

HoPLA: a Honeypot Platform to Lure Attackers

Elisa Chiapponi - EURECOM

Onur Catakoglu - Amadeus IT Group

Olivier Thonnard - Amadeus IT Group

Marc Dacier- EURECOM



Who are we?

Elisa Chiapponi

- Ph.D. student at EURECOM, collaboration with Amadeus IT Group
- Analysis and mitigation of the new generation of botnet

Dr. Onur Catakoglu

- Information Security Architect, GSO Amadeus IT Group

Dr. Olivier Thonnard

- Senior Security Expert, Tech Lead GSO Amadeus IT Group

Prof. Marc Dacier

- Head of Digital Security Department, EURECOM

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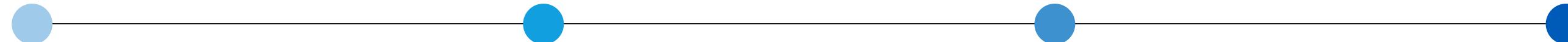
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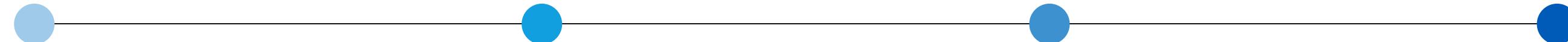
- Head of Digital Security Department, EURECOM

Our Agenda



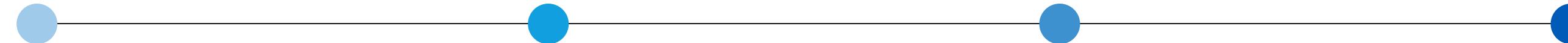
Our Agenda

1. Introduction and motivations



Our Agenda

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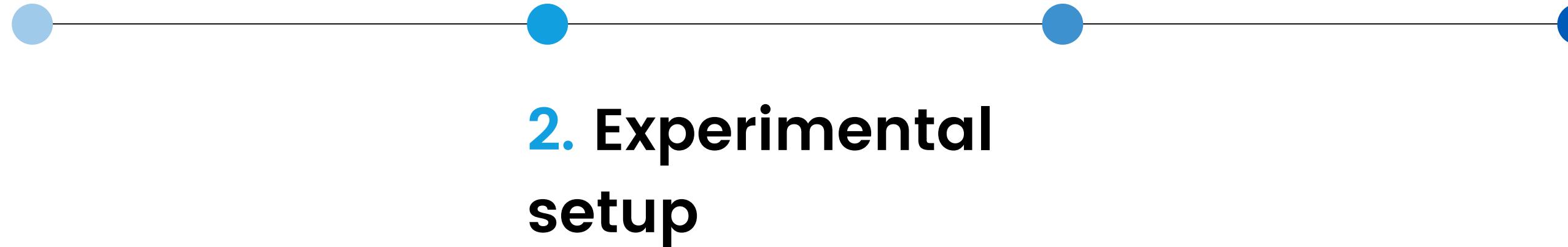


2. Experimental
setup

Our Agenda

**1. Introduction
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**3. Results and
discussion**



Our Agenda

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4. Conclusions
and future works

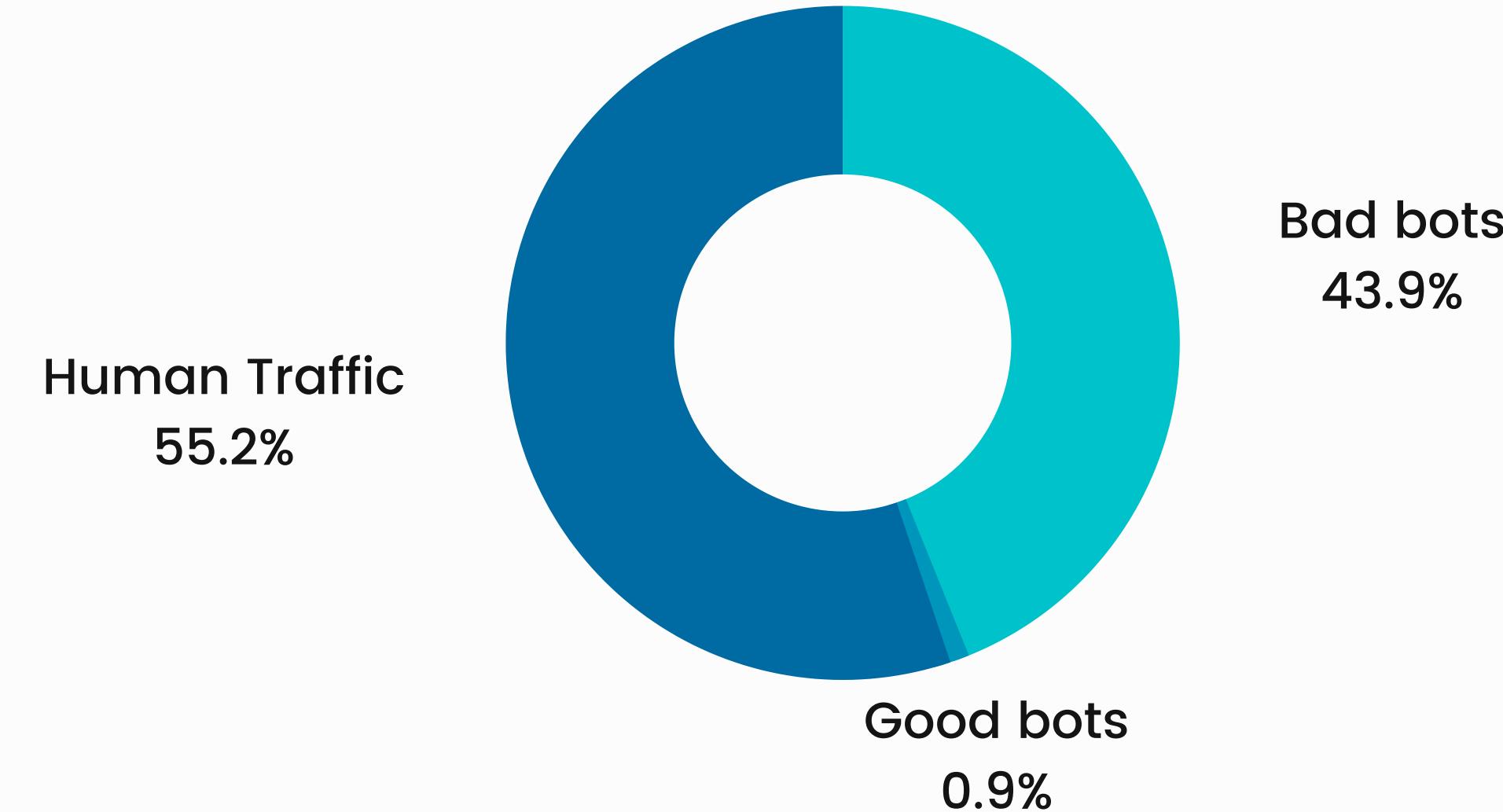
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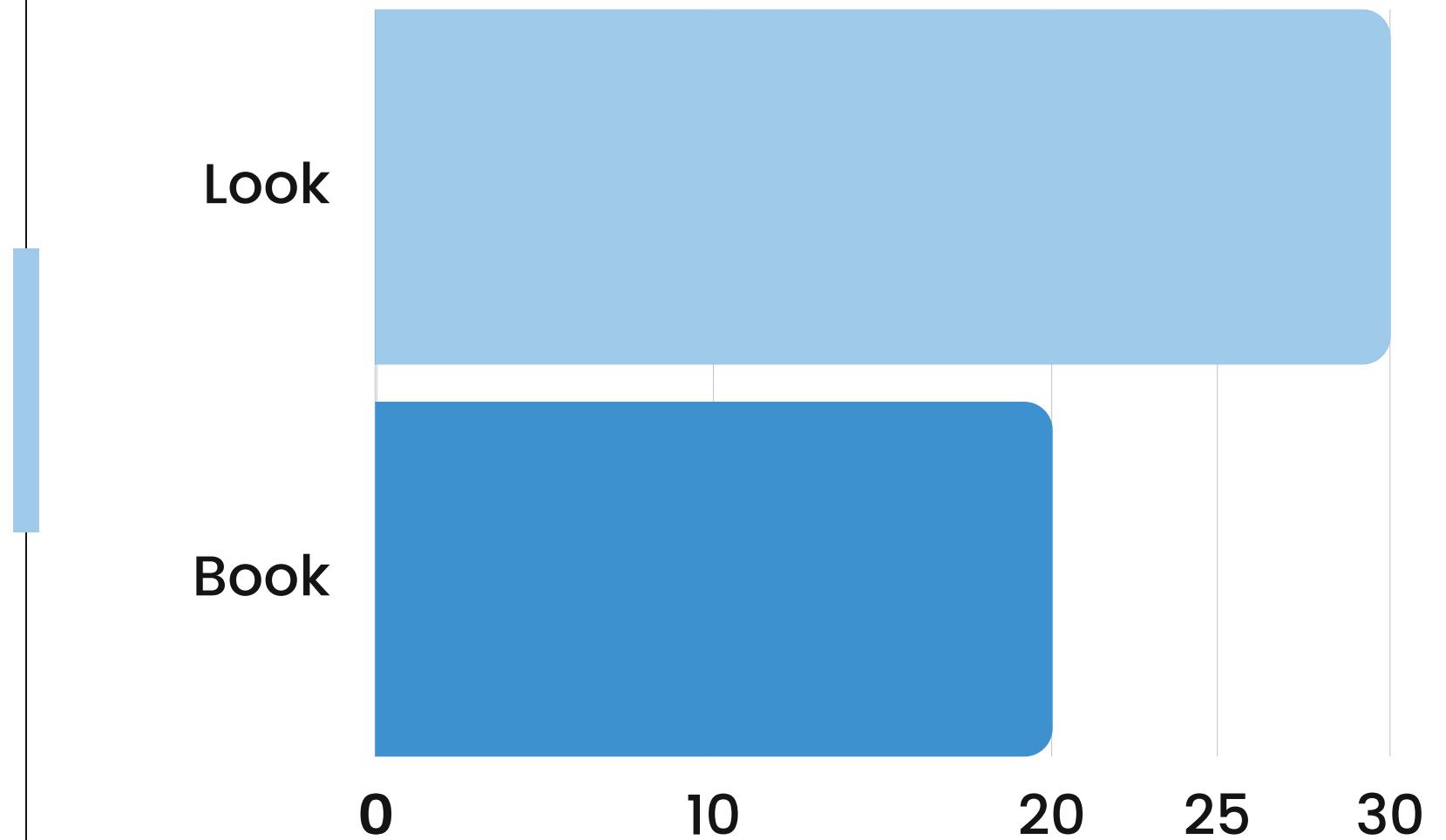
Web scraping

Web scraping is the periodical or continuous retrieval of accessible data and/or processed output contained in web pages.

