

Time analysis Based Attacks Simulation in Tor Networks.

Simulazione di Sistemi

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Table of contents

Introduction
Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

Simulazione di Sistemi



Introduction

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

Simulazione di Sistemi

Standard *shapes* of information security:

- ▶ Confidentiality
- ▶ Integrity
- ▶ Availability

There is a new security that we want to obtain: **Anonymity**

Anonymity [...] means that the personal identity, or personally identifiable information of that person is not known.



Introduction

anonymity methods

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

3

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

There are a lot of anonymity driven software online, like *i2p*, *freenet* or *Tor*, we will talk about the last one because is the most used and expanded in the real world (2 million of client per day!).



Onion Routing

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

Simulazione di Sistemi

The onion routing model is a way to gain anonymity on the net:

- ▶ Provides anonymity
- ▶ Protects from sniffing

Introduced by David Goldschlag, Paul Syverson and Michael Reed in the 1999.

It recalls an onion because every step **peel** a layer.

Let us see an implementation.



Tor

The onion router

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

Simulazione di Sistemi

5

Overview

Tor is a group of volunteers that operates to defend anonymity online. The system is based on an interconnection of machines, called **routers**. It operates over the network level 4.

It operates as follow:



Tor

Tor workings

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

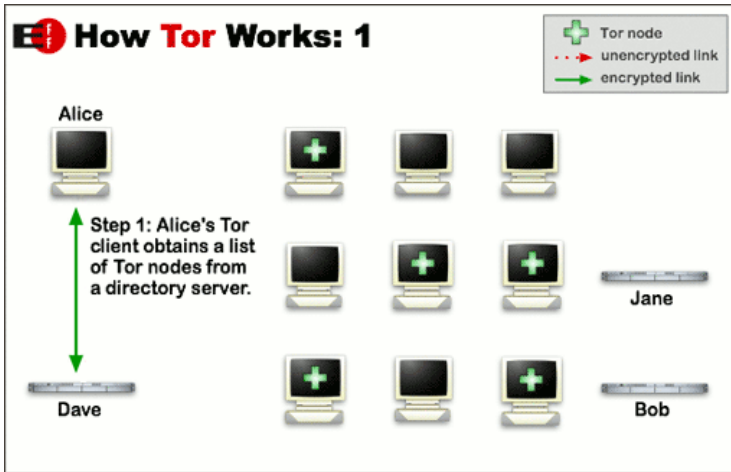
Analyzer

Empirical Results

Future Works

Simulazione di Sistemi

6





Tor

Tor workings

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

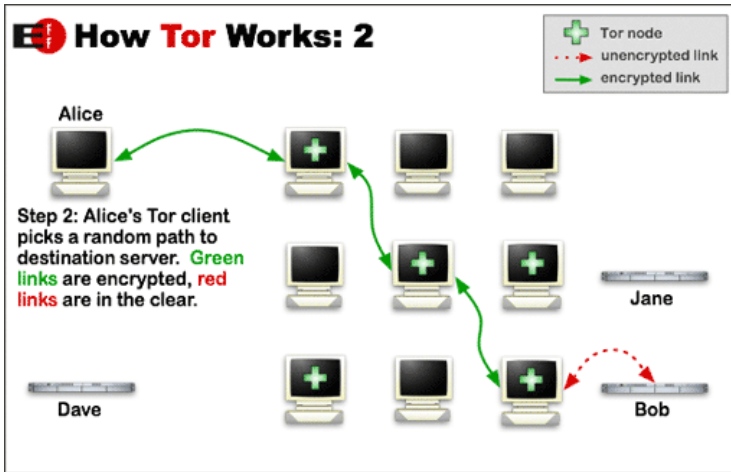
Analyzer

Empirical Results

Future Works

Simulazione di Sistemi

7





Tor

Tor workings

Time analysis Based Attacks Simulation in Tor Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

