020, TPUs)
- Web Server:
 - Commercial servers:
 - 1.2 Million requests/sec per second ([in 2016](#))
 - 5k req/s, random web server (`ab -n 10000 -c 100 http://mila.quebec/`)
 - Deep RL agents require a lot of environment interactions (~1M for simple problems).
 - Simulation speed is crucial



The Setup

The Environment

- **2 SQL tables**
 - User, the working table.
 - Private, which holds the flag.
 - **Processed responses:**
 - Unsanitized user input field:
 - At every attempt we select one out of three possible queries
 - And a variable number of column.
 - Probing is required for both.
- Best possible solution requires at most 5 actions:
- 3 actions to guess the correct number of columns
 - 2 actions to find the escape

Table 1. Users table schema.

| ID | username | firstName | lastName | age | nationality | create_at |
|-----|----------|-----------|----------|-----|-------------|-----------|
| ... | ... | ... | ... | ... | ... | |

Table 2. Private table schema.

| ID | user | account |
|-----|------|---------|
| ... | ... | ... |

```
SELECT cols FROM Users WHERE firstName = "<input>"
SELECT cols FROM Users WHERE nationality = '<input>'
SELECT cols FROM Users WHERE age = <input>
```

The Environment

- **One Algorithm, Multiple tasks**
 - Reconnaissance and exploitation are interleaved

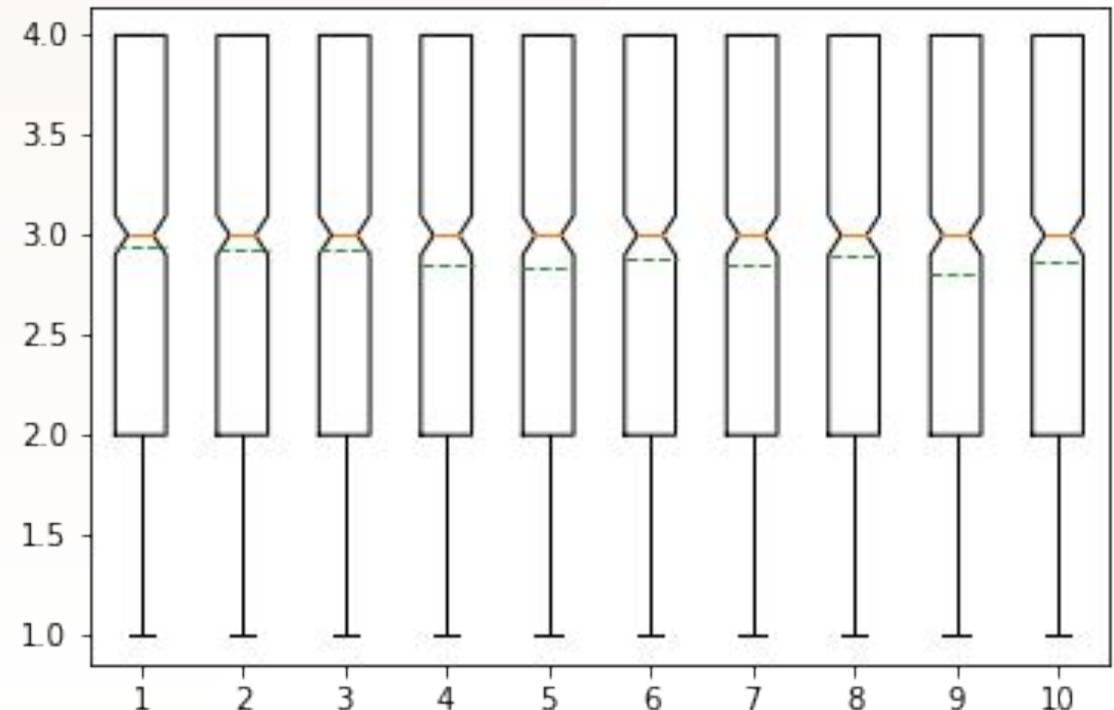
```
SELECT cols FROM Users WHERE firstName = "<input>"  
SELECT cols FROM Users WHERE nationality = '<input>'  
SELECT cols FROM Users WHERE age = <input>
```

