

Introduction

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TA on Tor

Tor recap

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Traffic Analysis

“Privacy-Enhancing Technologies” Course Talk

Willard Rafnsson

Department of Computer Science and Engineering
Chalmers University of Technology
Gothenburg, Sweden

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Outline

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Military

Deducing **information** from communication patterns.

- ▶ Frequent communication: Planning
 - ▶ Between same points: Chain of command
- ▶ Morse “hand”
- ▶ WWII: HMS Glorious

Traffic Analysis

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Military

Deducing **information** from communication patterns.

- ▶ Frequent communication: Planning
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- ▶ WWII: HMS Glorious

Low-quality compared to cryptanalysis, but easy/cheap to extract/process, and hard/expensive to counter.

- ▶ TA to select target for cryptanalysis.

Computer Security Setting

Traffic Analysis

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The use of **traffic data**, that is,

- ▶ transmission-time,
- ▶ length, and
- ▶ direction

of network packets to/from victim,
to extract information sensitive to the victim.

Note: **not content** of packets (encrypted?)

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Typical objective: *Deanonymization*.

Example

“Timing Analysis of Keystrokes and Timing Attacks on SSH” (Song et al.)

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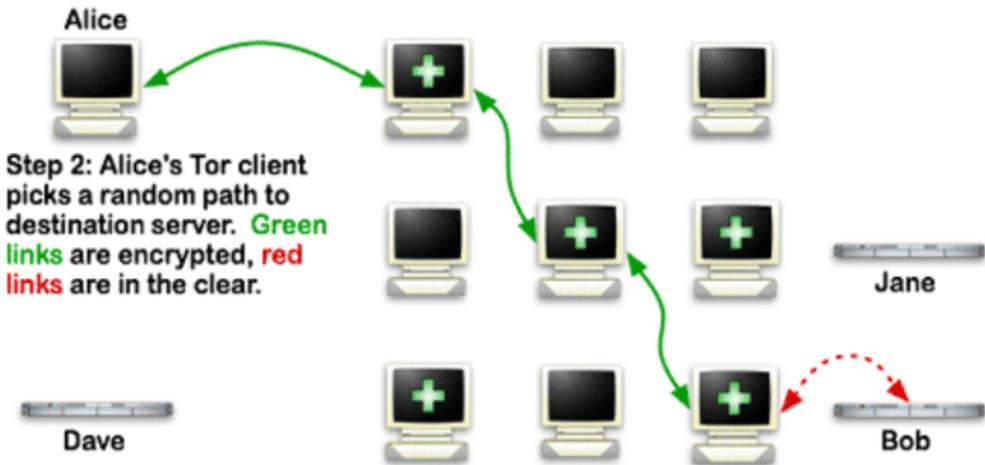
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Recall How Tor Works

How Tor Works: 2



Non-Global Attacker

Can

- ▶ observe,
- ▶ modify, and
- ▶ control

a fraction of the Tor network.

Traffic Analysis

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Non-Global Attacker

Can

- ▶ observe,
- ▶ modify, and
- ▶ control

a fraction of the Tor network.

Attacker can extract Tor connection path information.

Objective

Infer the nodes a stream goes through.

- ▶ know which OR stream begins at,
 - ▶ reduces anonymity.
- ▶ trace unrelated streams to same initiator.

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Objective

Infer the nodes a stream goes through.

- ▶ know which OR stream begins at,
 - ▶ reduces anonymity.
- ▶ trace unrelated streams to same initiator.

Why possible:

- ▶ Each OR processes its streams in a round-robin fashion.
 - ▶ empty streams skipped to save time (low-latency demand)
- ▶ Adding a stream to a OR delays processing of existing streams at OR slightly.