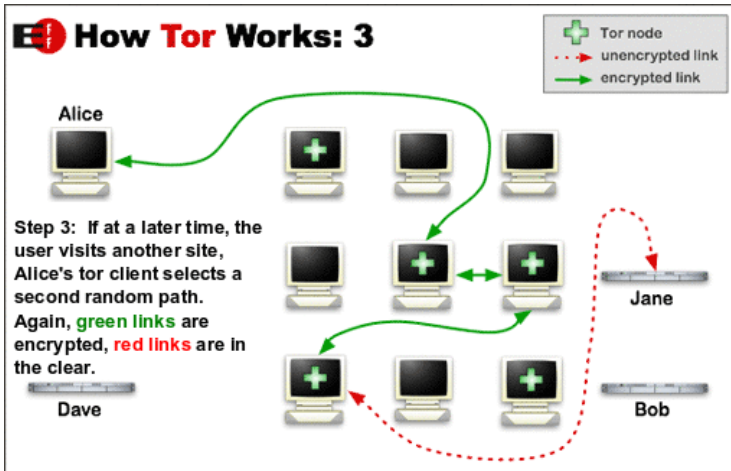
Analyzer

Empirical Results

Future Works

Simulazione di Sistemi

8





Tor

Tor encryption

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

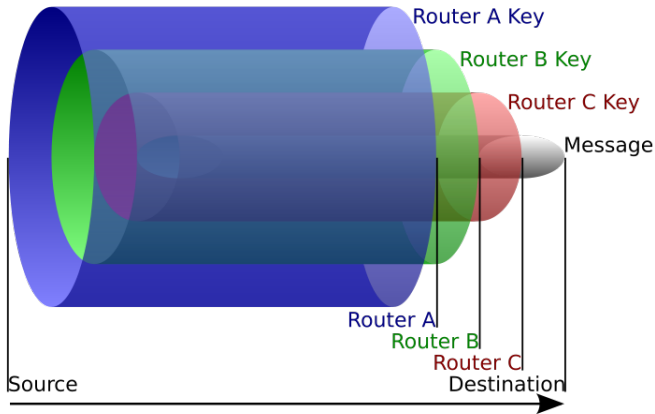
Simulation Handlers

Analyzer

Empirical Results

Future Works

9





Tor

Tor encryption

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

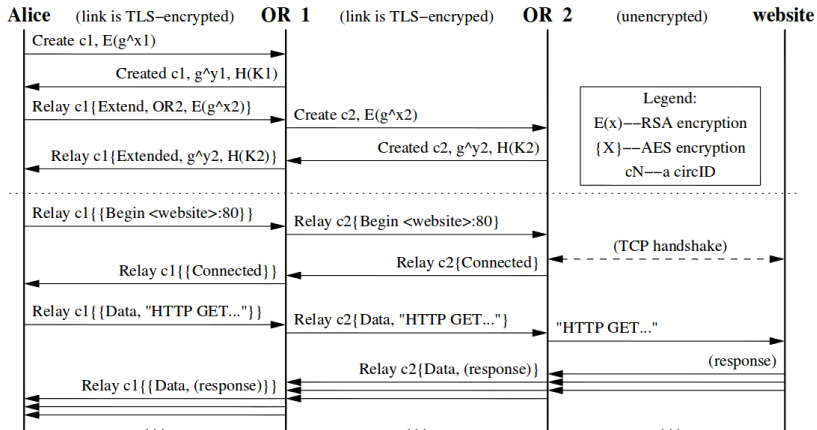
Analyzer

Empirical Results

Future Works

10

Simulazione di Sistemi





Attacks

A lot of attacks and vulnerabilities has been discovered for the system.

- ▶ Bad apple attack.
- ▶ Side channel attacks (tor bundle).
- ▶ Cypher attacks (Tor changed the cryptosystem a lot of time).
- ▶ *Time analysis based attacks*
- ▶ Sniper attack.
- ▶ Sybil attack.

11

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

Simulazione di Sistemi



Time analysis based attacks

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

12

“Tor does not provide protection against end-to-end timing attacks[...].”

We can place a tracker after the client node and another before the server node and check for the connection time to profile users and nodes (and later associate IP to users.)



Simulation

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

13

- ▶ From this idea we started our simulation work.
- ▶ But **OmNet++** doesn't have a reliable simulation model of Tor¹ and so **NS2/3**.
- ▶ We needed a simulation model for Tor.

¹And Tor is fully implemented in User-Space (over level 4)



Shadow

Introduction

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

- ▶ We used the **Shadow** simulator
- ▶ Developed by **Rob Jansen** (U.S. Naval Research Lab).

Users



14

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

Simulazione di Sistemi



Shadow

Simulator internals

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

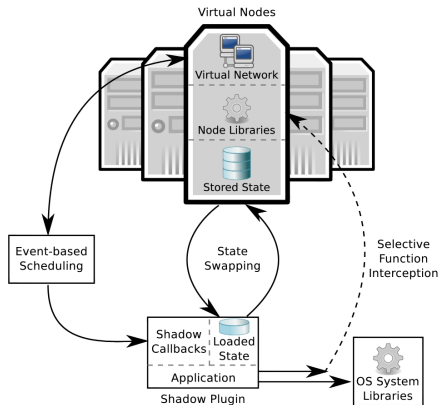
Future Works

Simulazione di Sistemi

The main feature of **shadow** is the capability of running real applications (like tor).

