

# | Side-Channel Security

## Chapter 7: Network Side Channels

Stefan Gast

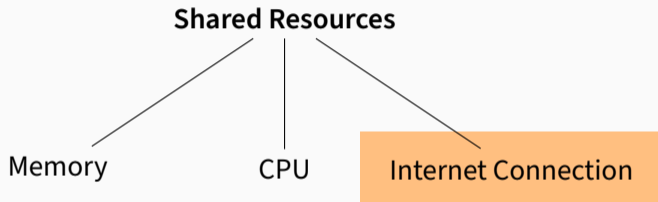
2025-04-03

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# Introduction

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# What to Attack?

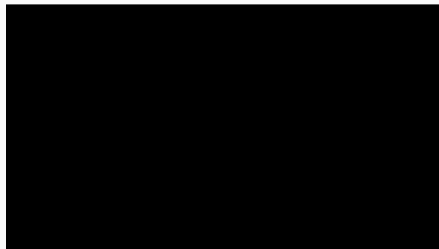
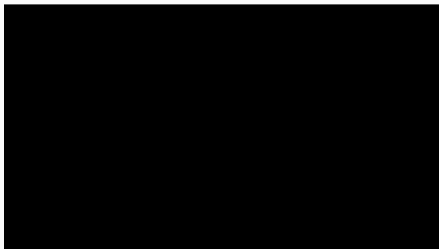


**Demo: Network Traffic Depends on Activity**

Every website causes a characteristic traffic pattern – a **fingerprint**:

- Hintz, 2003 [Hin03]: asset transfer sizes
  - Panchenko et al. , 2011 [Pan+11]: packet sizes, directions, order
  - Rimmer et al. , 2017 [Rim+17]: traffic shape (packet sizes, directions, timings), CNN classifier
  - ...
- attacker-in-the-middle, mostly used against privacy-enhancing tunnels

Which video segment uses more bandwidth?



<https://www.youtube.com/watch?v=LNI8rnxxVvQ>

# Video Fingerprinting

- Dynamic Adaptive Streaming over HTTP (DASH) [ISO22]
  - usually encrypted
  - split video into segments with a few seconds duration
  - send segments on demand
  - segment durations and sizes depend on content
- fingerprint!

- Reed and Kranch, 2017 [RK17]: Netflix
  - Schuster et al. , 2017 [SST17]: YouTube, Netflix, Amazon, Vimeo
  - Gu et al. , 2018 [Gu+18]: self-hosted DASH server
  - ...
- attacker-in-the-middle or with JavaScript



## Other Traffic Analysis Attacks

- SSH keystroke timings [SWT01]
- deanonymization of Tor users [RSG98; AYR15; Wan+11]
- language [Wri+07] and phonemes [Whi+11] of VoIP calls
- other privacy-critical information [Che+10; LM18]

# SnailLoad: Remote Traffic Analysis via TCP [Gas+24]

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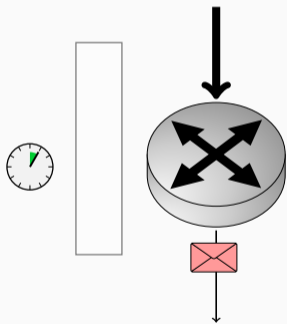
**Some of you probably know the  
effect...**

- DSL, Fiber, LTE, 5G: different throughput
- backbone connection **has orders of magnitude higher throughput**
- buffering before last mile is necessary!