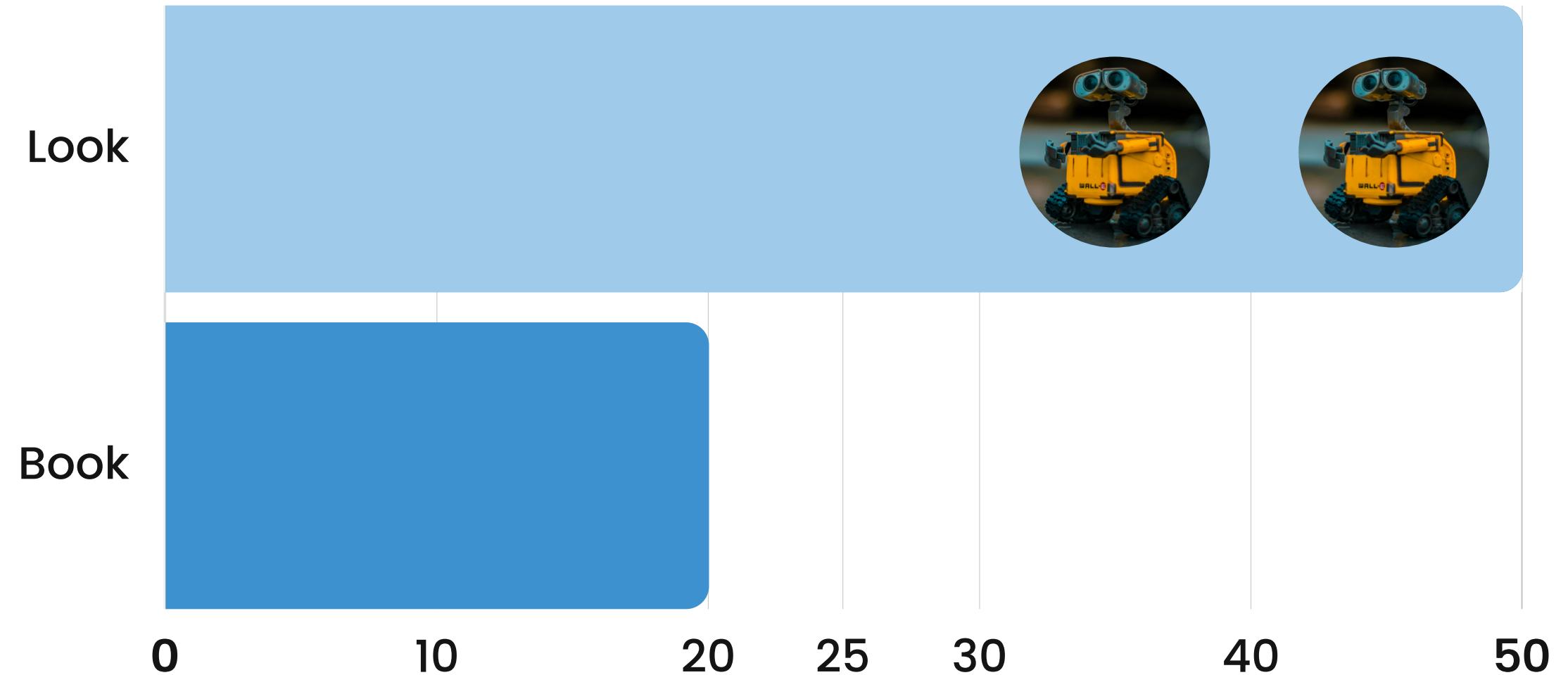
Scraping bots and airlines



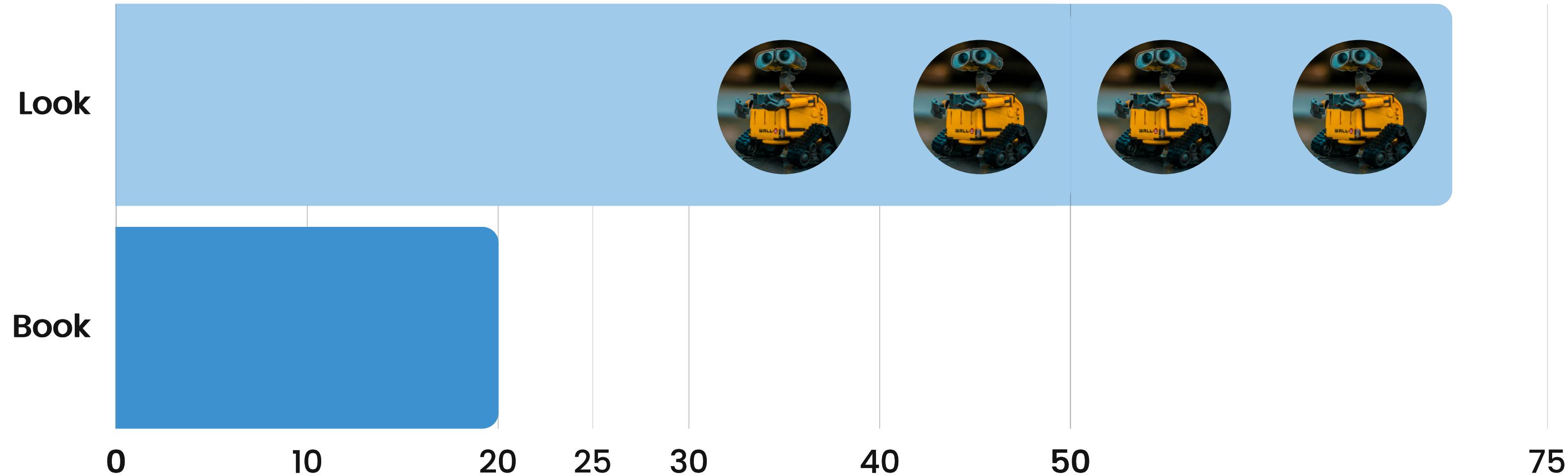
Look-to-book ratio



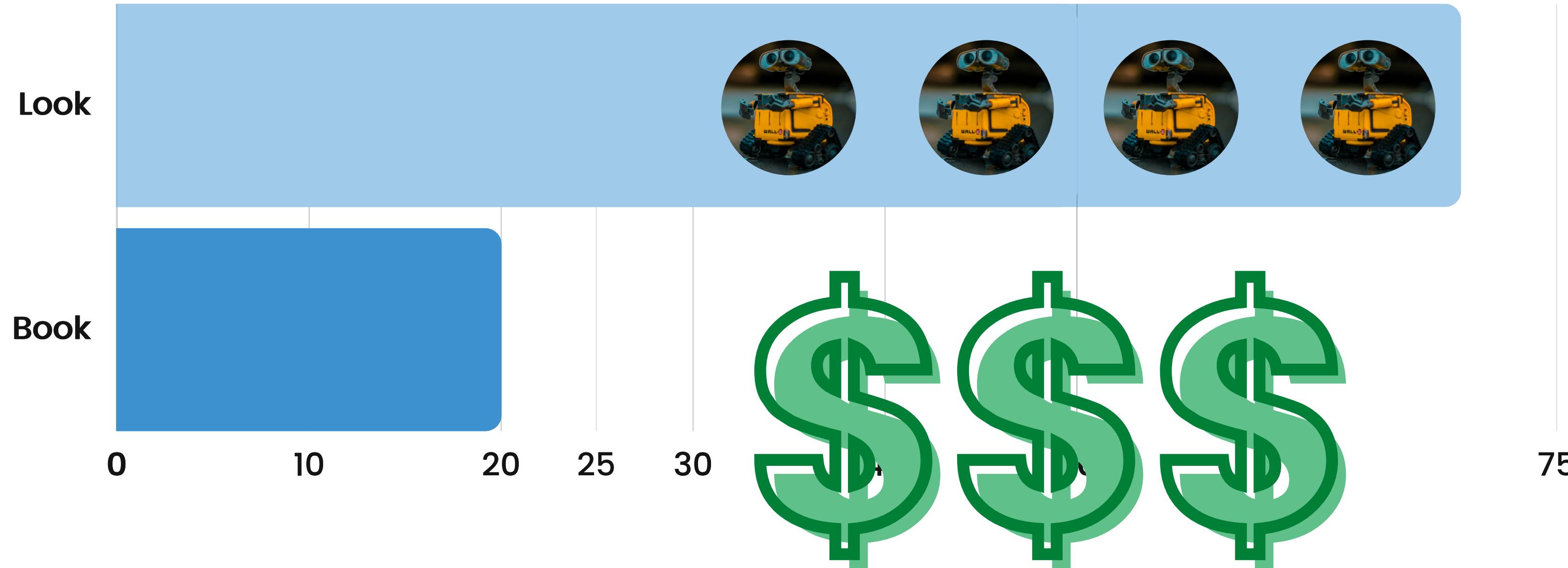
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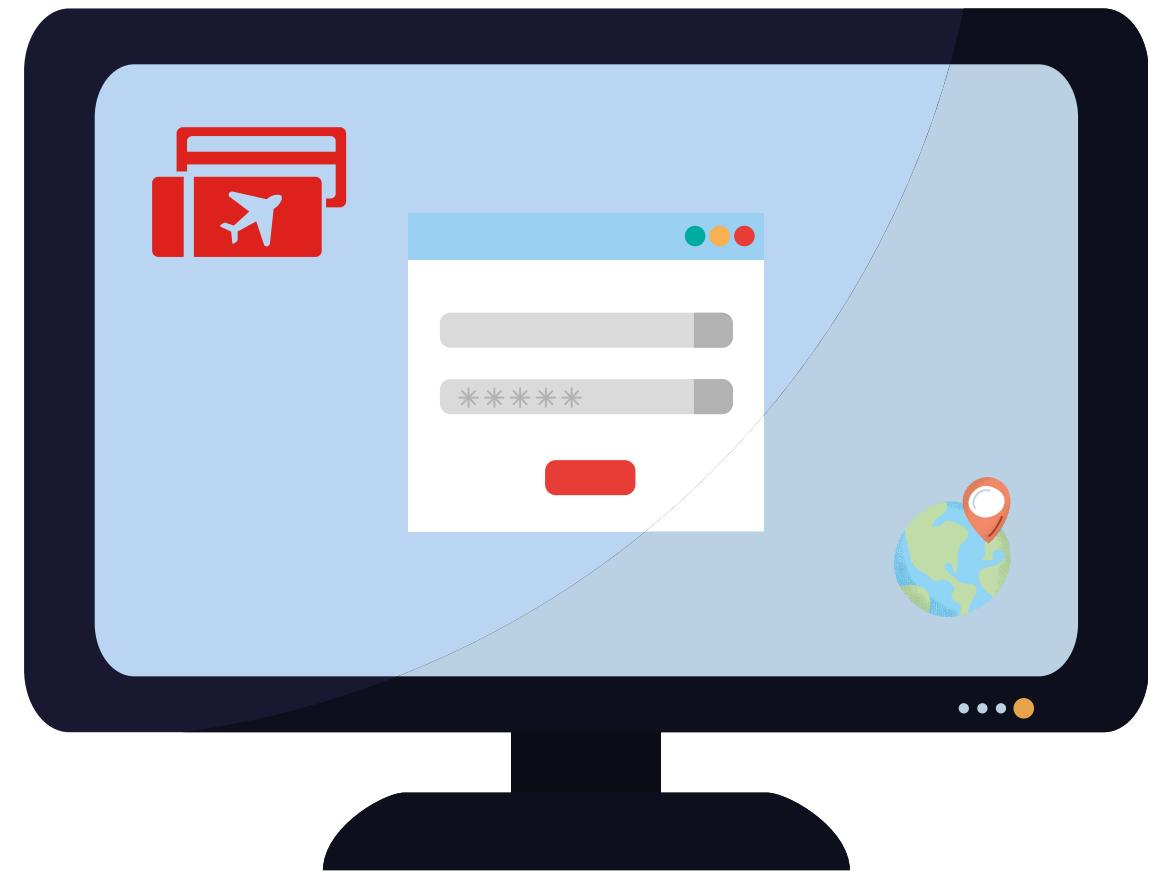
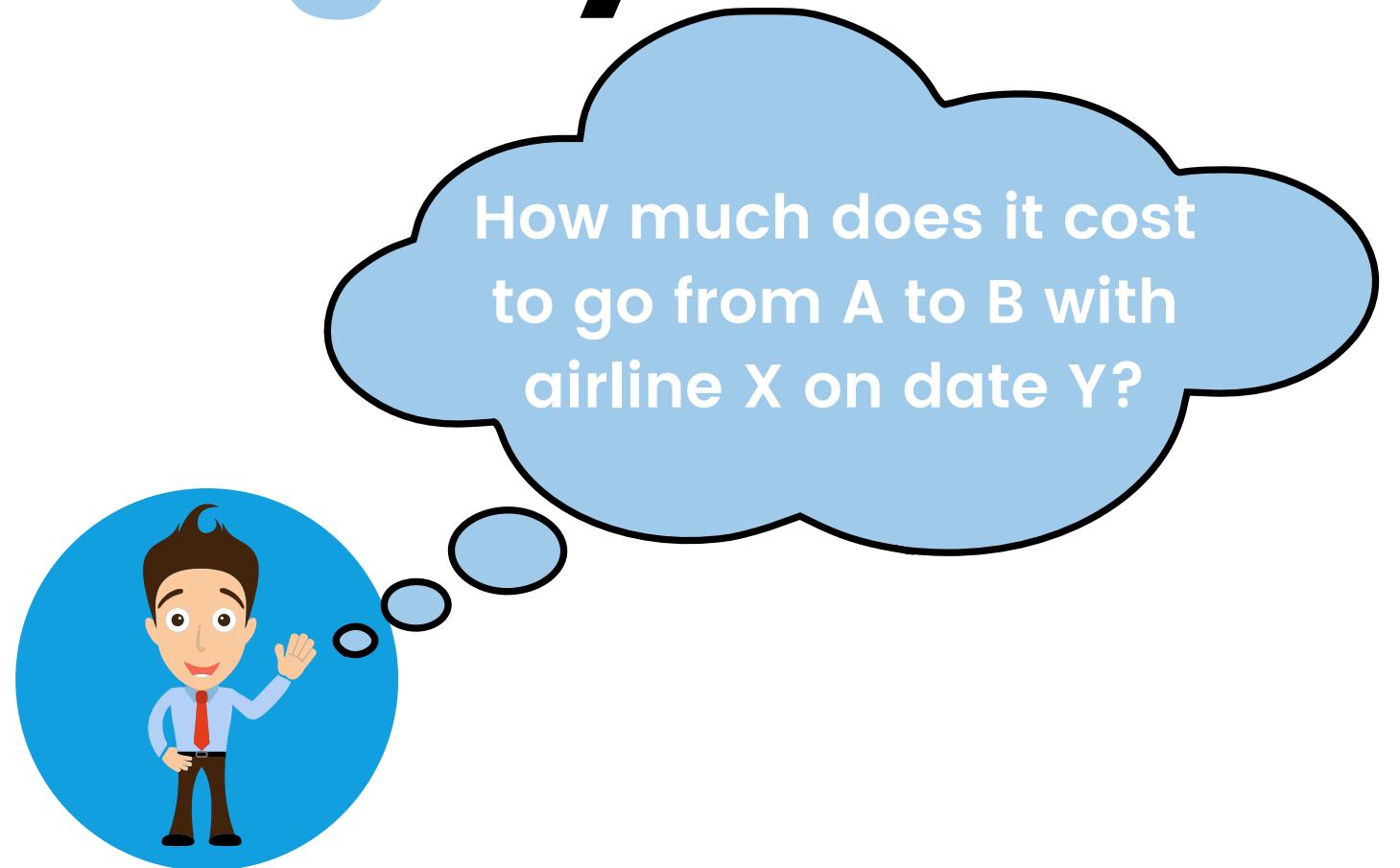
Booking systems