15

Shadow combines virtualization with simulation, it virtualize network stacks and act as an micro system hypervisor (partial virtualization).





Plug-ins

Shadow plugins

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

16

So we needed:

1. A client tracer shadow plug-in (proxy).
2. A server tracer shadow plug-in (proxy).
3. A logger plug-in.
4. A client plug-in (HTTP browser?).
5. A server plug-in (HTTP web-server?).

1,2 and 3 was not implemented by the shadow research team.



Plug-ins

Autosys plug-in

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

17

Data Analysis

Simulation Bunches

Simulation Handlers

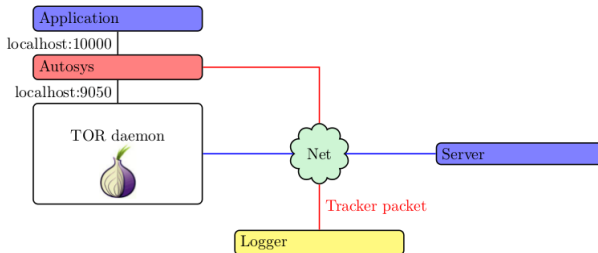
Analyzer

Empirical Results

Future Works

Simulazione di Sistemi

- ▶ Trace the SYN flag that pass trough Tor (on both ways)².
- ▶ Send a packet to the logger
< type(*Tracked_node*); Hostname(*Tracked_node*); timestamp >.



²A future work could be the trace of the SYN-ACK flag, to get the corresponding gap in the analysis part.



Plug-ins

Autosys plug-in

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

18

Can be implemented in a lot of different ways:

- ▶ As a sniffer installed on the routers which listen for every TCP SYN flag (the autonomous system).
 - ▶ But **Shadow** does not support Raw Sockets.
- ▶ We decided to implement that as a simple malware-like connection proxy installed on the client virtual node³.
- ▶ Otherwise the tracer can be installed on the guard relay (but we need to deal with re-association between traced clients and real clients because the path changes every 10 minutes).

FBI Spent \$775,000 on Hacking Team's Spying Tools

³A similar solution to the Hacking team one.



Plug-ins

Analyzer/Logger plug-in

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

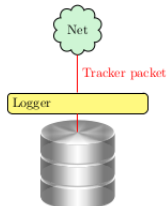
Analyzer

Empirical Results

Future Works

19

- ▶ After being captured by the sniffers the data must be stocked for late-processing.
- ▶ We used a public logger service that logs this informations.
- ▶ Based on UDP for lightness.
- ▶ (In a real-world scenario this entity would have some form of security and could be replicated/load balanced).





Plug-ins

Analyzer/Logger plug-in format

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

20

This plug-in so will save the data that it receives from the sniffers with a common format:

host_type; hostname; timestamp

- ▶ *host_type*: C or S if the tracked is a client or a server (the communication is going in or it exiting from Tor?).
- ▶ *hostname*: The hostname of the tracked (got by autosys).
- ▶ *timestamp*: The temporal reference of the connection (This will be used to compute distances and gaps).

This will be processed in the phase 2 to get the matches.



Plug-ins

SimpleTCP plug-in (Client)

We needed a simple client that sends his hostname on the network to compute the **matching accuracy** later.

- ▶ Do some connections to a fixed server.
- ▶ A future work should make it capable of multiple connections to multiple serves.
- ▶ This plug-in must have SOCKS5 capability to run over Tor.

21

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

Simulazione di Sistemi



Plug-ins

SimpleTCP plug-in (Server)

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

22

The server part, by opposite:

- ▶ Listen for some connections from the clients.
- ▶ Add a time stamp to the current received packet (correspondent host name).
- ▶ Save this data to a common file (per server).

This data will be used in the phase 2 to compute the **matching accuracy**.