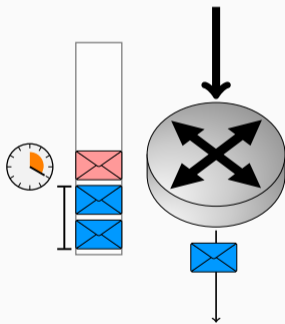




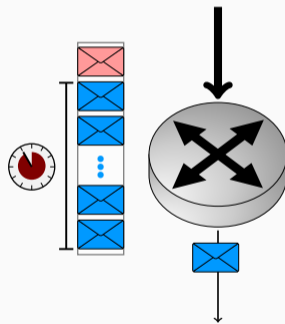




**Figure 1:** Connection idle



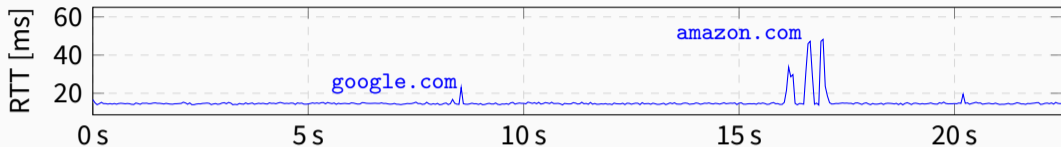
**Figure 2:** Connection busy



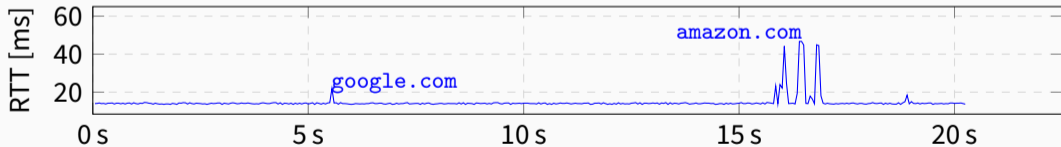
**Figure 3:** Bufferbloat



# Network Activity Causes Latency Spikes



**Figure 4:** Same machine pinging 8.8.8.8



**Figure 5:** Different machine sharing the same internet connection pinging 8.8.8.8

## Idle and Busy Round-Trip-Times

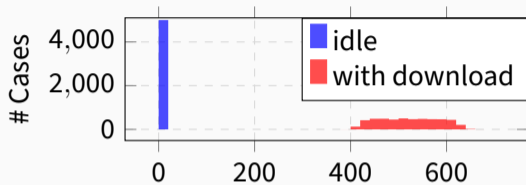


Figure 6: RTT [ms], ADSL-1, 50 Mbit/s

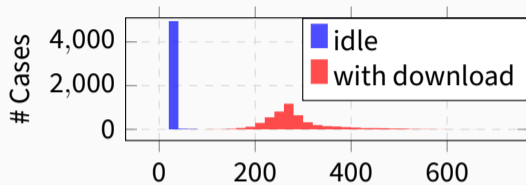


Figure 7: RTT [ms], LTE, 75 Mbit/s

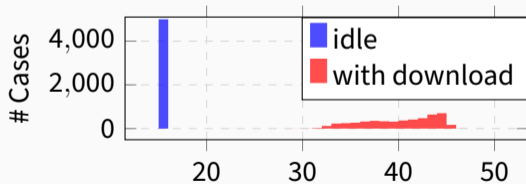


Figure 8: RTT [ms], FTTH-1, 80 Mbit/s

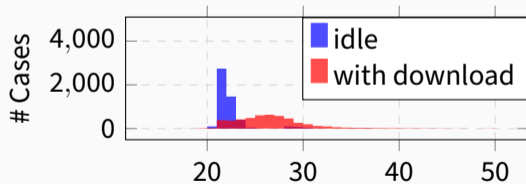
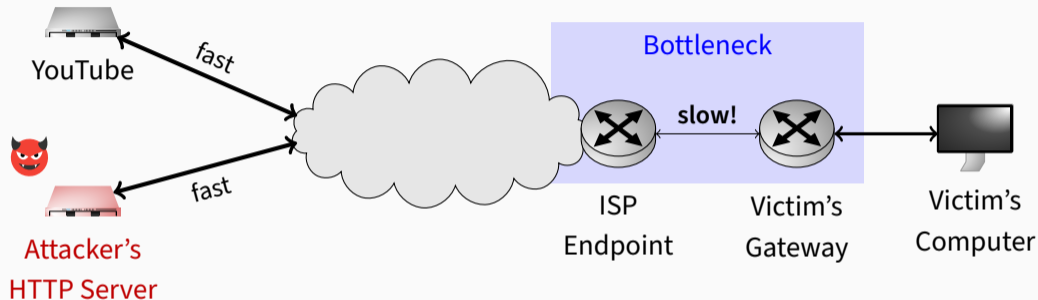


Figure 9: RTT [ms], Cable, 80 Mbit/s

## Attack Setup



- Various scenarios: Compromised websites, malicious ads, emails, and more
- Different ways attackers can exploit network traffic to perform attacks



## Polling the Server's Send Buffer To Measure RTTs

**begin**

```
acked ← false;
start ← get_current_time();
send(sock, b, 1, 0);
repeat
  | if ioctl(sock, SIOCOUTQ) = 0 then
  |   |   acked ← true;
  |   end
until acked;
end ← get_current_time();
return end - start;
```

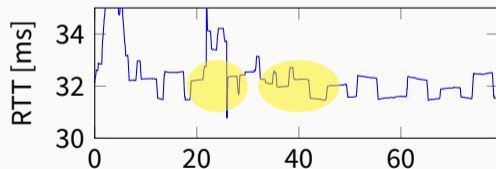
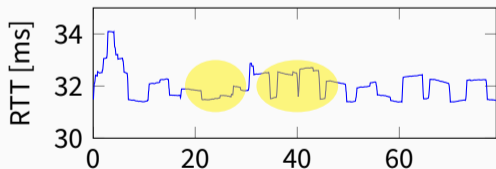
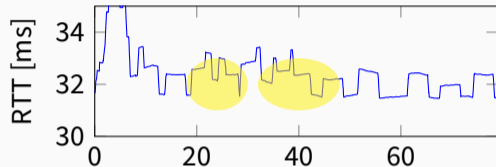
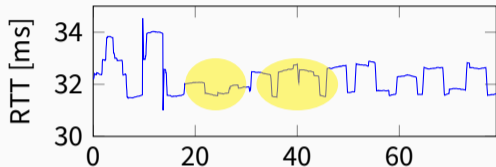
**end**