Structured agent

- **Hard coded requests (25 actions)**
 - 6 escape probing (e.g. 1' or 1=1 --)
 - 9 exploratory (e.g. 1" union select NULL --)
 - 9 exploit attempts (e.g. 1" union select account, NULL from private --)
 - 1 other "\0"
 - **Processed responses:**
 - A list of 25x1 numbers, one for each possible action.
 - 0 at an index means that the action has been tried and an sql error was returned
 - 1 action never tried
 - 2 action tried -> no data returned
 - 3 action tried -> something
 - 4 flag found
- action 1: -> obs [2, 1, 1, 1, 1, 1, 1, ...] # valid sql, no data
- action 3: -> obs [2, 1, 3, 1, 1, 1, 1, ...] # valid sql, data was returned

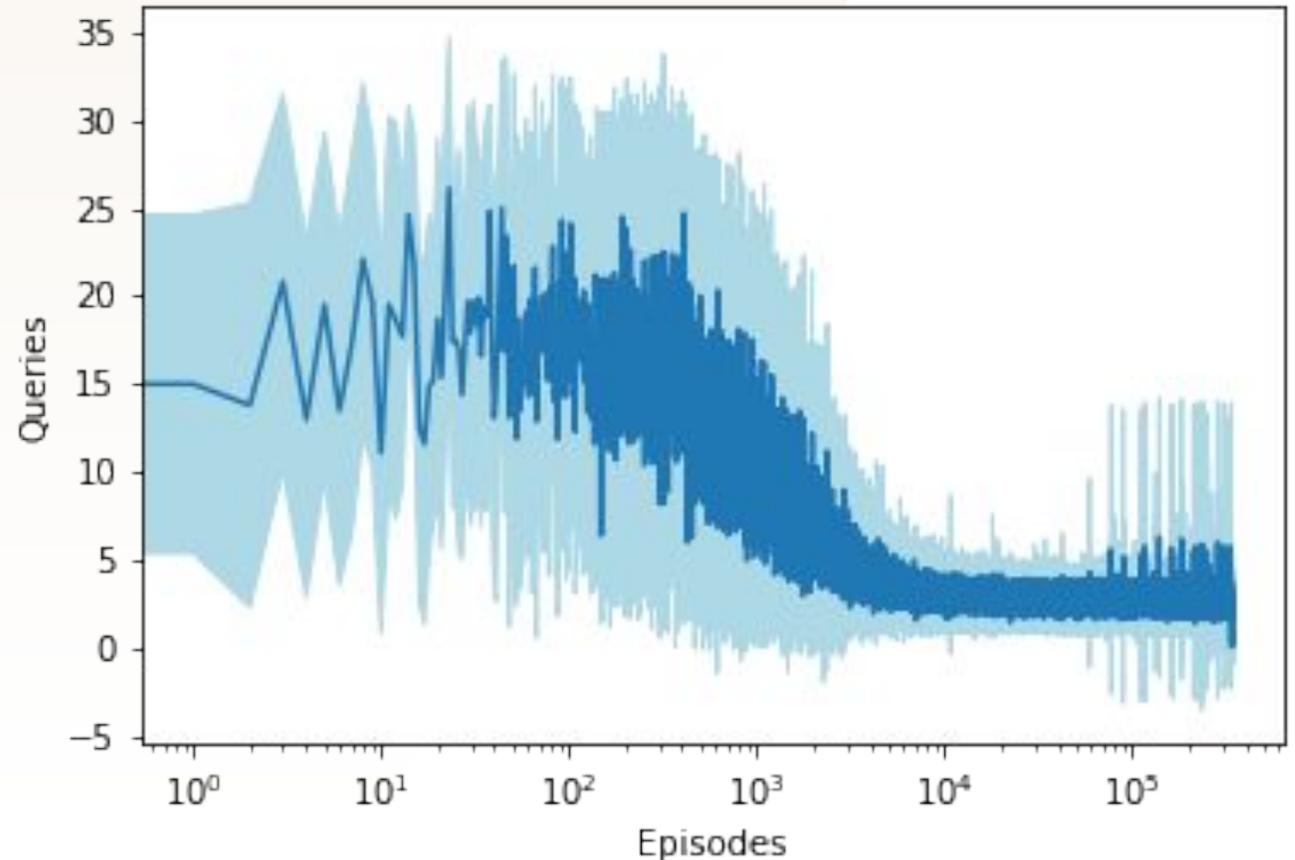
Structured agent, Final Performance

- Performance of 10 different agents:
- Orange line gives the median number of queries (~3), green mean
- 25%, 50%, 95% confidence interval for the median



Structured agent, Training

- Number of queries to capture the flag
- ~10k steps to convergence
- 3 queries per attempt
- w.r.t. 17 queries by trying randomly



Structured agent

- **Pros:**
 - Easy to encode human knowledge.
 - Higher efficiency to solve tasks.
- **Cons:**
 - As good as the encoded knowledge.
 - Unlikely to find surprising results.