Plug-ins

The big picture

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

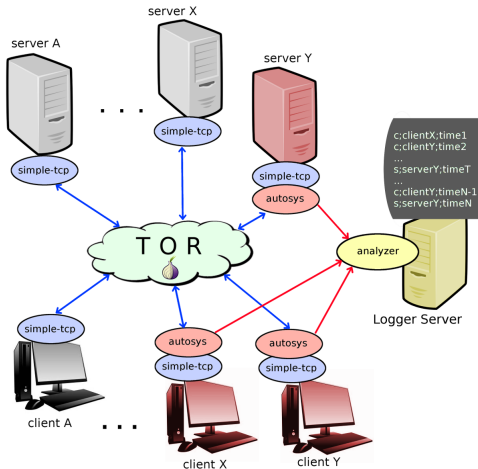
Simulation Handlers

Analyzer

Empirical Results

Future Works

23





Data Analysis

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

24

- ▶ Simulation Bunches
- ▶ Simulation Handlers
- ▶ Analyzer
- ▶ Empirical Results



Simulation Bunches

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

25

1. Traced clients fixed to the 100% and increasing traced servers at each macro bunch run ($0\% \rightarrow 100\%$).
2. Traced servers fixed to the 100% and increasing traced clients at each macro bunch run ($0\% \rightarrow 100\%$).
3. Increasing both traced clients and traced servers (traced portion) at each simulation ($0\% \rightarrow 100\%$).



Simulation Bunches

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

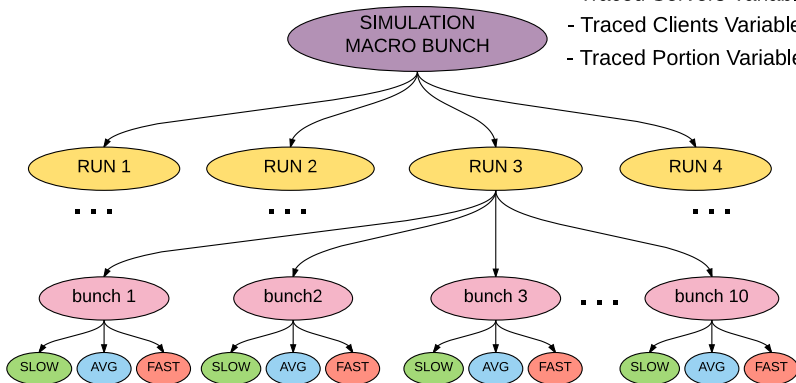
Empirical Results

Future Works

26

Simulazione di Sistemi

- Traced Servers Variable
- Traced Clients Variable
- Traced Portion Variable





Simulation Handlers

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

27

- ▶ Netbuilder
- ▶ Launcher



Netbuilder

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

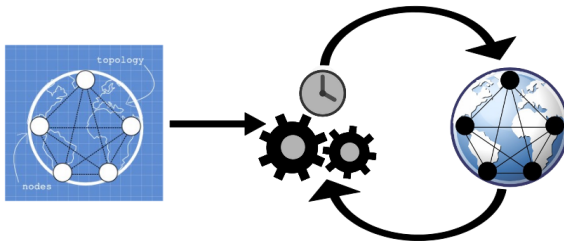
Analyzer

Empirical Results

Future Works

28

Generates an XML file that describes the network





Netbuilder

Configuration

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

29

Allow the network configuration through:

- ▶ The number of TOR exit nodes in the simulation.
- ▶ The number of TOR 4authorities⁴ nodes in the simulation.
- ▶ The number of clients (simpletcp) of the simulation.
- ▶ The number of servers (simpletcp) of the simulation.
- ▶ The percentage of clients tracked by an autosys plug-in.
- ▶ The percentage of servers tracked by an autosys plug-in.
- ▶ The density of the network-requests.

⁴A 4 Authority node is simply the database that keep track of the state of the TOR network and the list of the TOR relays/exit-nodes



Netbuilder

Densities

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

30

The connection densities are the sleep time thresholds between each client connection request:

- ▶ Slow: 800 (mean) - 2000 (mean) milliseconds
- ▶ Average: 80 (mean) - 1000 (mean) milliseconds
- ▶ Fast: 20 (mean) - 100 (mean) milliseconds