- use machine learning to analyze network traffic and infer user actions
- pre-process traces with an STFT
- KERAS (Tensorflow)

**Table 1:** CNN Parameters

Type	Parameters	Activation
Conv2D	filters=32, kernel size=[5,5], strides=[1,1]	ReLU
MaxPooling2D	pool size=[2,2], strides=[2,2]	-
Conv2D	filters=64, kernel size=[3,3], strides=[1,1]	ReLU
MaxPooling2D	pool size=[2,2], strides=[2,2]	-
Conv2D	filters=128, kernel size=[3,3], strides=[1,1]	ReLU
MaxPooling2D	pool size=[2,2], strides=[2,2]	-
Flatten	-	-
Dense	output size=1024	ReLU
Dense	output size=512	ReLU
Dense	output size=10	Softmax

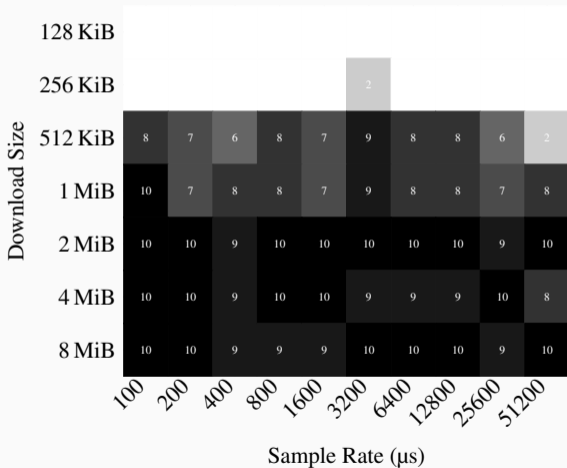


**Figure 10:** Video A, Time in seconds on x axis

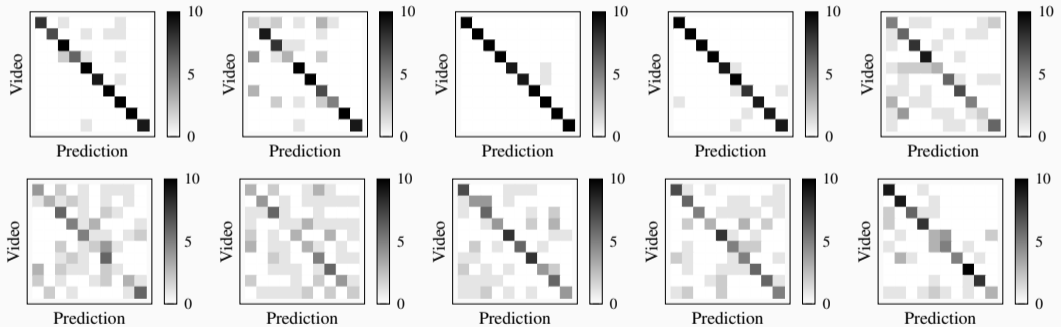
**Figure 11:** Video B, Time in seconds on x axis



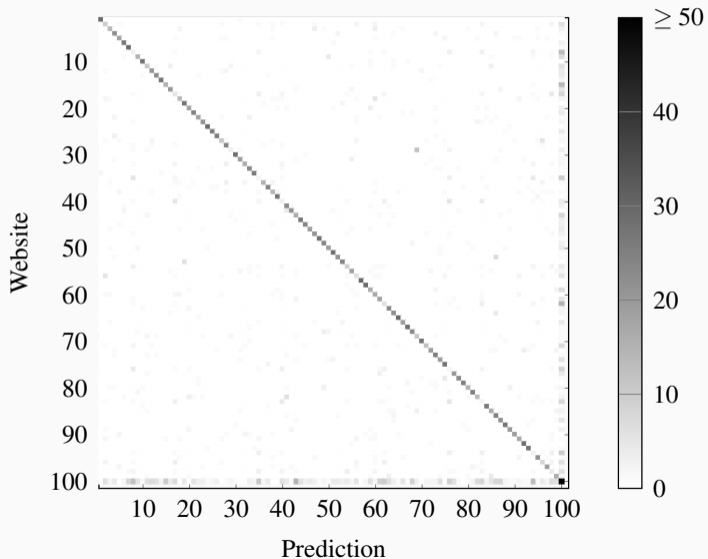
# How large does the website have to be?