**Booking domain
of airline X**

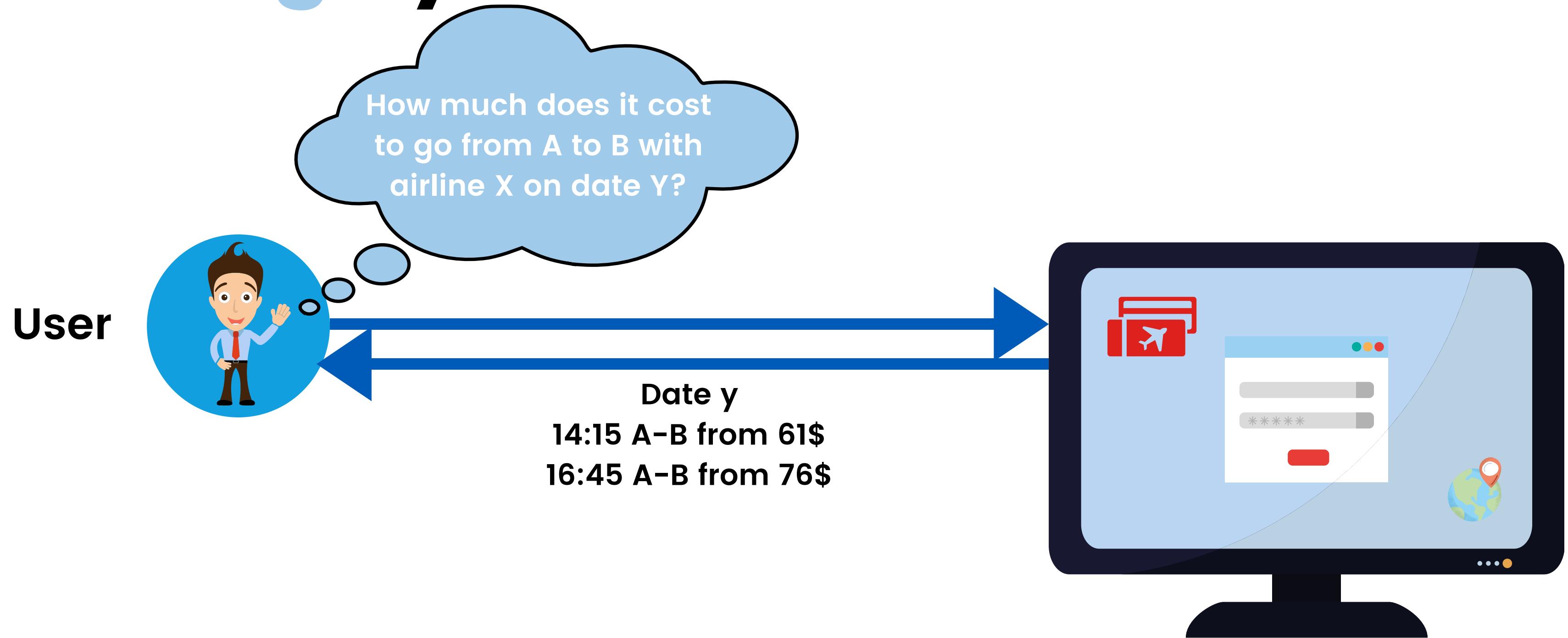
Booking systems

User

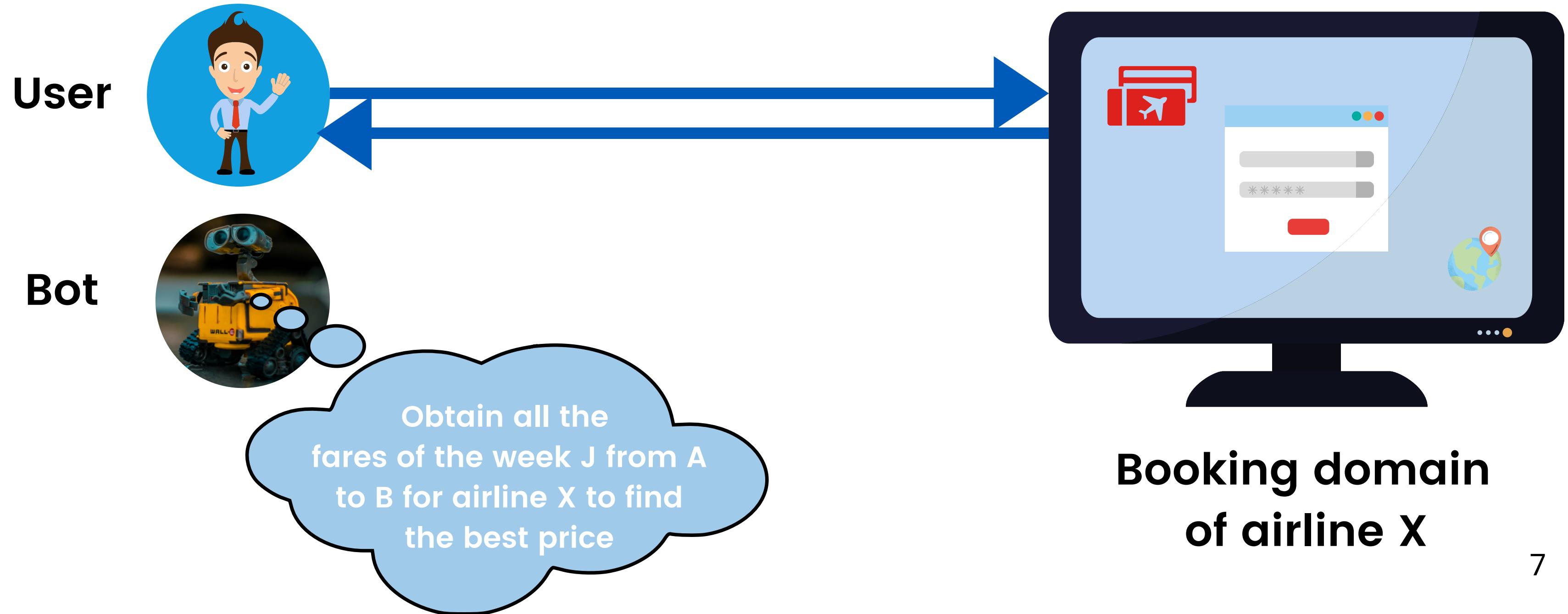


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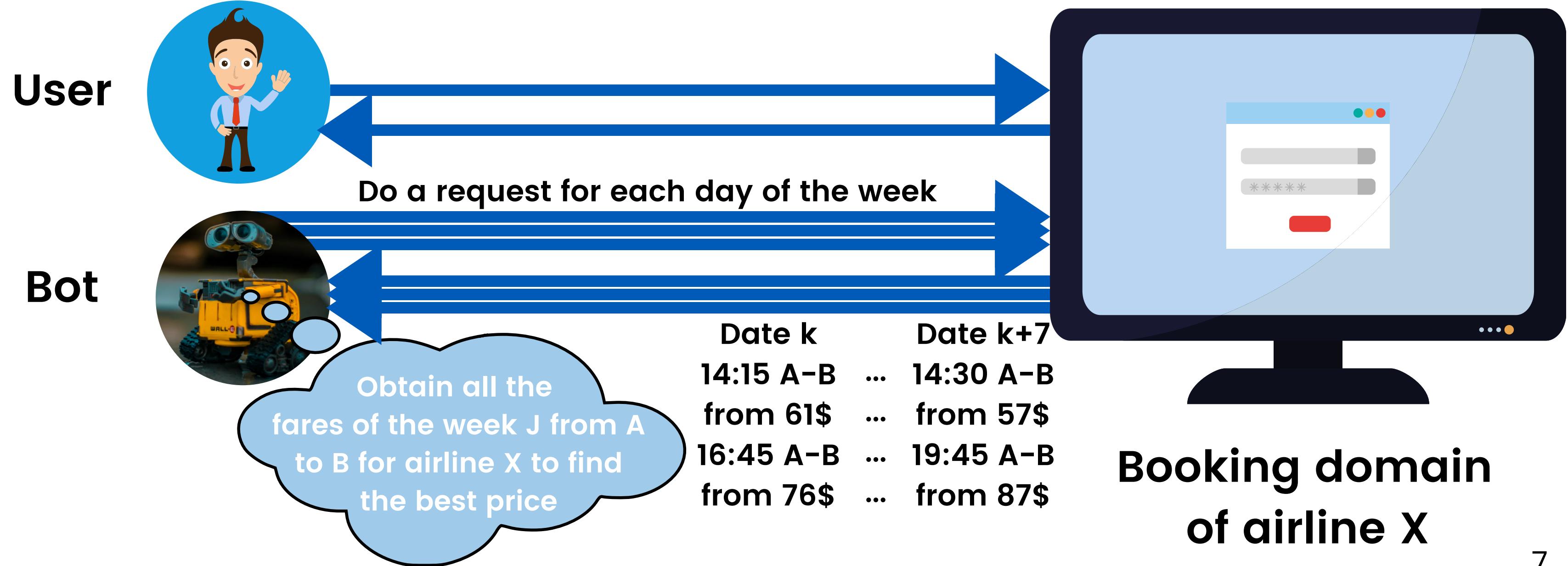
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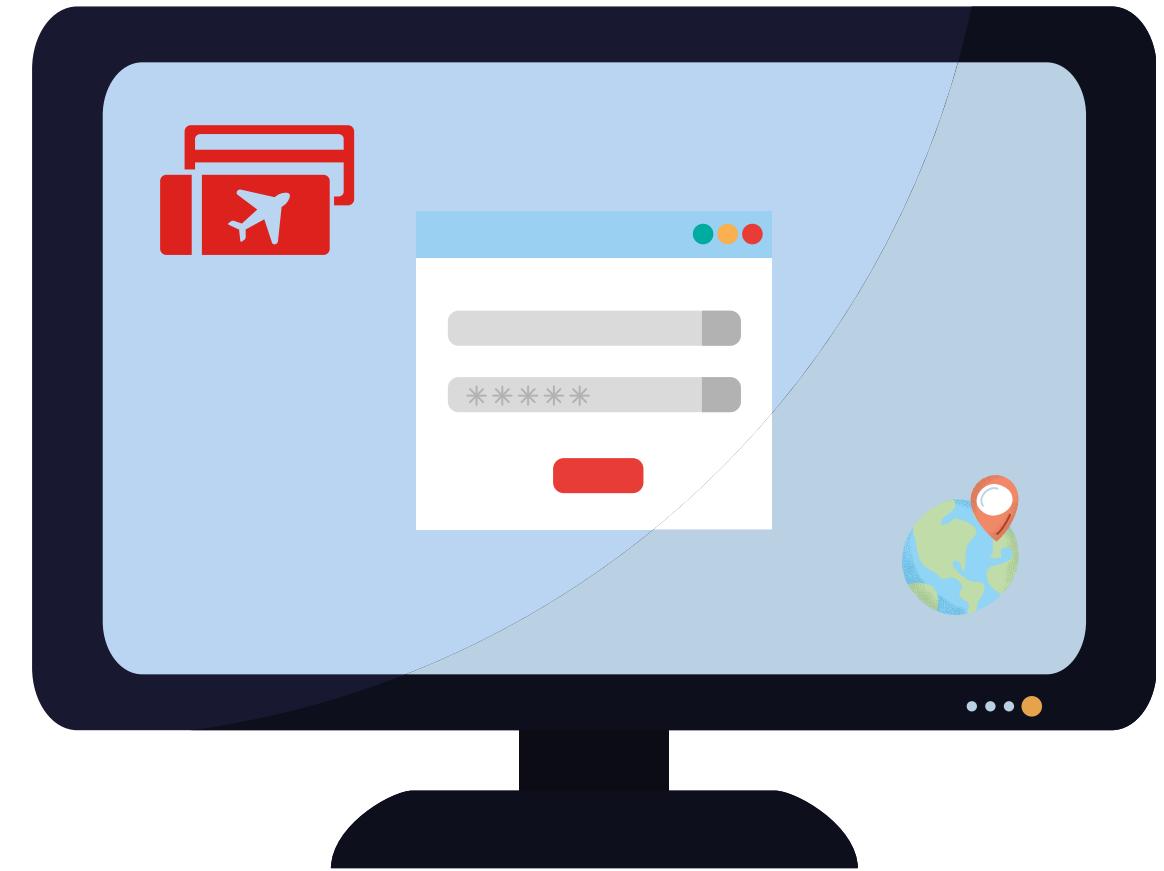
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Anti-bot solutions

User

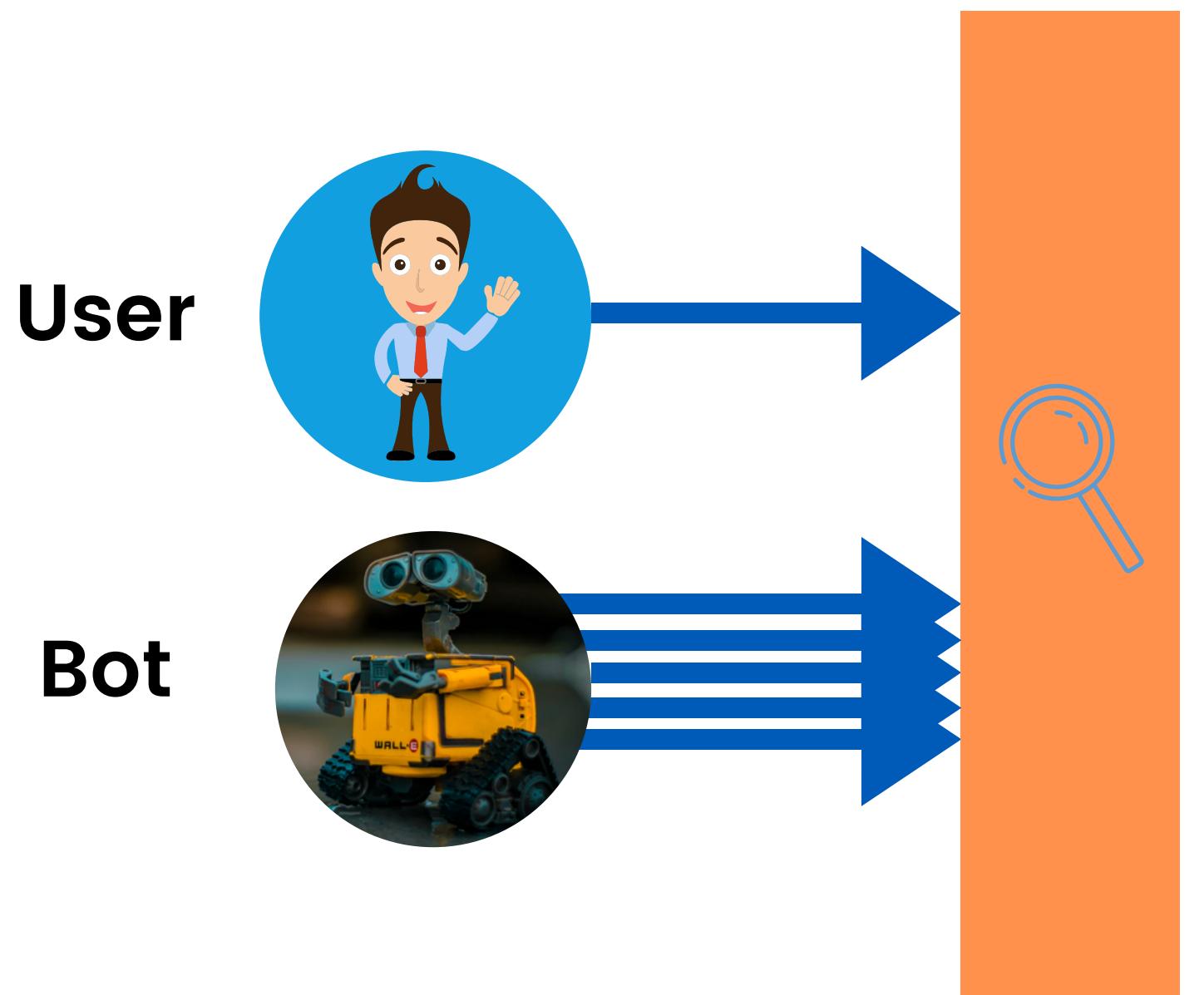


Bot



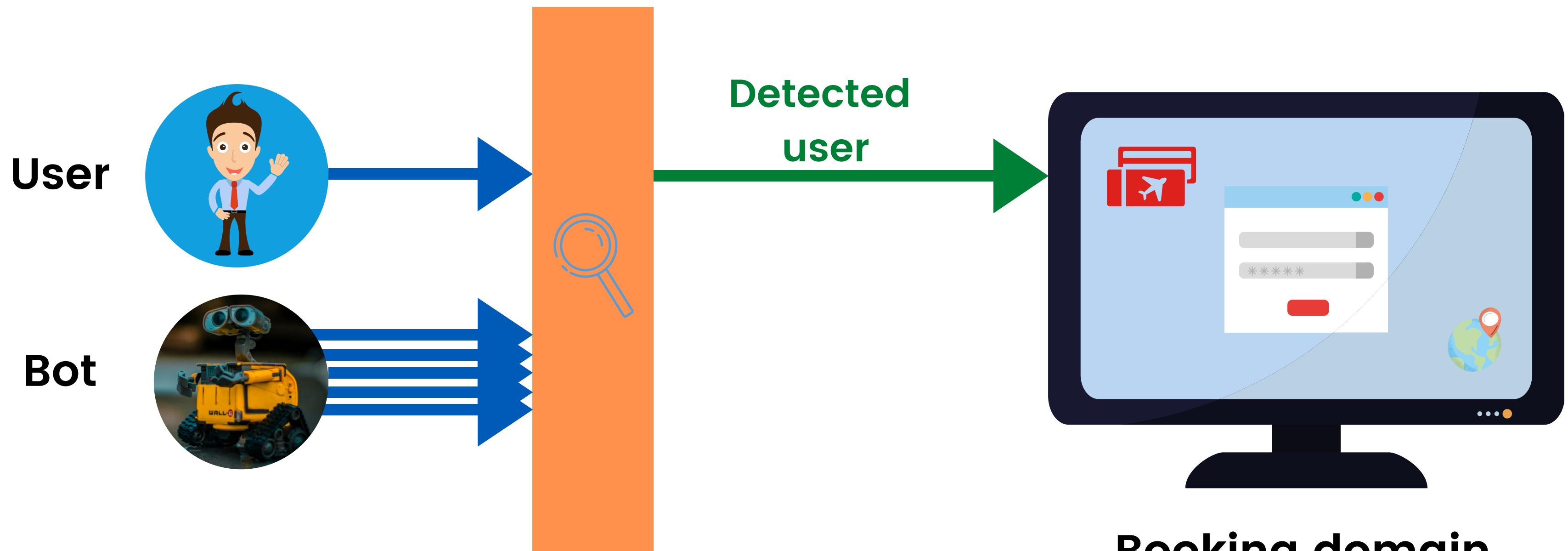
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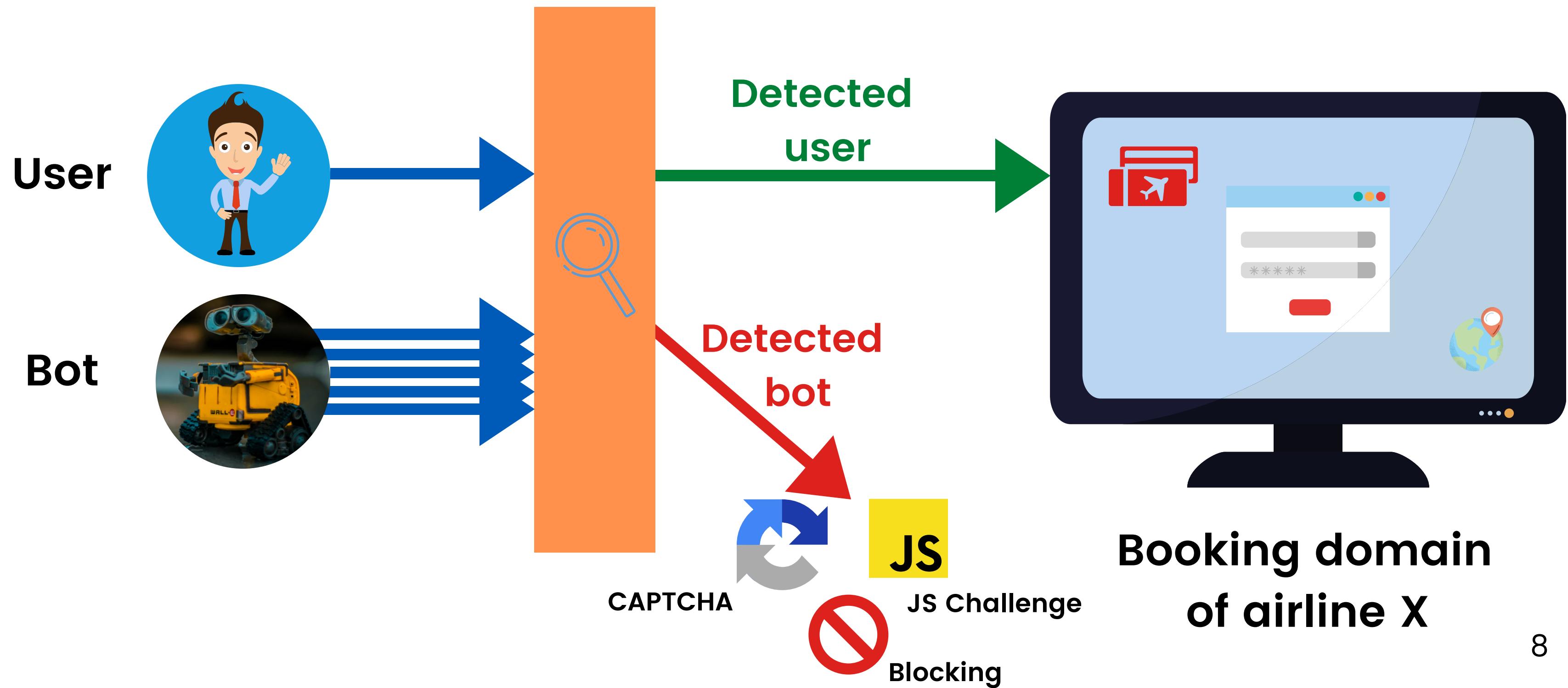