Launcher

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

31

Simulazione di Sistemi

Algorithm 2 Launcher script

```
for (simulation_run  $\leftarrow$  1; simulation_run  $\leq$  steps; simulation_run++) do
2:   for (sim_id  $\leftarrow$  1; sim_id  $\leq$  simulations_per_step; sim_id++) do
       for all density in (slow, fast, average) do
4:       if The client trace percentage is not fixed then
           client_trace_value  $\leftarrow$  sim_id/simulations_per_step
6:       end if
       if The server trace percentage is not fixed then
8:       server_trace_value  $\leftarrow$  sim_id/simulations_per_step
       end if
10:      if A configuration is present for  $\langle \textit{sim\_id}, \textit{density} \rangle$  And the percentages are fixed then
           Use the previous configuration
12:      else
           Generate a new configuration with net-builder
14:      end if
           Launch the Shadow Simulator with the appropriate configuration.
16:      end for
       end for
18: end for
```



Analyzer

Log file

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

32

Simulazione di Sistemi

...	(*)
c;client10;1420000000	s;server7;1421023888
s;server7;1420008031	s;server2;1421156205
c;client6;1420005867	c;client8;1421160529
s;server9;1420146660	s;server3;1421318345
s;server6;1420205384	s;server0;1421332488
s;server8;1420252482	c;client7;1421487295
c;client0;1420680882	c;client4;1421634744
c;client1;1421017740	s;server9;1421726485
(*)	...



Analyzer

Scan

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

- ▶ For each client connection request $creq$, it looks for candidate server acceptances
- ▶ Nested loop “temporally” limited between thr_{MIN} (100ms) thr_{MAX} (6sec)

Time distance

33

Let $\Delta_t(creq, s)$ be the time distance between a $creq$ time-stamp and a server candidate acceptance s time-stamp.



Analyzer

Scan

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

34

$\Delta_t < thr_{MIN} \rightarrow$

...

c;client10;1420000000
s;server7;1420008031
c;client6;1420005867
s;server9;1420146660
s;server6;1420205384
s;server8;1420252482
c;client0;1420680882
c;client1;1421017740
(*)

(*)

s;server7;1421023888
s;server2;1421156205
c;client8;1421160529
s;server3;1421318345
s;server0;1421332488
c;client7;1421487295
c;client4;1421634744
s;server9;1421726485
...

← already considered



Analyzer

Matching Probability

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

35

The likelihood for a server acceptance to be related to a client request can be related to their time distance.

$pmatch$

$$pmatch(creq, s) = 1 - \frac{\Delta_t(creq, s) - thr_{min}}{thr_{max} - thr_{min}} \quad (1)$$



Analyzer

Matching Probability

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

36

Simulazione di Sistemi

candidate	pmatch
server9	0.992
server6	0.982
server8	0.975
server7	0.846
server2	0.823
server3	0.769
server0	0.794
...	...

The *pmatch* is higher when the server connection is closer to thr_{min} .



Analyzer

Time Gap

Time analysis
Based Attacks
Simulation in Tor
Networks.

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Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

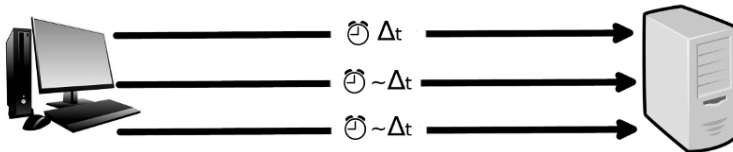
Simulation Handlers

Analyzer

Empirical Results

Future Works

Simulazione di Sistemi



Acceptance Delay Correlation

37

If a server receives a connection request from a client after a certain time Δ_t , that server will likely receive again another connection from the same client after a time that is close to Δ_t if the Tor communication path is the same as before



Analyzer

Time Gap Average

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

38

As the $pmatch$ is defined as the Δ_t normalization, let us define the *gap* average of a server s marked as candidate for a client c

$$gap_{AVG}(c, s) = \frac{\sum_{i=0}^{N(c,s)} |pmatch(creq_{i+1}, s) - pmatch(creq_i, s)|}{N(c, s)} \quad (2)$$

where $N(c, s)$ is the number of c connection requests that have been likely accepted from s .



Analyzer

Score

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

39

The *score* gained by a server s marked as candidate for a client c

$$score(c, s) = \frac{\sum_{i=0}^{N(c,s)} pmatch(creq_i, s)}{gap_{AVG}(c, s) + 1} \quad (3)$$



Analyzer

Best Candidate

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

Simulazione di Sistemi

client633		client637		client349	
candidate	score	candidate	score	candidate	score
<u>server8</u>	9.44	<u>server3</u>	59.17	<u>server0</u>	14.86
server0	7.01	server2	15.14	server1	13.81
server2	6.88	server8	13.96	server5	11.94
server5	6.83	server5	8.33	server2	11.20
...