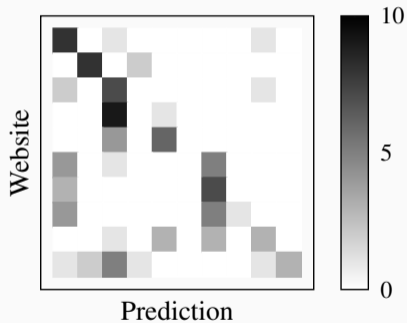
# Video Fingerprinting on 10 different connections



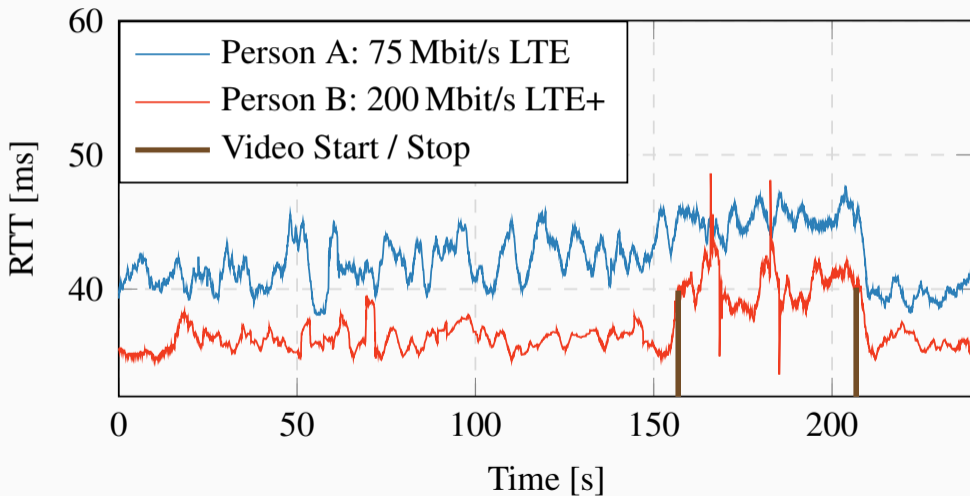
# Top-100 Open-World Website Fingerprinting



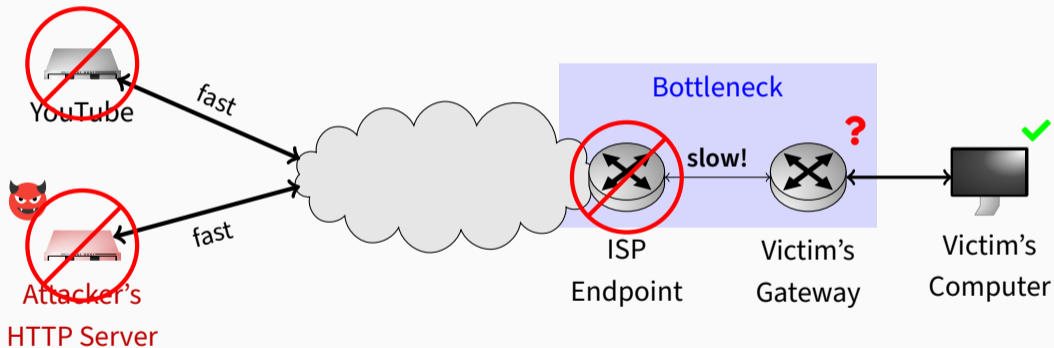
# Cross-Connection Website Fingerprinting



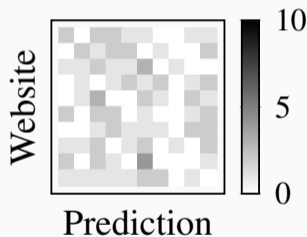
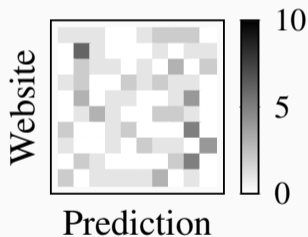
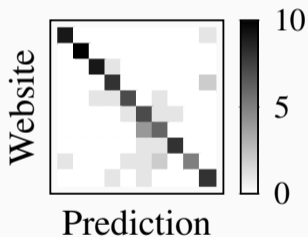
# Video Call Detection



# What about Mitigations?



# Impact of Noise on Website Fingerprinting



- SnailLoad is a generic problem of heterogenous networks (with different throughputs)
- Many “remote” attacks can now be transformed to truly remote attacks
- We disclosed to Google / YouTube
  - they investigated the issue for several weeks
  - concluded that it is a generic problem



## Take Aways

- Any connection to a remote server can obtain high-resolution traces of your activity
- Traces can leak websites and videos watched
- Throughput difference is the root cause → not trivial to fix
- Paper + Demo: <https://snailload.com>

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