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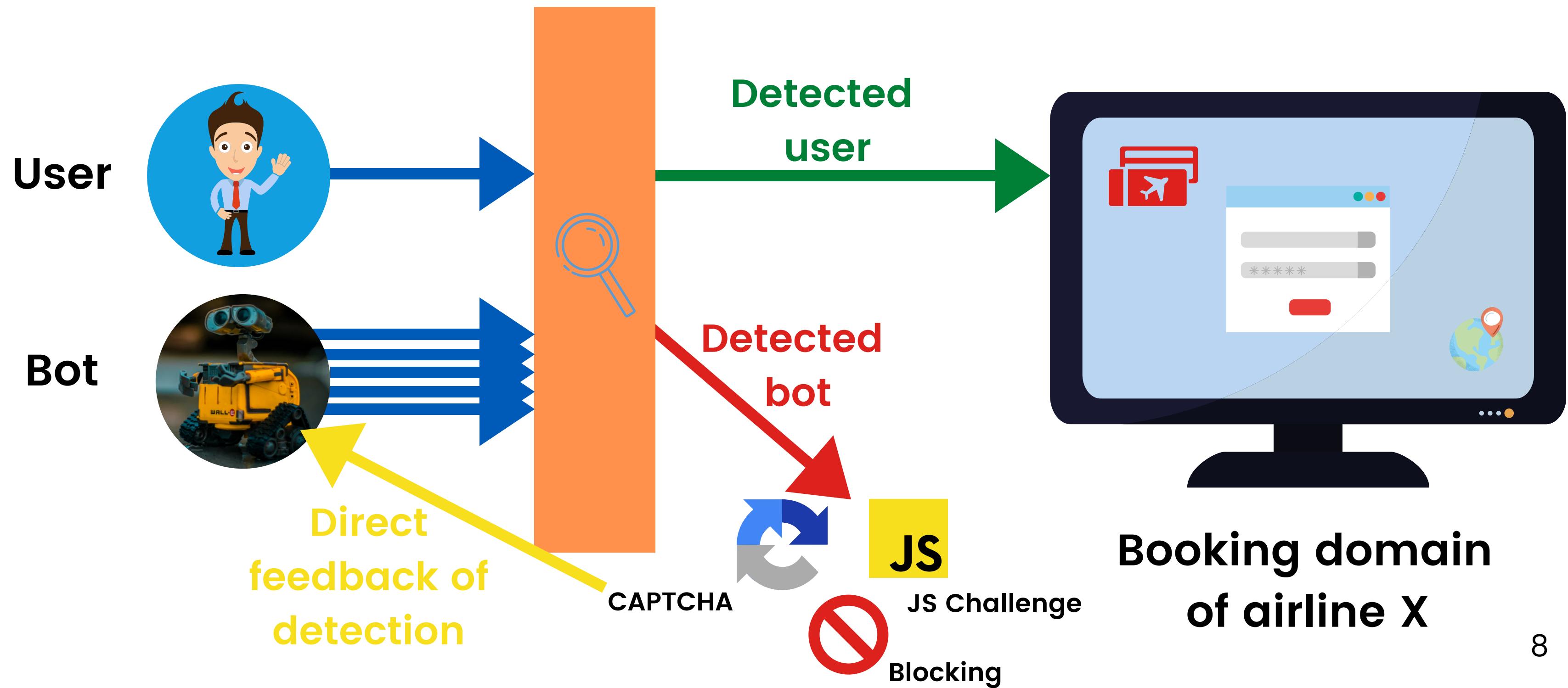
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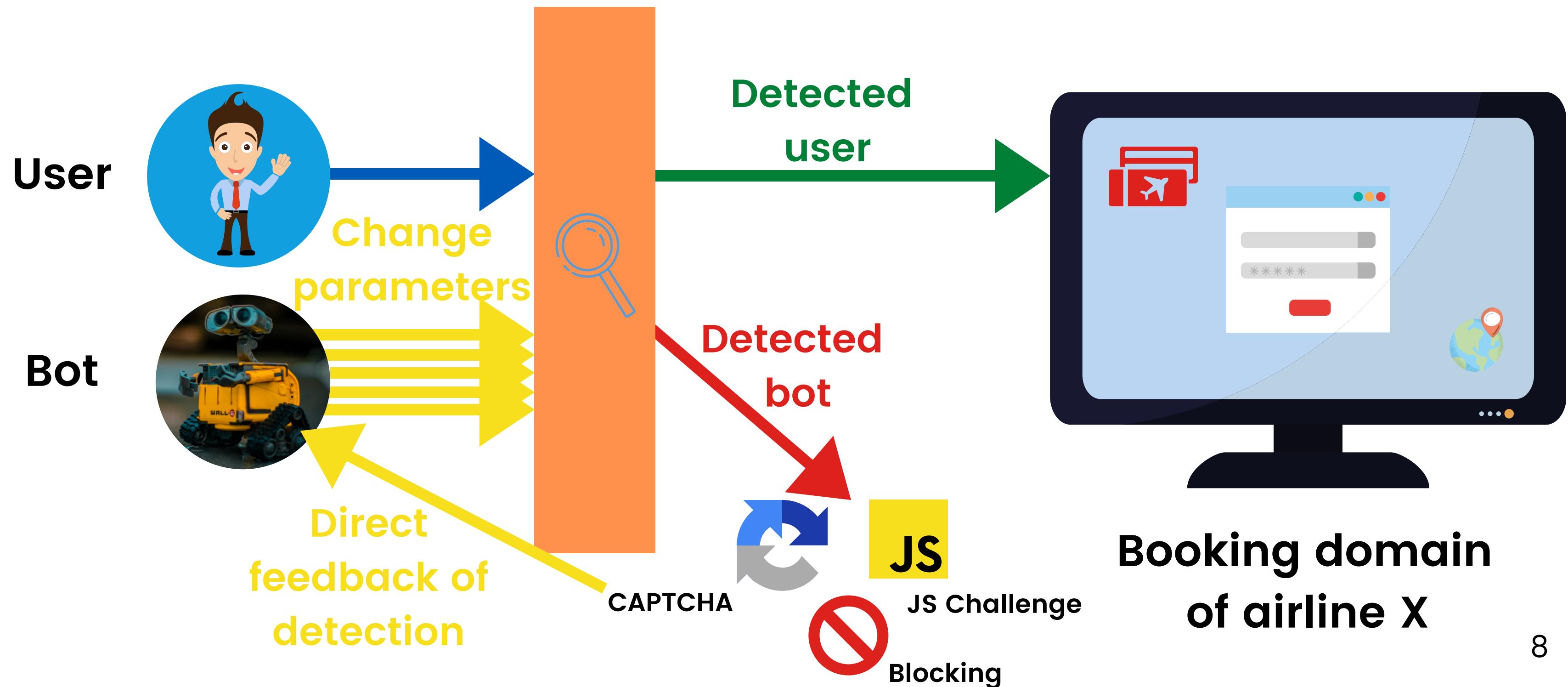
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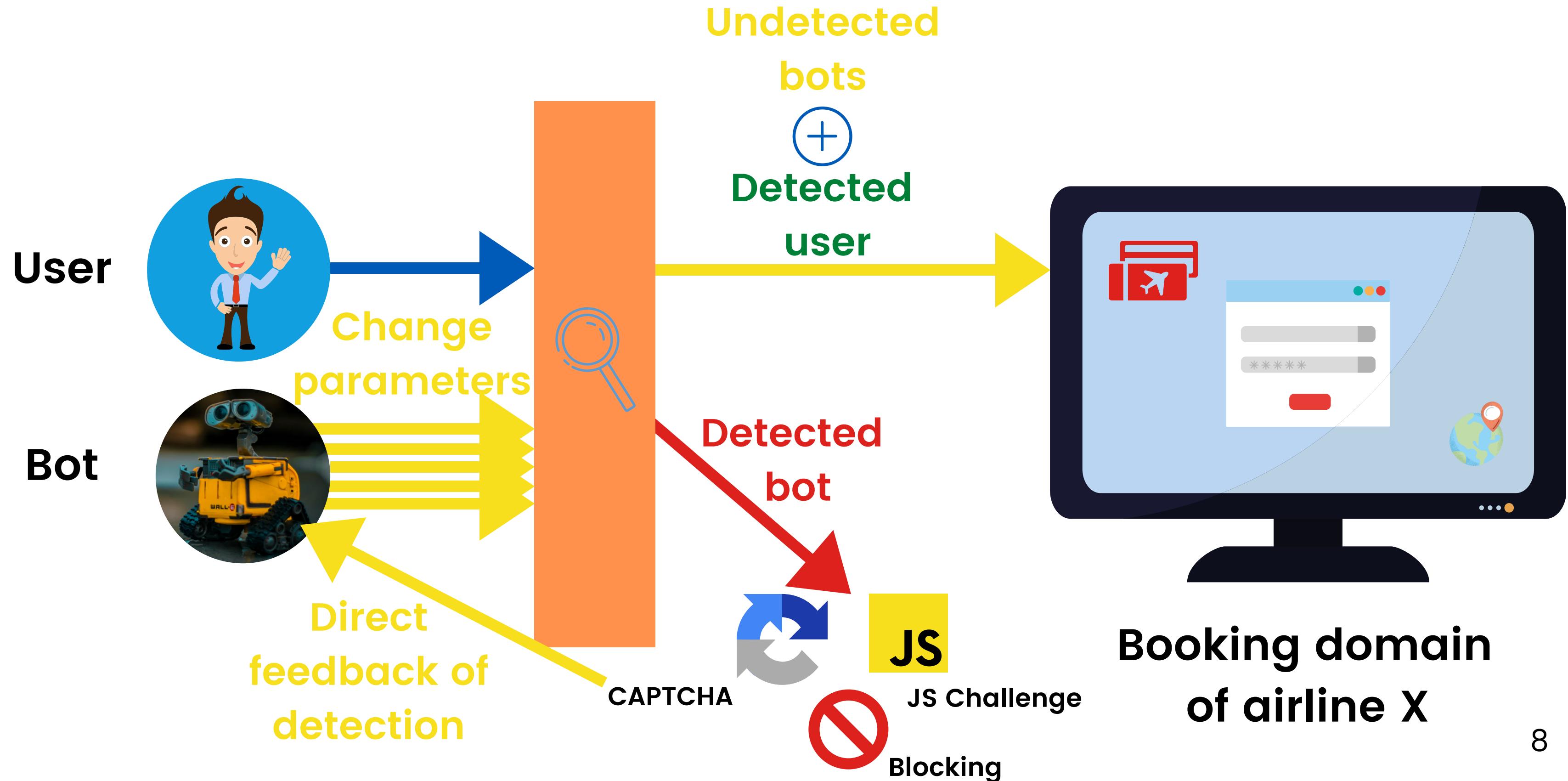
Anti-bot solutions



Anti-bot solutions



Anti-bot solutions



What 3rd parties works tell us

Blocked bots die

- Continuous verification of stealthiness and efficacy of the bots
- Rapid modification to avoid detection

Harnessed information is verified

- Continuous verification of the correctness of the information
- Incorrect information is a direct feedback they have been detected

New
approach

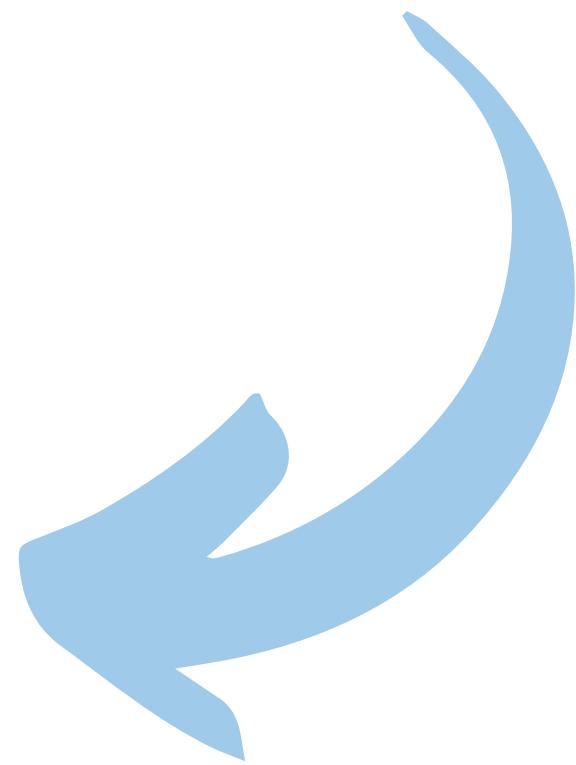
New approach

Prevent bots to know
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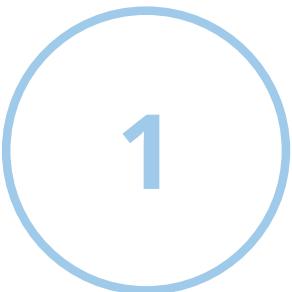
New approach

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Provide bots
incorrect but
plausible answers