40

Best Candidate

The server candidate with the **highest score** is the best candidate for a certain client.



Analyzer

Real Data

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

41

- ▶ How much are the estimated results close to the real ones?
- ▶ Use of real connections logged by the simple-tcp applications.
- ▶ Matched accuracy estimation
- ▶ Matched portion estimation



Analyzer

Matched Accuracy

Time analysis
Based Attacks
Simulation in Tor
Networks.

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Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

42

For each client check if the best candidate is the real server that accepted its connections and mark it as **matched**.

If so calculate the client accuracy as the distance between the number of estimated connections N and the number of real connections M :

$$accuracy_c \leftarrow \frac{MIN(M, N)}{MAX(M, N)} \quad (4)$$

The *matched accuracy* is the average of matched client accuracies.



Analyzer

Matched Portion

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

43

The matched portion indicates how many traced clients found their real server:

$$matched_portion = \frac{matched_clients}{traced_clients} \quad (5)$$



Matched Portion

Servers traced portion augmenting

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

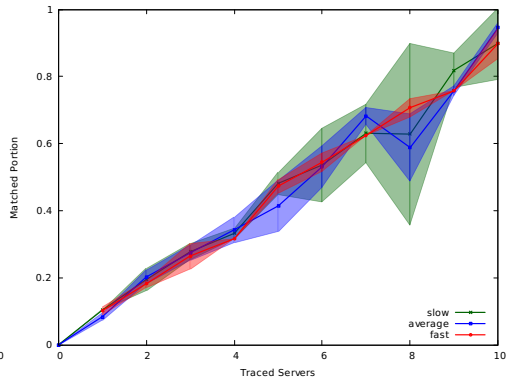
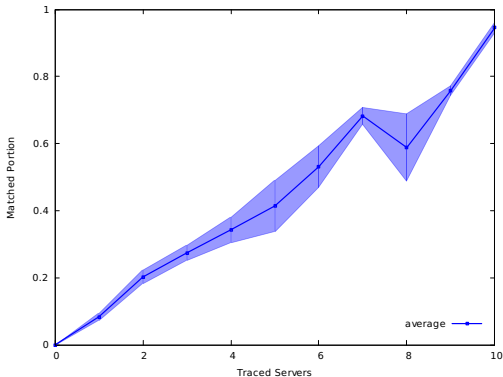
Analyzer

Empirical Results

Future Works

44

Simulazione di Sistemi





Matched Portion

Clients traced portion augmenting

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

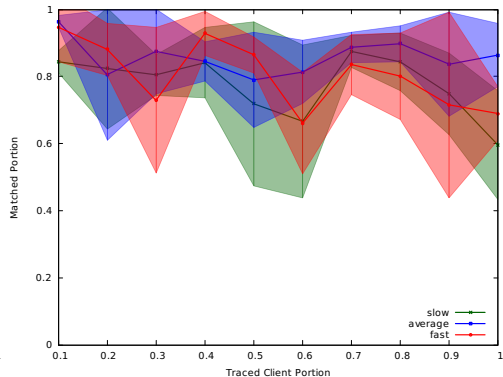
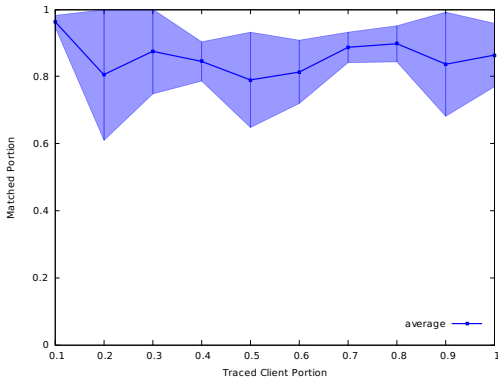
Simulation Handlers

Analyzer

Empirical Results

Future Works

45





Matched Portion

Details

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

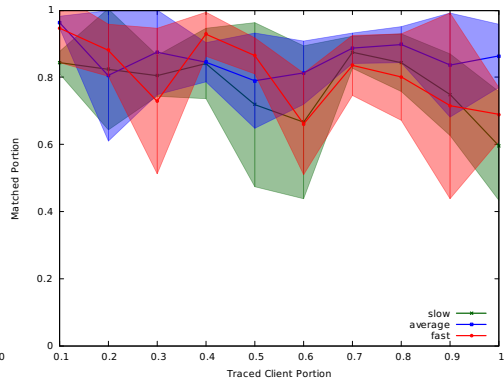
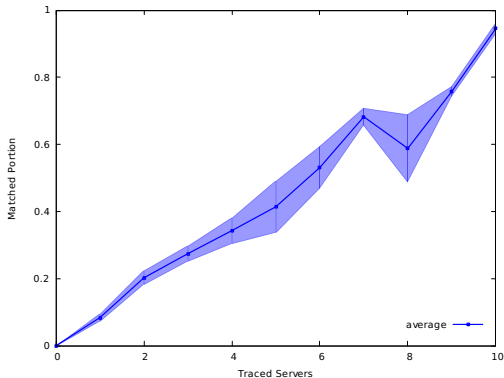
Empirical Results