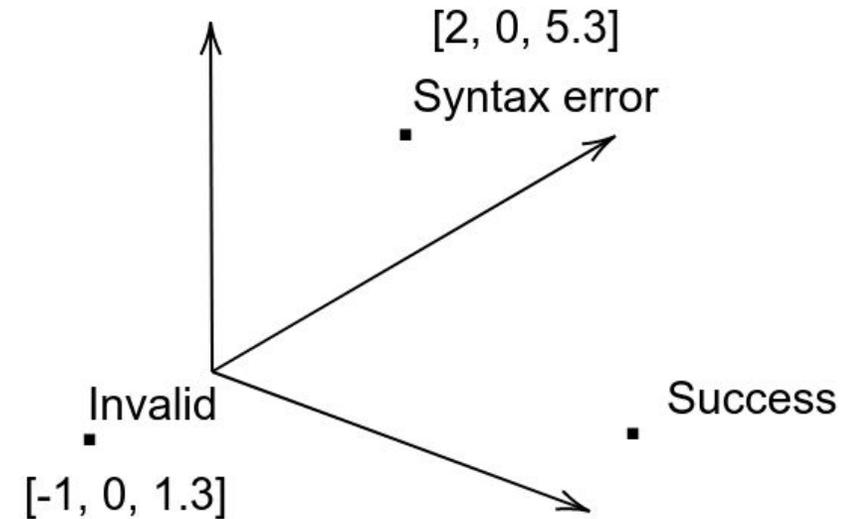
Unstructured Agent

- **Huge action space >100k distinct actions**

- Actions are selected from a dictionary of:
 - SQL keywords
 - A “wordlist”
- An action is a mix of the dictionary words
- $a = [34, 8, 9, 12, 0]$

- **Observations**

- A transformation of the SQLite error message
- e.g. “SELECTs to the left and right of UNION do not have the same number of result columns” becomes “columns”
- any variation of “syntax error” becomes “Syntax error”
- Then the string then becomes a multidimensional embedding, $[-2, 0, 5.3]$



Training progress

1000 steps

1 NULL, 1 UNION SELECT a FROM p --

UNION SELECT a FROM p --

1 UNION SELECT UNION SELECT a FROM p --

NULL, UNION SELECT a FROM p -- NULL,

NULL, ' "

16000 steps

' UNION SELECT NULL, NULL, a FROM p --

" UNION SELECT NULL, NULL, a FROM p --

' UNION SELECT NULL, a FROM p --

UNION SELECT " UNION SELECT NULL, a FROM p --

1 UNION SELECT NULL, a FROM p --

Structured agent

- **Pros:**

- Does not rely on human knowledge (and limitations).
- Scales with compute, not with man-hours.
- Can find surprising solutions.

- **Cons:**

- Complex algorithms to write.
- Requires Reinforcement Learning + Security + Natural Language Processing understanding.
- Young field, no turnkey solutions.
- Good solutions emerge after 1-2 Million interactions.

Why Reinforcement Learning now

- **We have the resources and algorithms:**
 - Web servers are fast.
 - These algorithms scale with compute speed.
 - Can find surprising solutions.
 - Can handle as many edge cases as there are.
- **Experts define the problem.**
 - What makes a successful solution.
 - How does the program solution look like (neural net architecture)
 - Compute finds the program.
- **Exponential scale gives exponential results**
 - Often is just a matter of **more compute**.
 - DeepMind, OpenAI showed we did not yet hit the limit of scale

Questions? Contacts

- Manuel Del Verme (manuel.delverme@gmail.com)
- Åvald Åslaugson Sommervoll (aavalds@ifi.uio.no)
- Laszlo Erdodi
- Simone Totaro
- Fabio Massimo Zennaro



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