Our discoveries



Identification of specific super stealthy bots, with extreme distribution of activity (one request per IP in most cases)

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Identification of a behavioral pattern among these requests, which brings hope for another form of detection

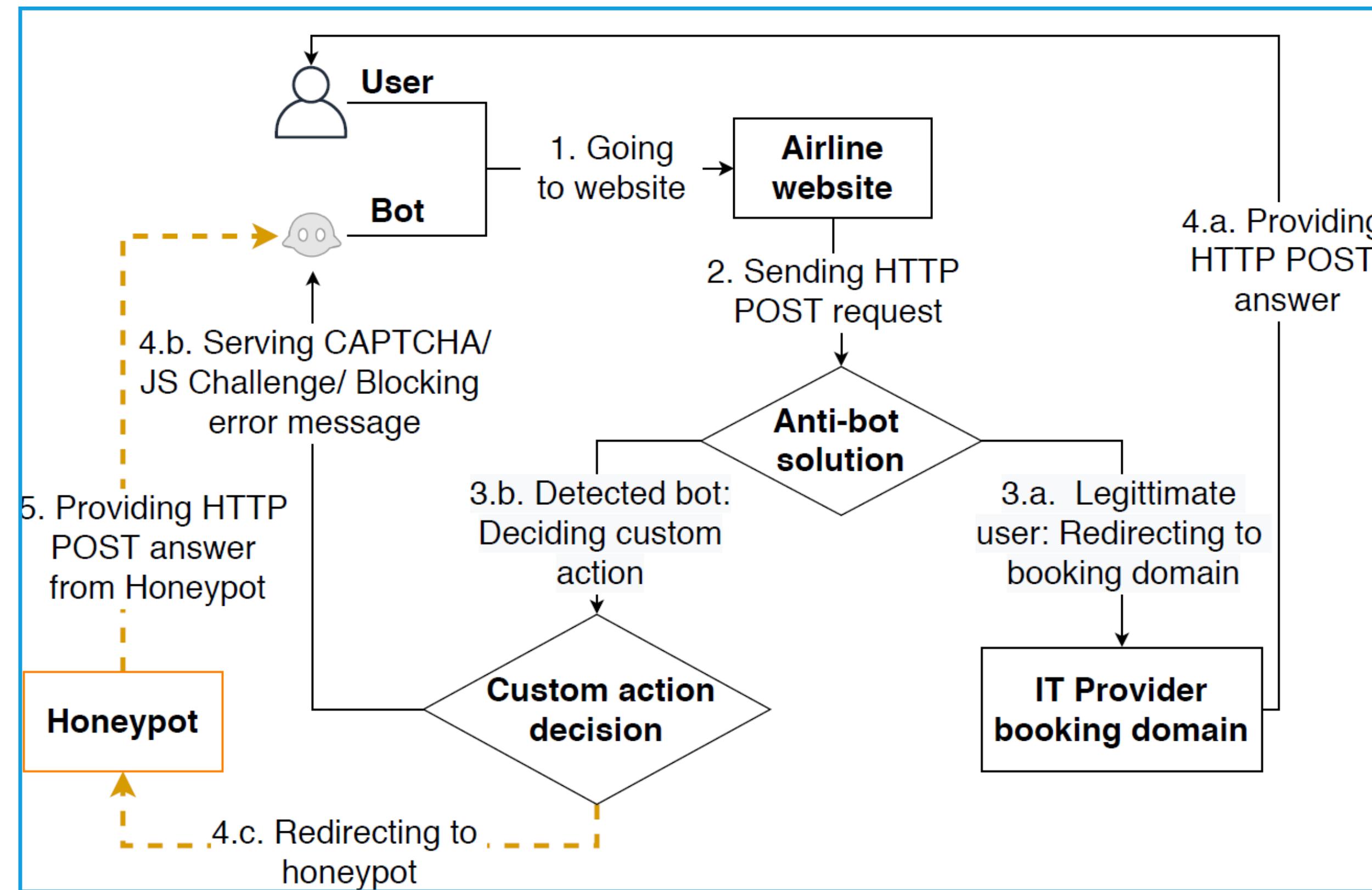
Our discoveries

- 1 Identification of specific super stealthy bots, with extreme distribution of activity (one request per IP in most cases)
- 2 Identification of a behavioral pattern among these requests, which brings hope for another form of detection
- 3 Success in providing inaccurate information without being detected

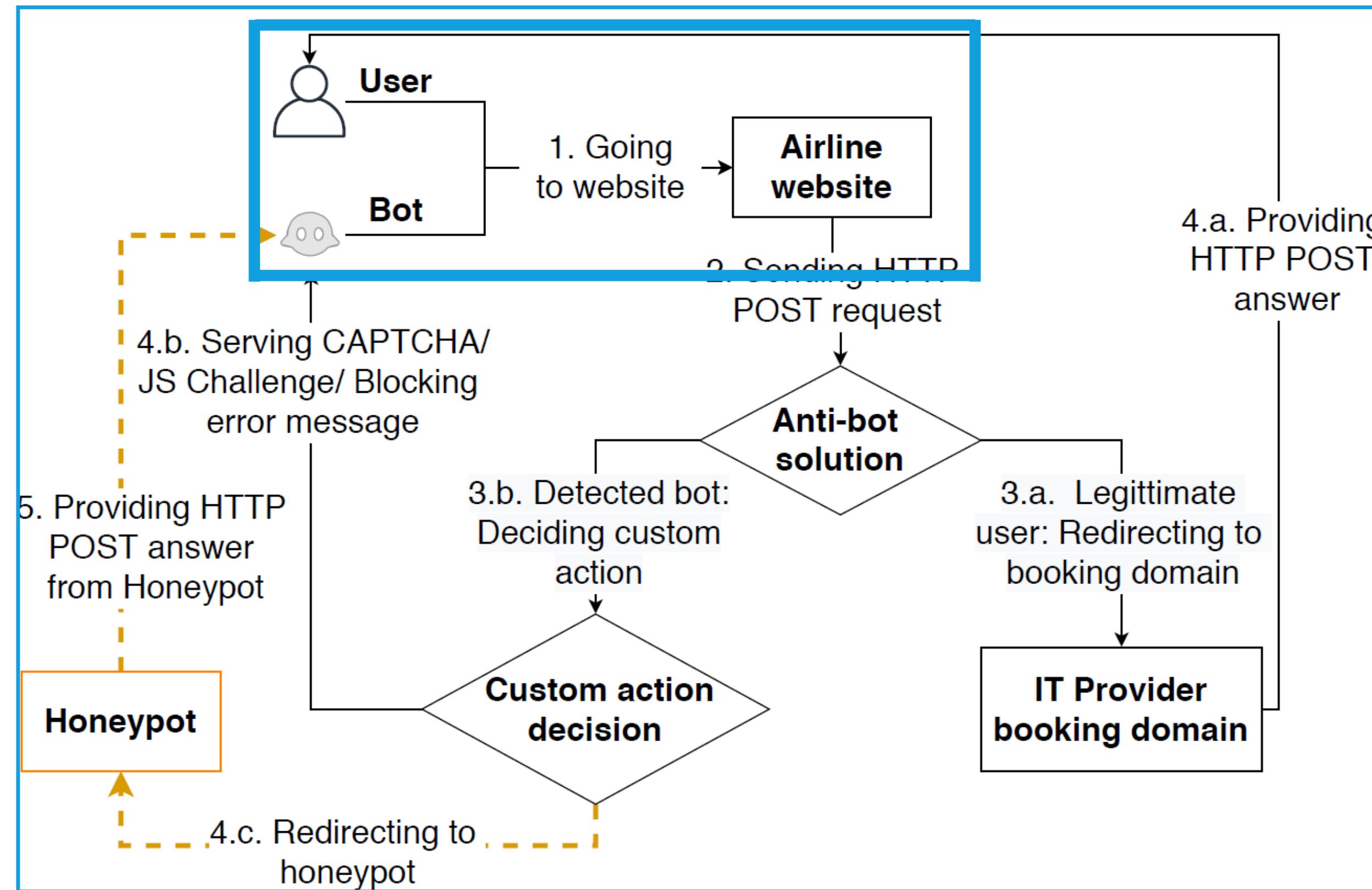


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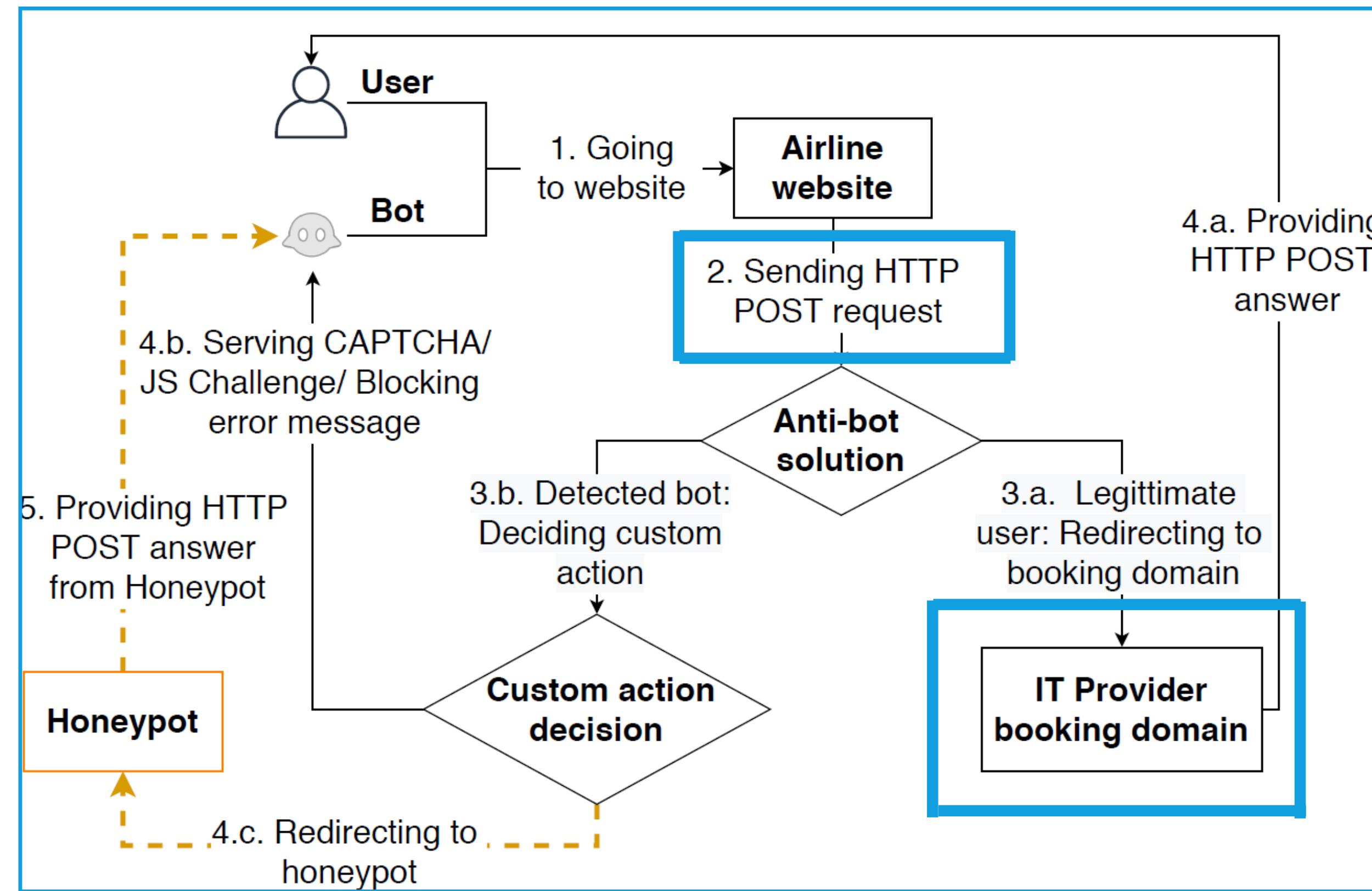
The architecture