Future Works

46

Simulazione di Sistemi

We can see that the matched portion tends to be constant around 80% (as with an high portion of traced servers).





Matched Portion

Considerations

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

47

Future Works

Simulazione di Sistemi

- ▶ This behaviour was expected because the analysis is conducted from the clients side.
- ▶ Increasing the connection density the function trend seems to be more precise.
- ▶ Linearly dependent by the number of traced servers.



Matched Accuracy

Servers traced portion augmenting

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

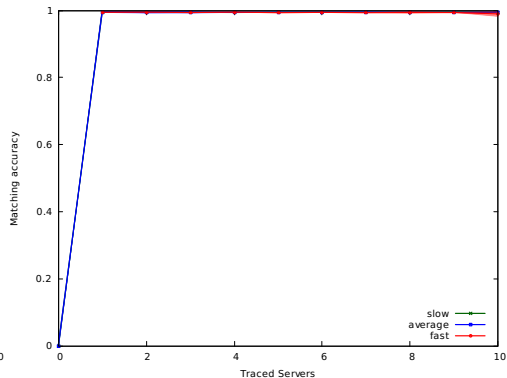
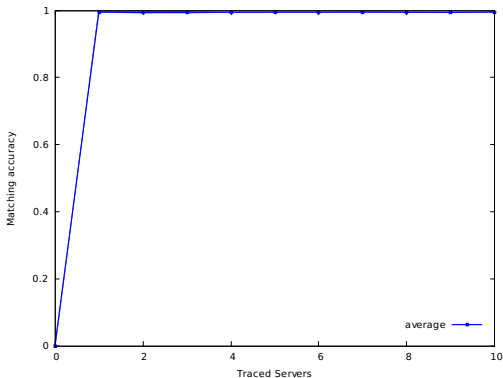
Simulation Handlers

Analyzer

Empirical Results

Future Works

48





Matched Accuracy

Clients traced portion

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

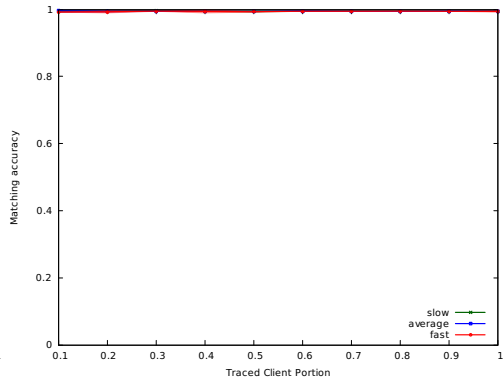
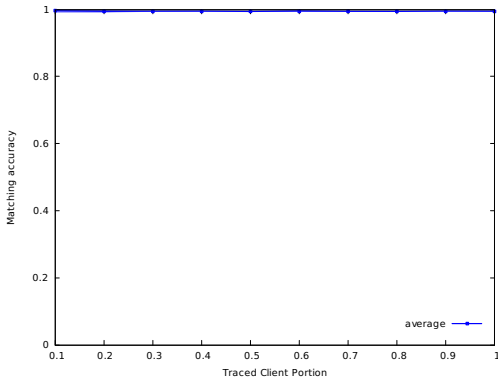
Simulation Handlers