The Attack

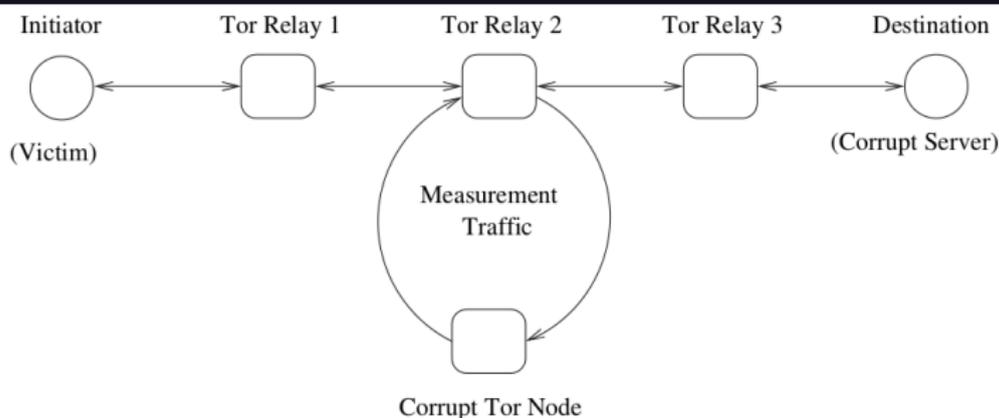


Figure 1. The attack setup

It's Cheap

To pull this off, you must be

- ▶ at the end of stream
 - ▶ compromised web server (trick victim)
 - ▶ man-in-the-middle (evil web hosting),
 - ▶ an exit node?
- ▶ able to probe all Tor nodes
 - ▶ ≥ 1 low-latency machine (same as end?)

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How it scales:

- ▶ N probe streams required.
- ▶ Attack cost: $O(N)$.

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How it scales:

- ▶ N probe streams required.
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For more:

“Low-Cost Traffic Analysis of Tor”, S&P 2005

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Classifier

Recall what is observed:

- ▶ transmission-time,
- ▶ length, and
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of network packets.

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Classifier

Recall what is observed:

- ▶ transmission-time,
- ▶ length, and
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of network packets.

Classifier: function which, given (*length, direction*) (of a packet p), returns a (guess of the) destination for p .

Classifier

Recall what is observed:

- ▶ transmission-time,
- ▶ length, and
- ▶ direction

of network packets.

Classifier: function which, given (*length, direction*) (of a packet p), returns a (guess of the) destination for p .

Typically **machine learning** algorithms, trained on a large data set of traffic.

Observables

Recall what is observed:

- ▶ transmission-time, ← avoid change (latency)
- ▶ length, and
- ▶ direction ← cannot change

of network packets.

Padding seems like a good idea.

Padding

SSH/TLS/IPSec padding:

- ▶ Session Random 255 byte padding
- ▶ Packet Random 255 byte padding

Other:

- ▶ Linear (nearest $k128$)
- ▶ Exponential (nearest 2^k)
- ▶ Mice-Elephants
- ▶ MTU
- ▶ Packet Rnd. MTU ($\text{rnd} \{0, 8, \dots, M - l\}$)

C: Too much overhead ($\geq 40\%$) to be practical.

#Packets Obfuscation

Direct target sampling Define/derive distrib.

D_{AB} over packets from A to B . When A sends a packet p of length i to B , instead,

- ▶ sample D_{AB} for smallest #lengths $i_1 \dots i_k$ s.t. $\sum_{j=1}^{k-1} i_j < i \leq \sum_{j=1}^k i_j$,
- ▶ send first i_1 bytes of p as a packet, then next i_2, \dots (pad last to i_k).

Traffic morphing (same idea, more complex to understand, more efficient in practise)

Too much overhead (40-80%) to be practical.

Transmission Time & Bandwidth

Not explored in literature:

- ▶ Total session time,
- ▶ Total bandwidth of data transmitted each direction,
- ▶ Transmission time of each packet
 - ▶ “burstiness” of packets.

Too much buffering / junk to be practical.

All Known Countermeasures Fail

Individual packet lengths need not be considered for high-accuracy classification

- ▶ padding & packet splitting ineffective countermeasure
 - ▶ fixed padding does not change bandwidth substantially
 - ▶ random padding “averages” out
 - ▶ burst information very informative
- ▶ best classifier still $> 80\%$ accuracy with privacy set size 128.

Simple Classifiers Are Accurate

Best classifier only marginally better than a naive Bayes classifier which *only* considers

- ▶ total session time,
- ▶ per-direction per-website bandwidth, and
- ▶ burst patterns.

VNG++ classifier developed.

VNG++ Counterme. Impractical

Buffered Fixed Length Obfuscator (BuFLO):

- ▶ fixed-interval send of
- ▶ fixed-length packets for a
- ▶ fixed minimum transmission time.

With well-configured parameters, best classifier down to 5.1% accuracy for privacy set size 128 (random guess is $\frac{1}{128} = 0.78\%$)

For more:

“Peek-a-Boo, I Still See You: Why Efficient Traffic Analysis Countermeasures Fail”, S&P 201[12]