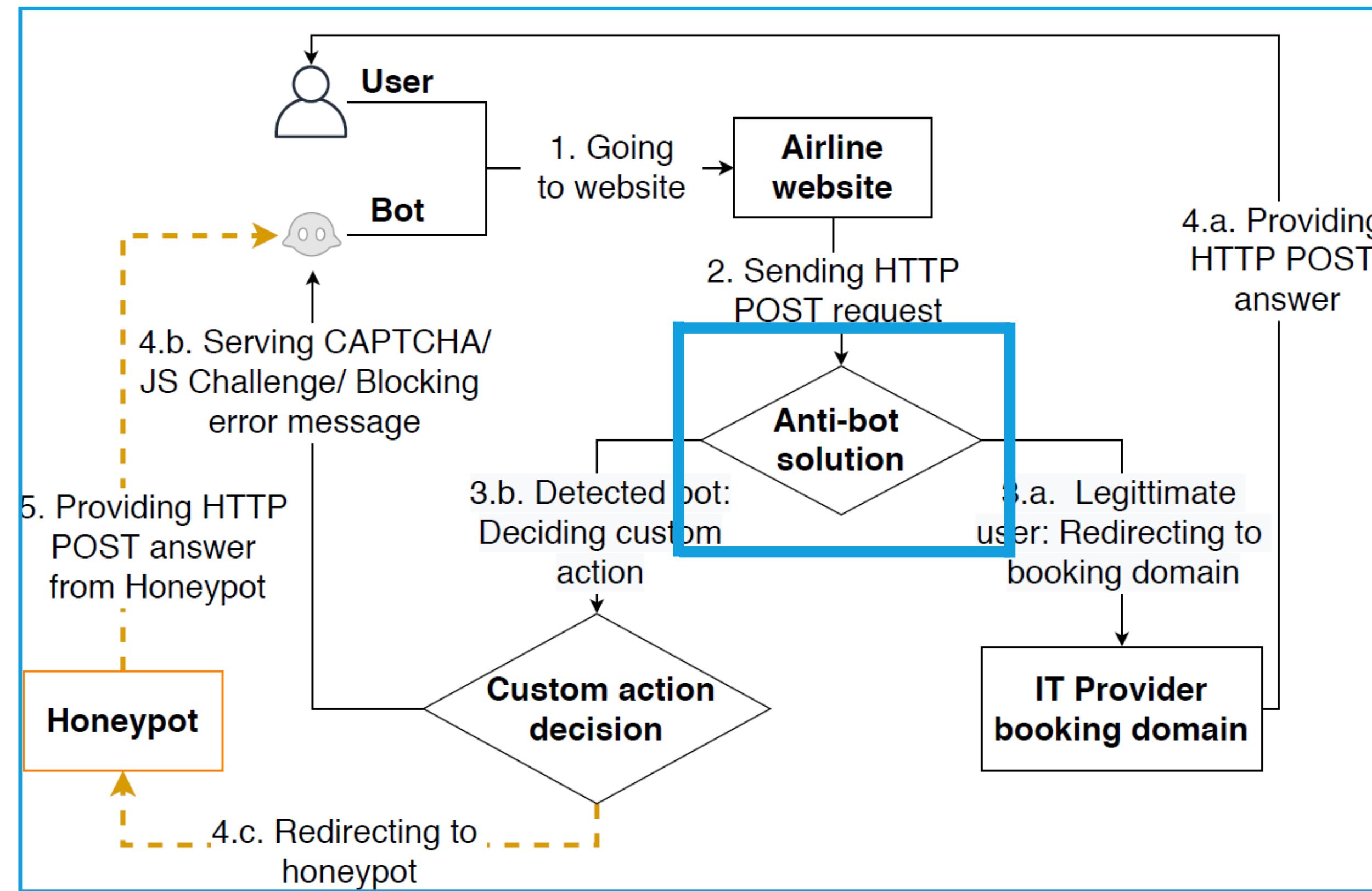
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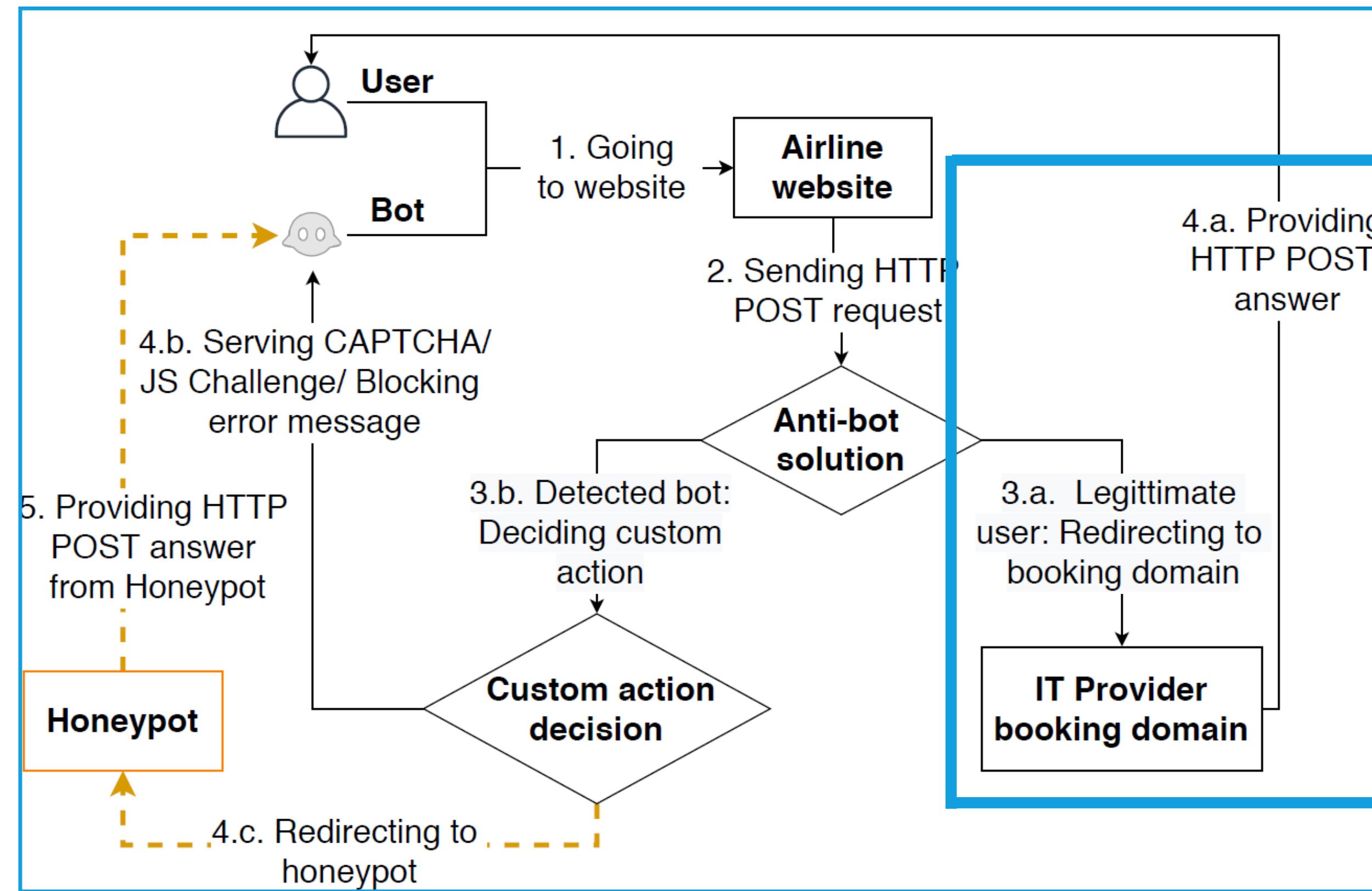
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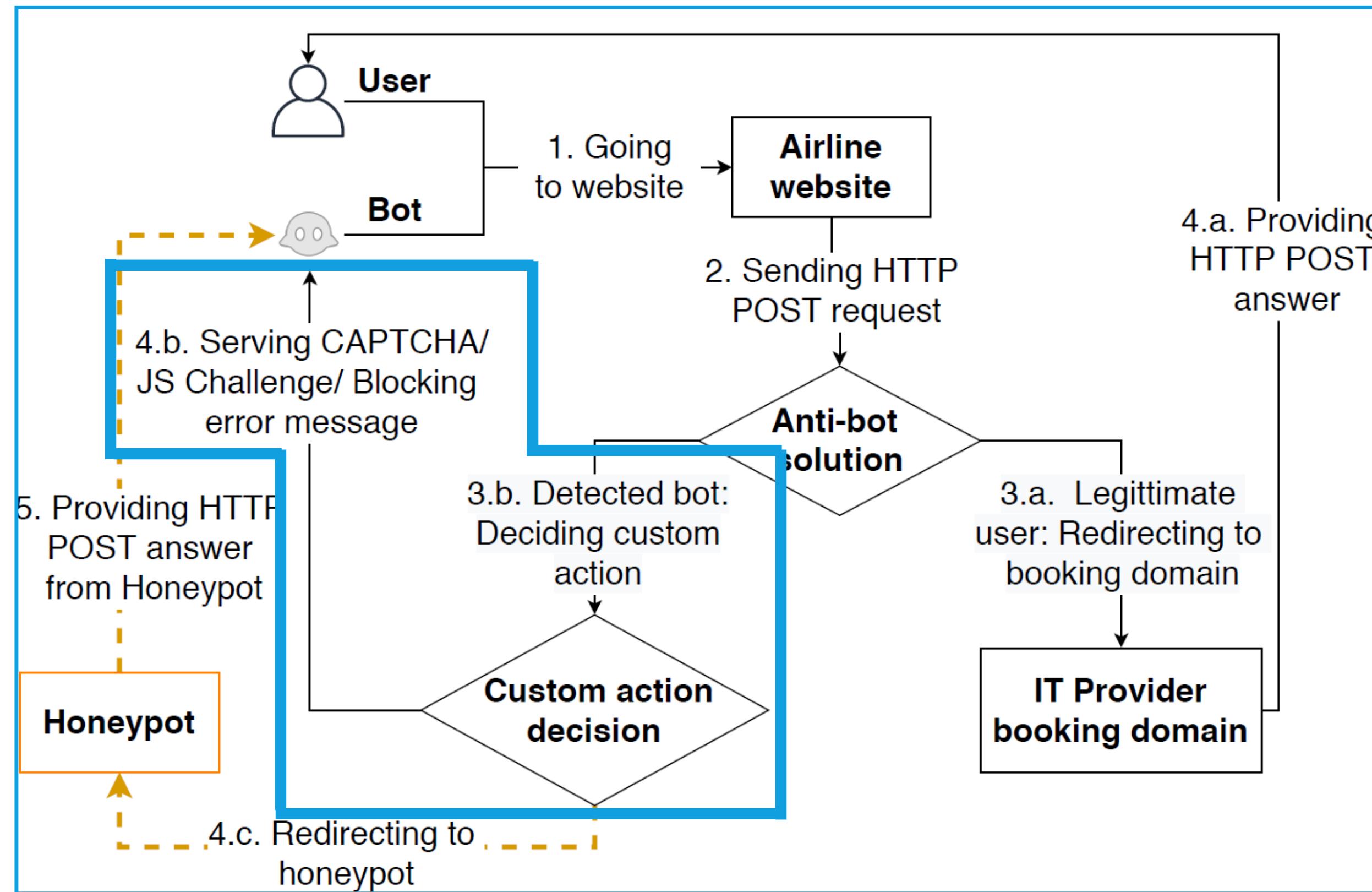
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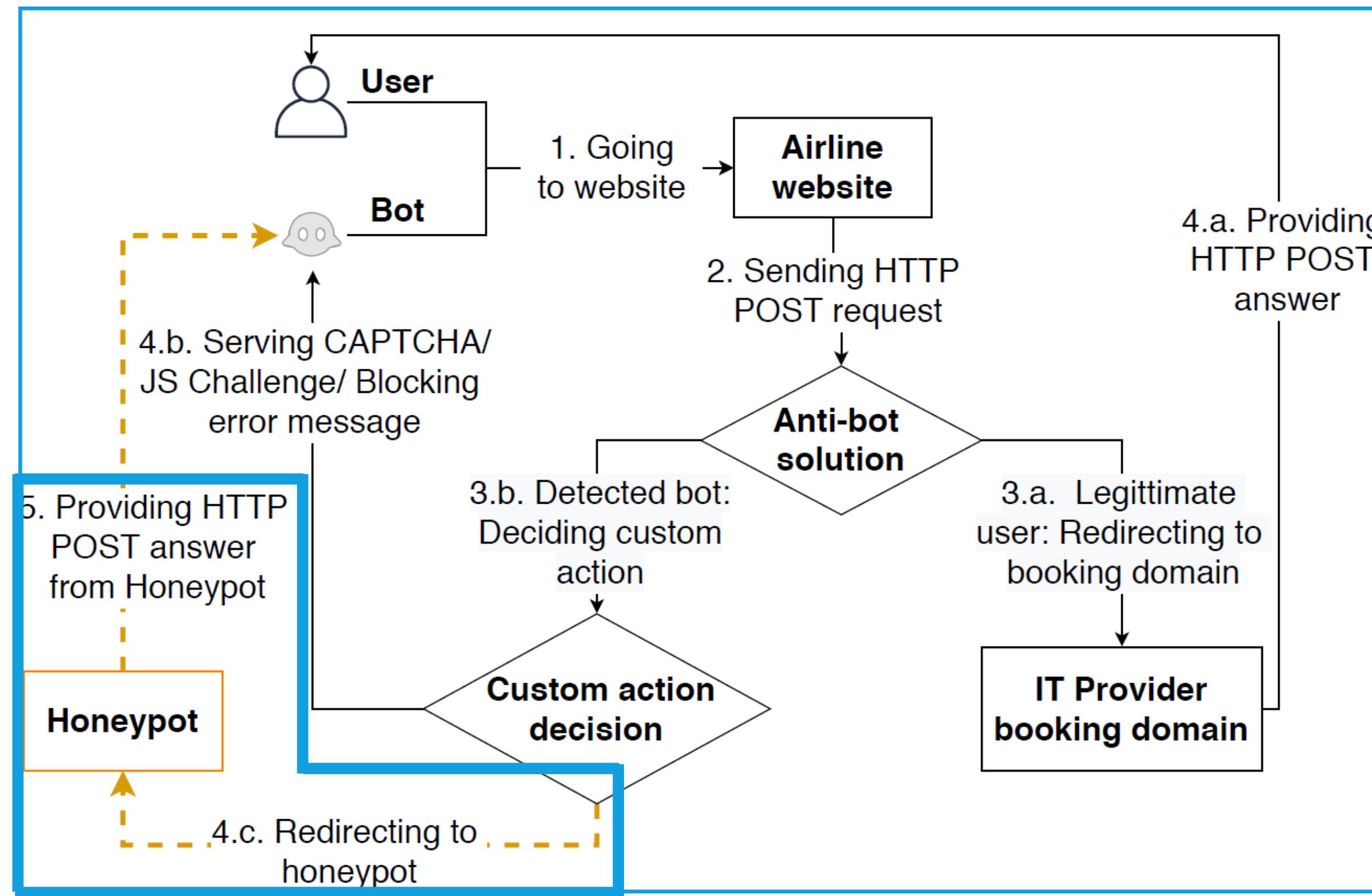
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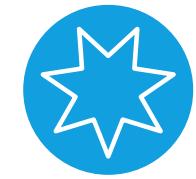
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Some numbers...



Pilot airline

Ticketing system: 1 million requests per day, 40% detected as coming from bad bots

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Chosen bot

40-minutes daily time window of activity, no bookings, 400-500 requests per day

Pricing strategy



After 3 days, modification
of fares: increase 10% of
the requests by 5%

Goal: understanding if
the bot master was
checking for anomalies in
the price

3. Results and discussion



General analysis

- Experiment running for 56 days (interruption linked with COVID-19 restrictions on flights)

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General analysis

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- ✓ Reception of 22,991 HTTP requests at the Honeypot
- ✓ No change of behavior from before and during the experiment