Analyzer

Empirical Results

Future Works

49





Considered values

Both portions

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

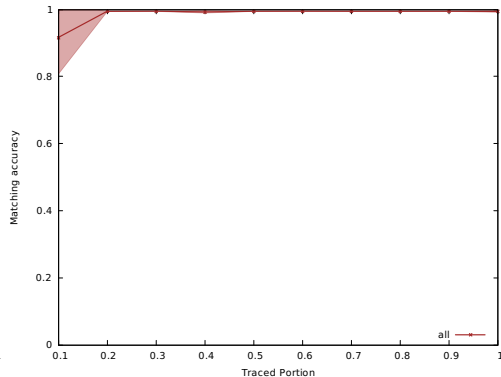
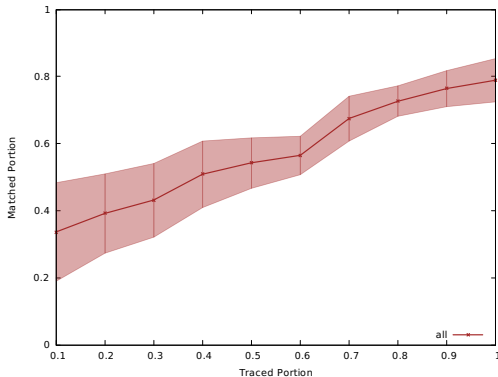
Simulation Handlers

Analyzer

Empirical Results

Future Works

50





Considered values

Considerations

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

51

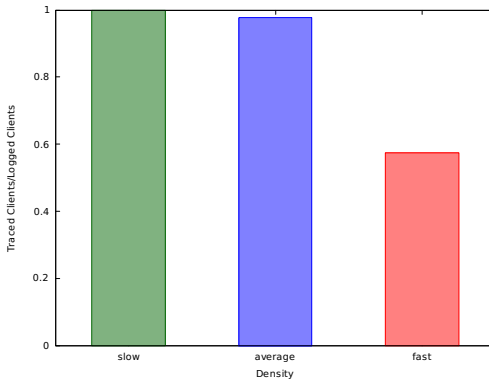
Future Works

- ▶ Most realistic scenario
- ▶ Respect the avg sum of the other two experiments
- ▶ An attacker should be interested in trace as much Tor network nodes as possible.



Communication density

Does not seems to highly interfere with the matched portion.





How to distinguish correct guesses?

Time analysis
Based Attacks
Simulation in Tor
Networks.

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Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

53

Future Works

- ▶ An attacker can, so far, get some maps between servers and clients.
- ▶ Let us see a 4 dimensional graph.



Correct guesses spatial distribution

Time analysis
Based Attacks
Simulation in Tor
Networks.

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Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

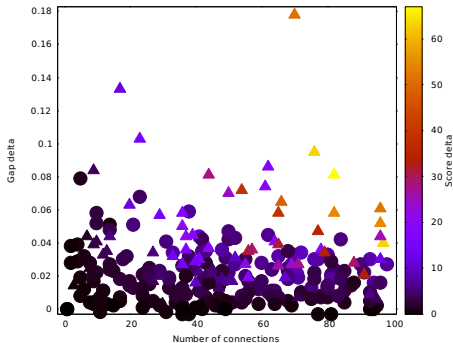
Analyzer

Empirical Results

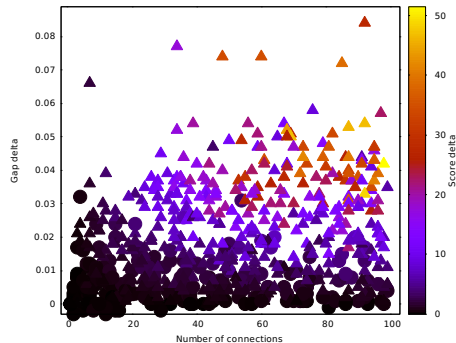
Future Works

54

Simulazione di Sistemi



40% of traced portion



90% of traced portion



Correct guesses spatial distribution

Considerations

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

55

Future Works

Simulazione di Sistemi

- ▶ As we can see the correctly guessed servers (triangles) take place in the upper-right section in a yellowish color.
- ▶ We can choose some parameters to get the "Trusted matching set".
- ▶ An attacker so can blindly select some matchings.



Future works

Time analysis
Based Attacks
Simulation in Tor
Networks.

Davide Berardi,
Matteo Martelli

Introduction

Attacks

Simulation

Shadow

Plug-ins

Data Analysis

Simulation Bunches

Simulation Handlers

Analyzer

Empirical Results

Future Works

56

- ▶ A simulation with the alternative new-born Tor client "Astoria".
- ▶ An analysis of the i2p network model and the freenet network model.
- ▶ An analysis for some modification based on the paper "Mix network model".
- ▶ The modification to the **simple-tcp** plug-in to make it capable of connecting to multiple servers in single instance.
- ▶ Go on with the score delta driven analysis.
- ▶ Repeat the experiment with a bigger Tor network.

Thank you for your attention.