Lessons learned modifying values

No ground truth to compare returned values

Plausibility check not sophisticated enough for small changes

IP addresses study

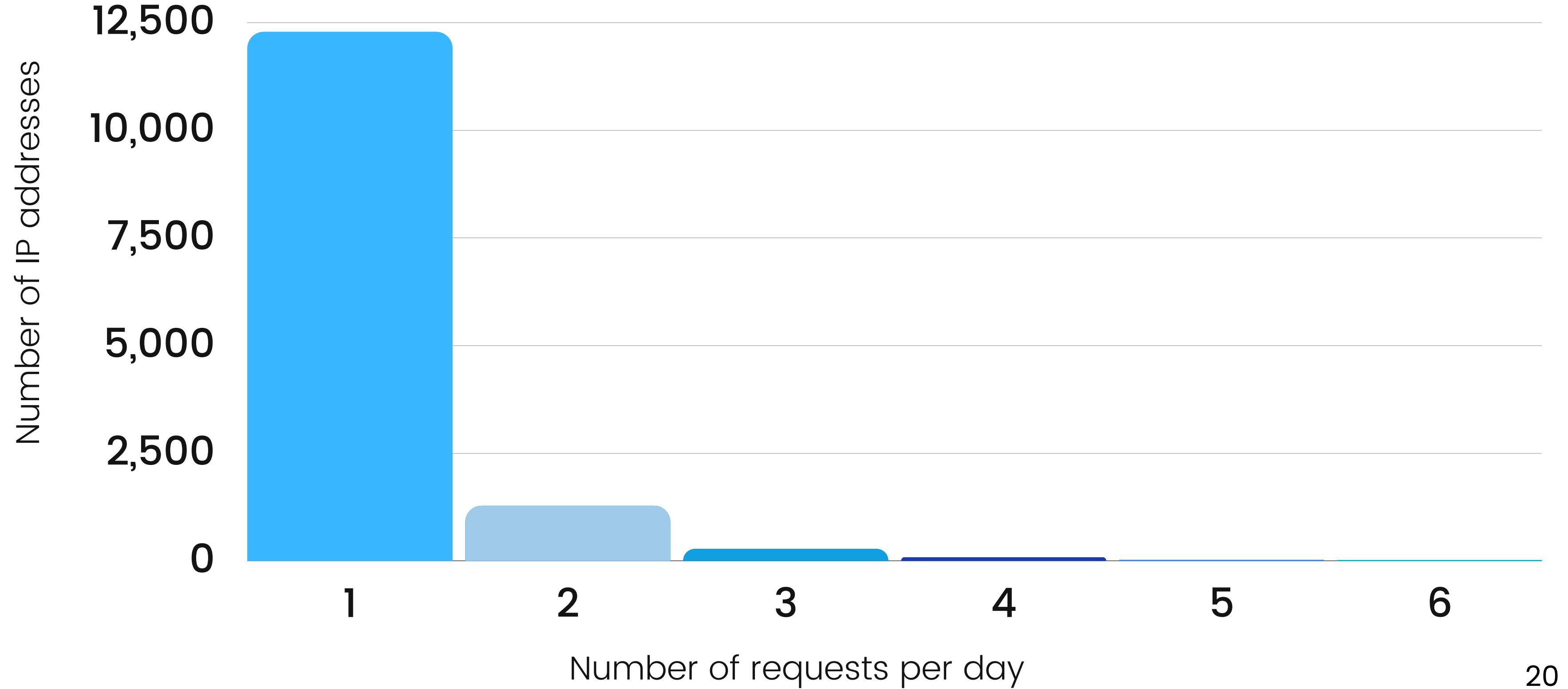
13,897
different IP
addresses

1,187 /16
blocks

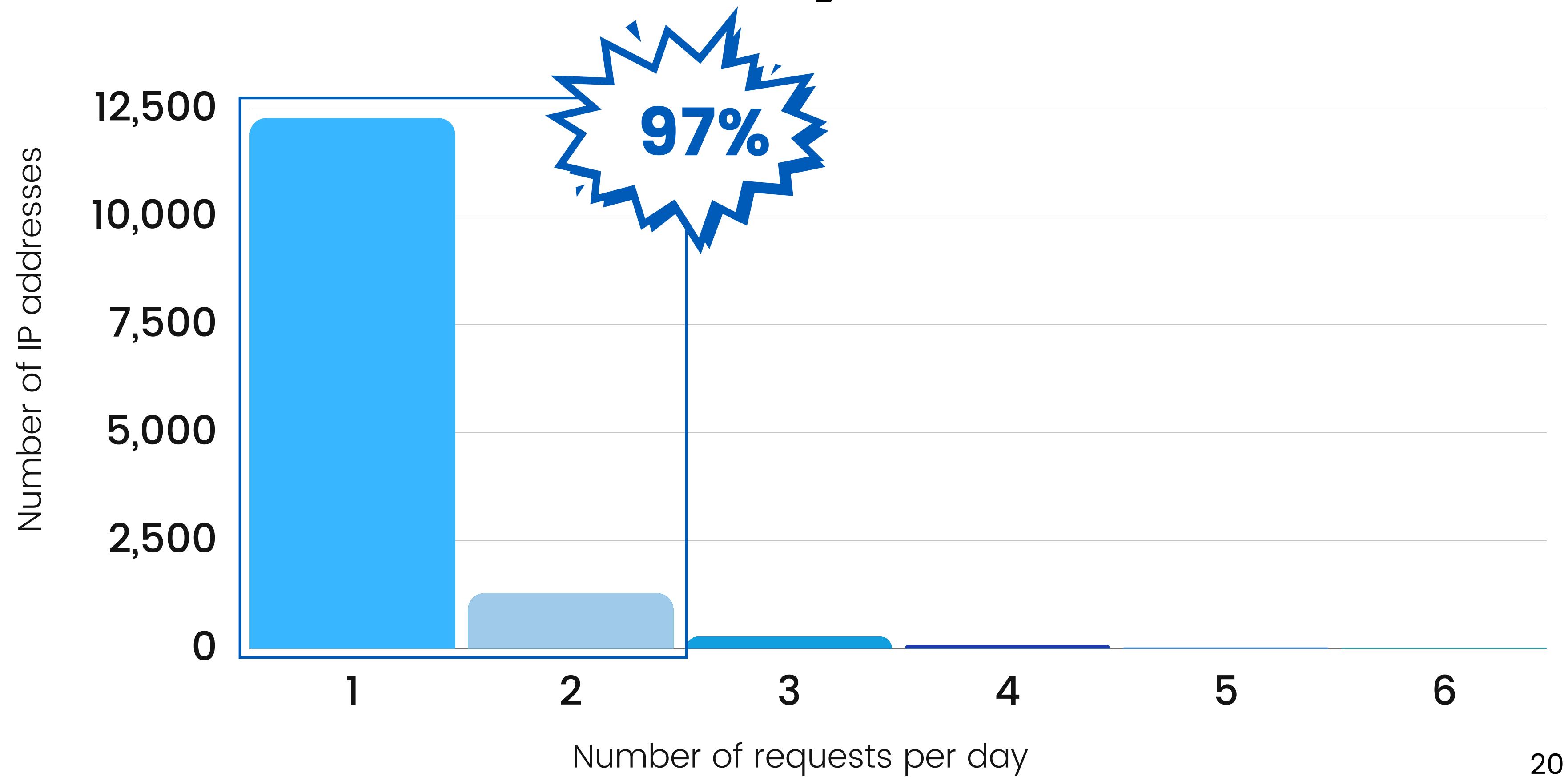
790 distinct
cities

86 countries

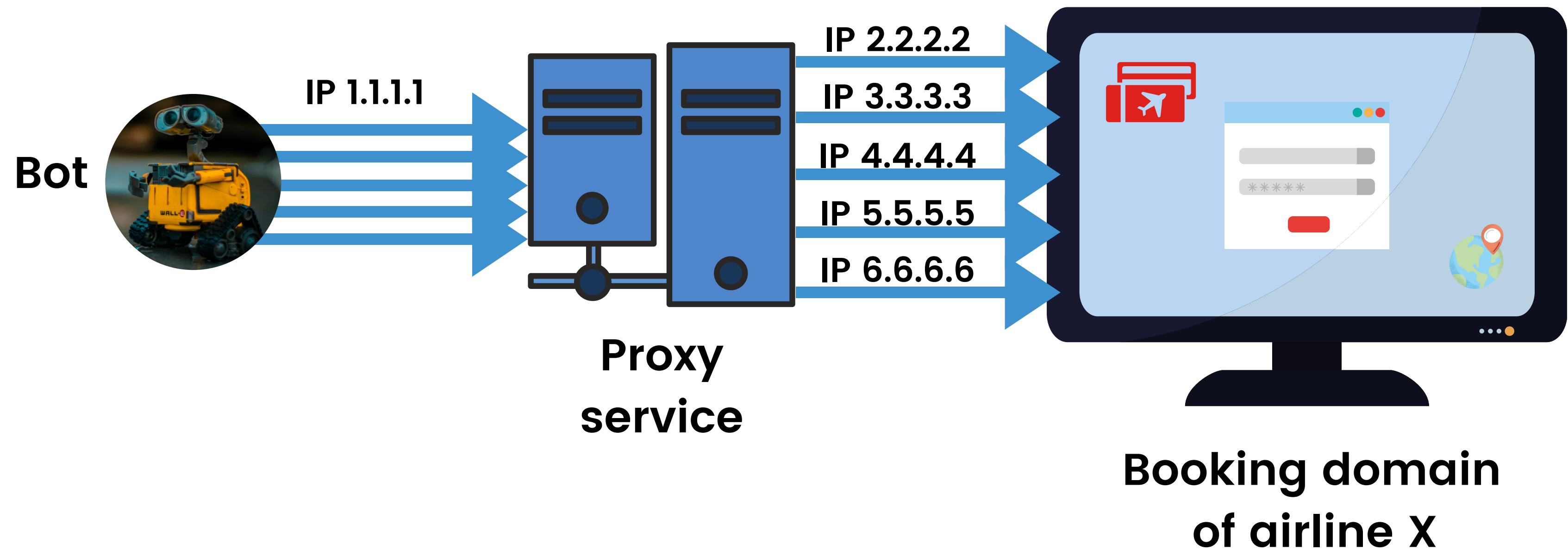
IP addresses study



IP addresses study



Residential proxy as a service



Proxy services IPs

Proxy services offer
millions of IP
addresses

Proxy services IPs

Proxy services offer
millions of IP
addresses

BUT

Proxy services IPs

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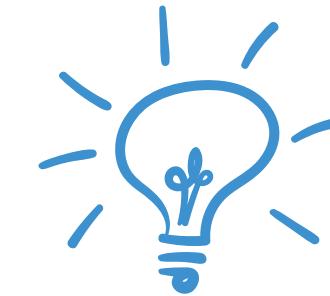
32% of the IPs
appeared on
more than one
day!

TOR network

Date	Number of TOR IPs
2020-02-14	12
2020-02-15	24
2020-02-18	20
2020-02-19	40
2020-02-24	12

TOR network

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Bots tried to take advantage of TOR nodes but they did not continue

Behavioral analysis



Behavioral analysis



51,5% of requests for return flights

Behavioral analysis

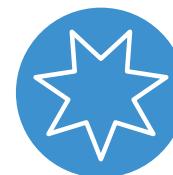


51,5% of requests for return flights



Return flights: 7 days period

Behavioral analysis



51,5% of requests for return flights



Return flights: 7 days period



Only 25 combination of departure and arrival airports, small fraction of the airline's offer

Time interval



Time Interval=Departure date - Date of the request

Time interval

- ▲ Time Interval=Departure date - Date of the request
- ▲ Value between 0 and 14 days or 21, 30, 45, 60, 90, 120 days

Time interval

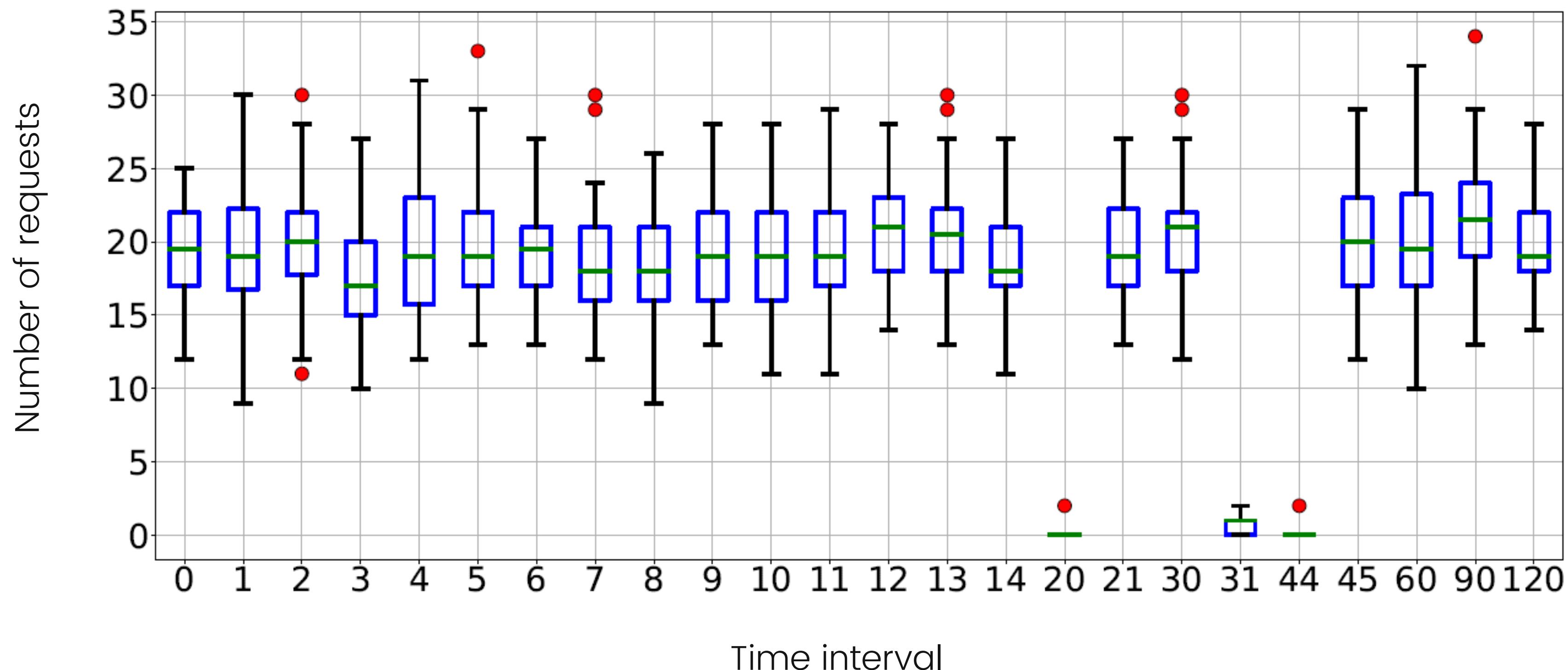
- ▲ Time Interval=Departure date – Date of the request
- ▲ Value between 0 and 14 days or 21, 30, 45, 60, 90, 120 days
- ▲ Only 0.2% of the requests exhibit different values but out of the 40 minutes daily time window

Distribution among segments

Segment=combination of one way/return flight, departure and arrival location

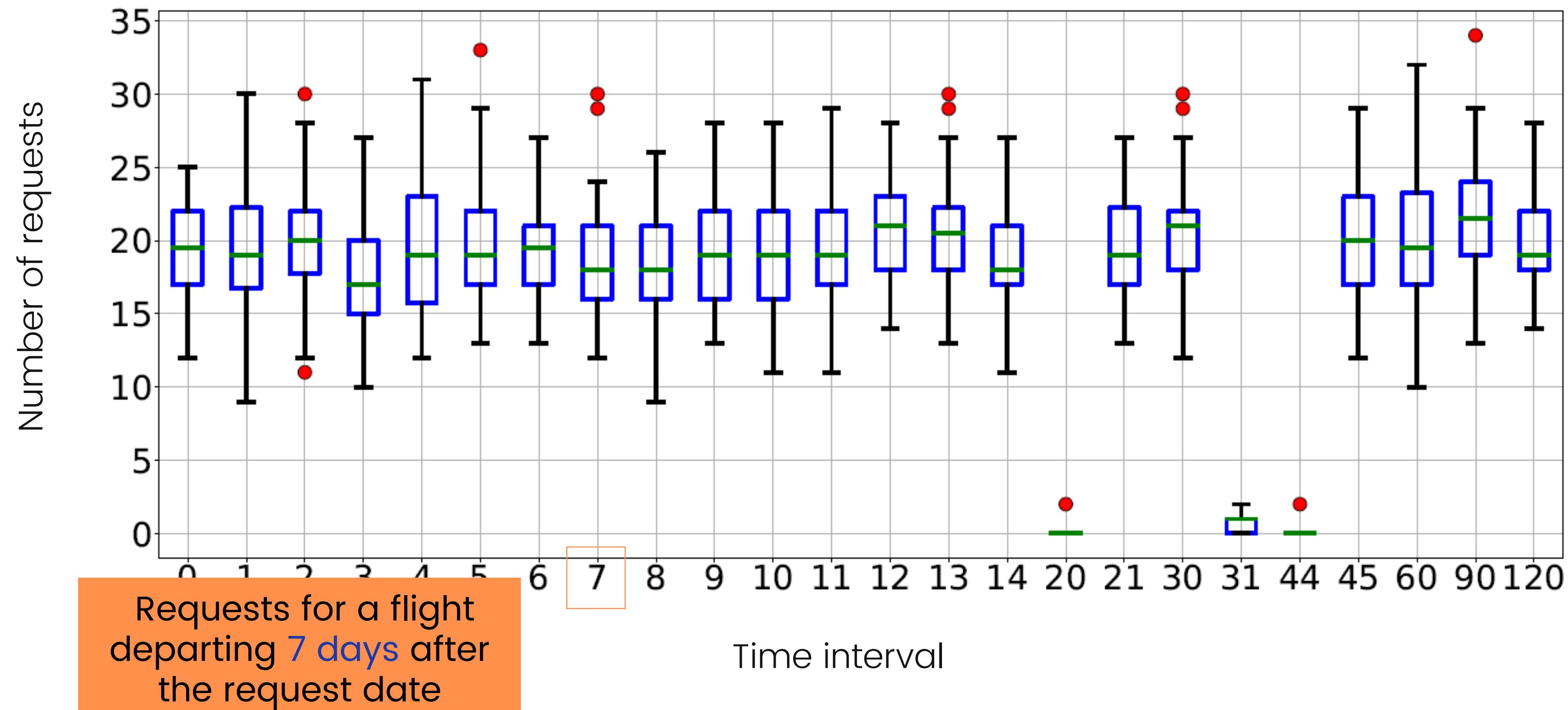
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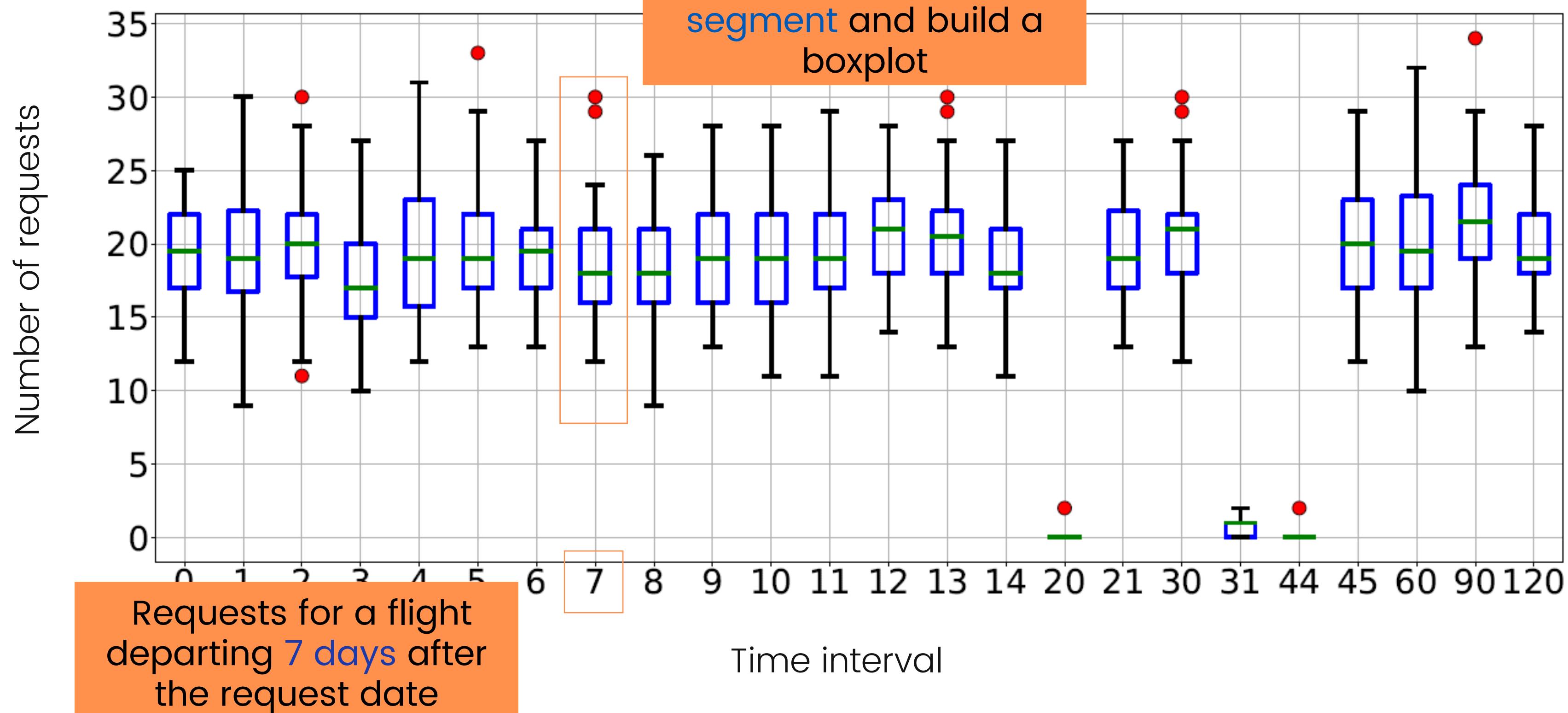
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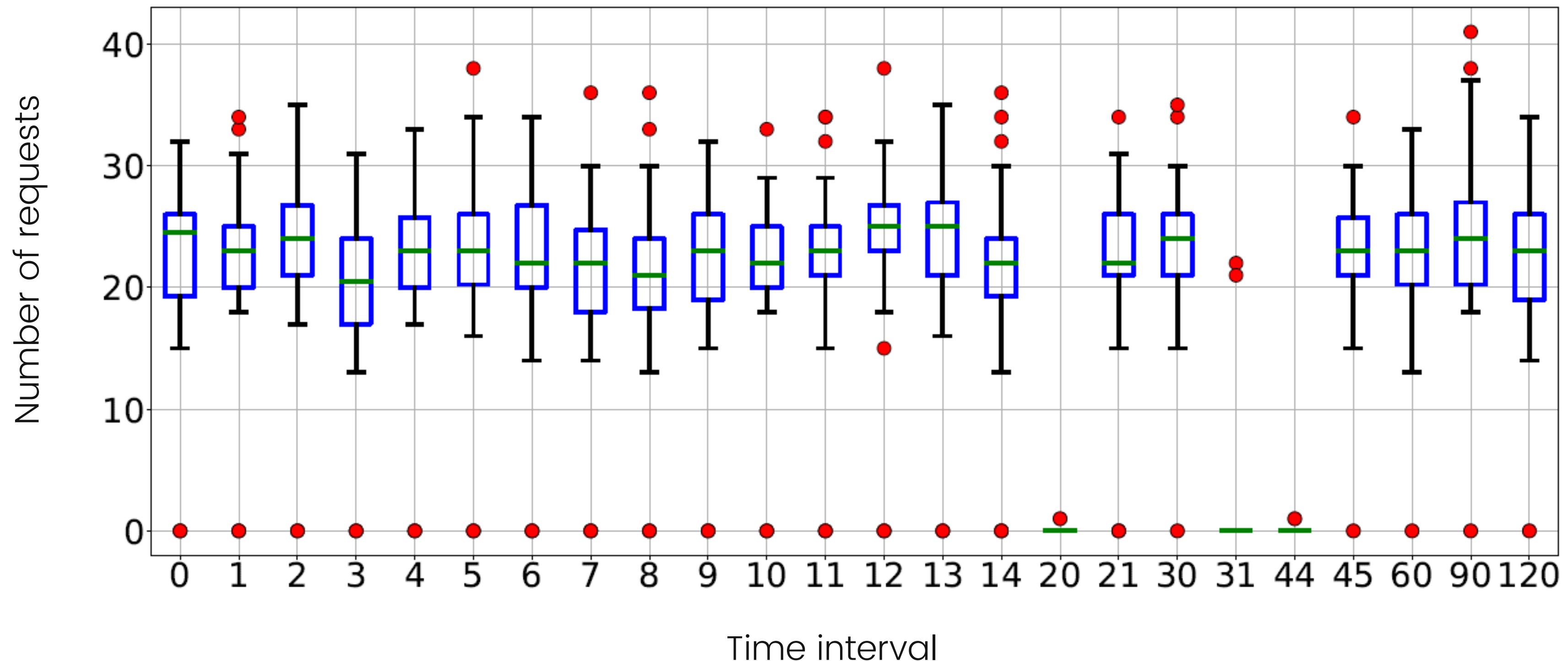


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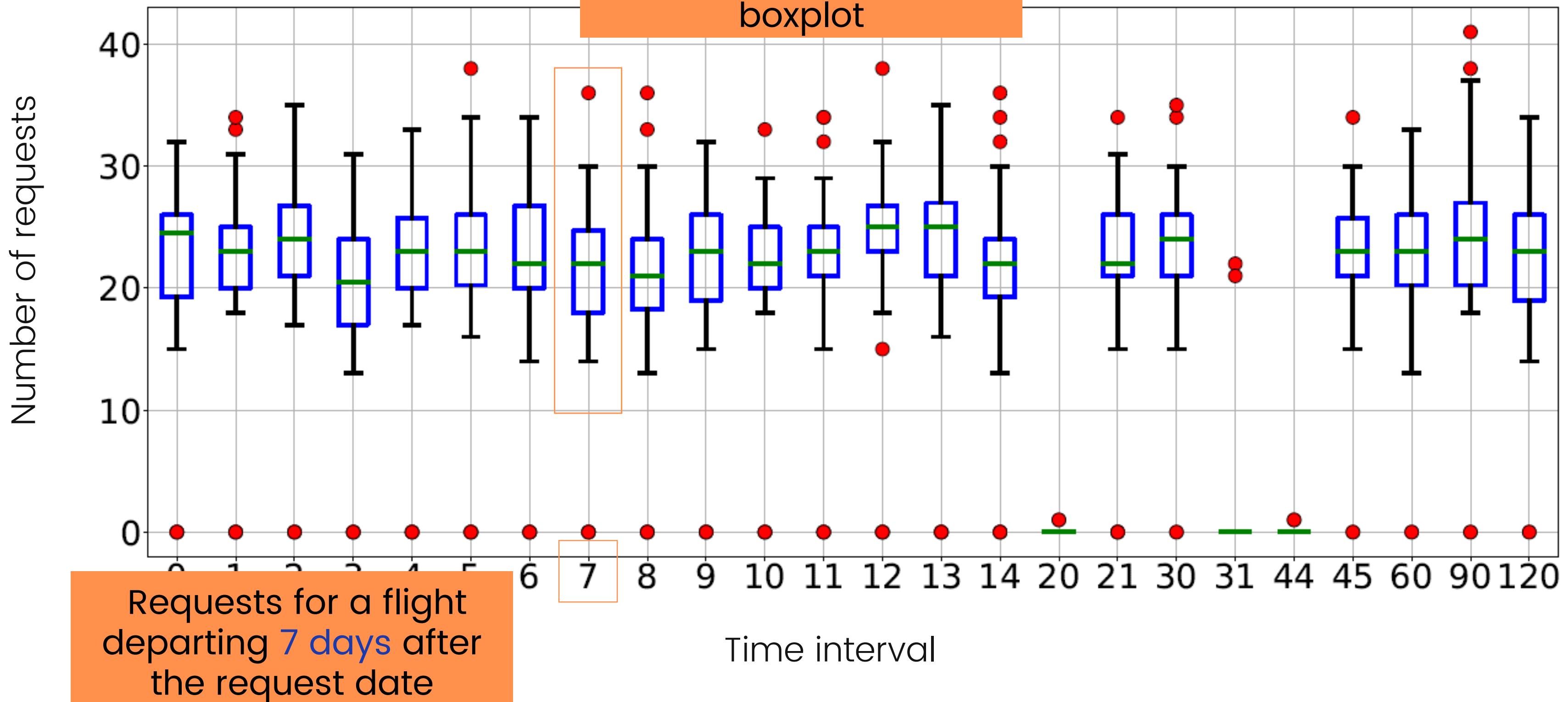


Distribution for request date



Distribution for request date

Divide them for date in which the request was made and build a boxplot



Tuples statistics



4-tuples made of:

- ▶ Departure airport
- ▶ Arrival airport
- ▶ Time interval
- ▶ Type of flight (one way/return)

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982 distinct tuples vs 410 average daily requests

Tuples statistics

- Each tuple is asked on average 23.41 times

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 - All requests done in little span of time
 - Maximum difference: 5 minutes and 30 seconds

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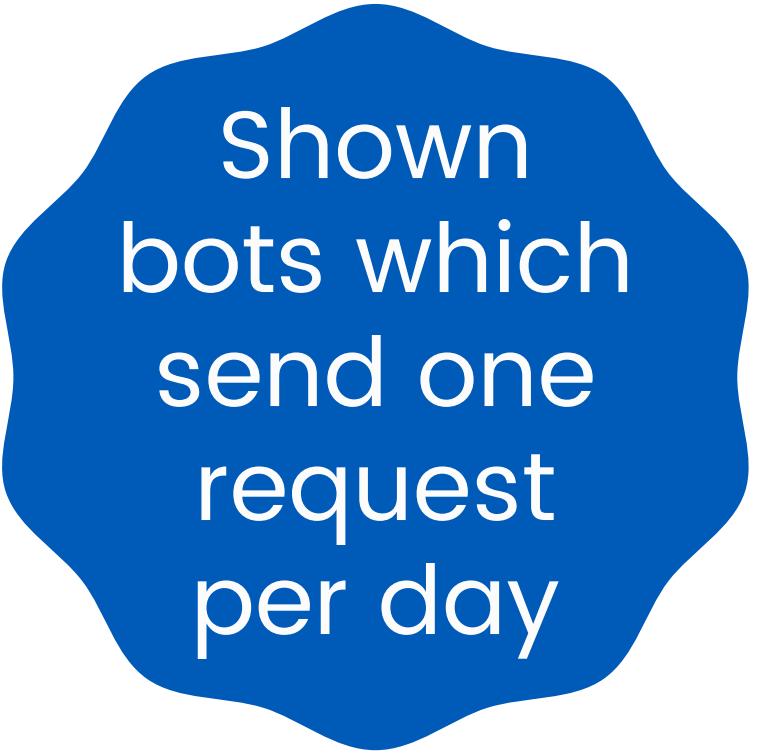
Check for consistency, but not sophisticated enough

4. Conclusions and future work



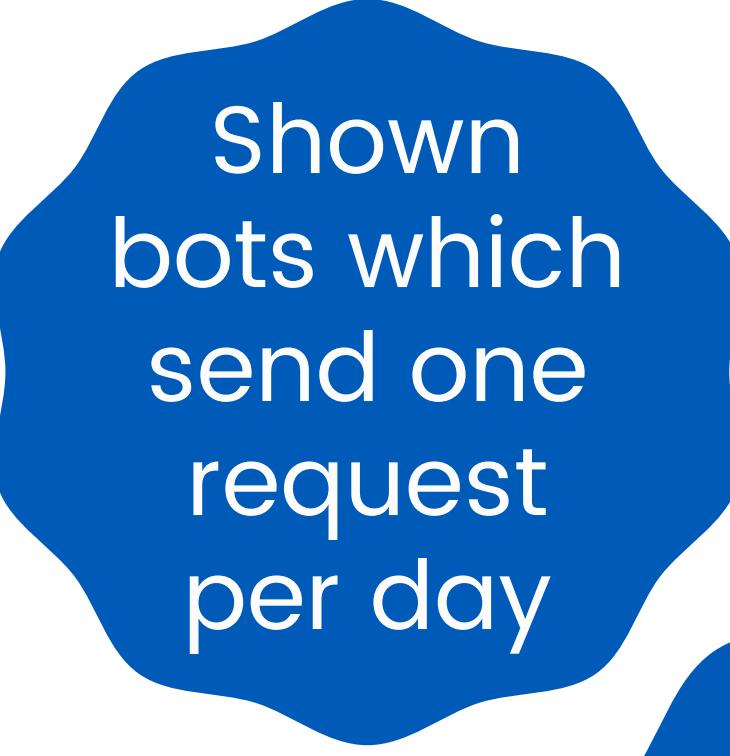
What we have done

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Found behavioral
pattern that
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Future works

Serve bot with
cached prices

Because of:

- Repetition of requests
- Loose verification

To do:

- Sensitivity analysis

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New method
to block IPs

To do:

- Longitudinal large-scale analysis of suspicious IPs exhibiting the same behavior

Thank You



How to reach us

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