



# Beauty And The Burst

Remote Identification of Encrypted Video Streams

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# Agenda

- Background
- MPEG-DASH Standard
- Attack Overview
- Deep Neural Networks
- Adversarial Models
  - On-Path Attack
  - Off-Path Attack

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**Why video traffic is so interesting ?**

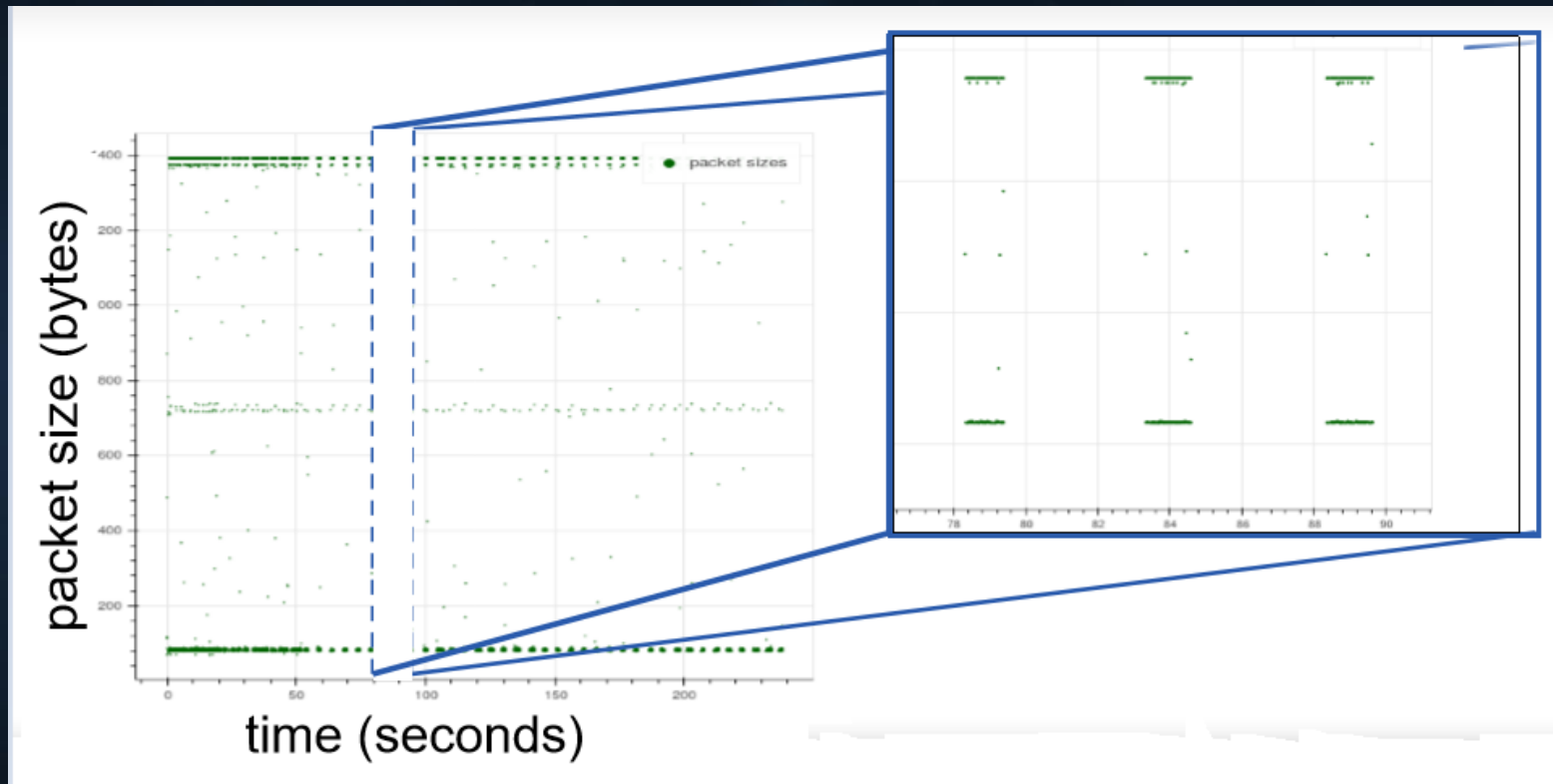
# Background

- Targeted marketing purposes
- Market characterization efforts
- Not everybody wants to volunteer this information about their habits
  - Video traffic is encrypted
  - HTTPS has been in wide deployment

# What still can be learned ?



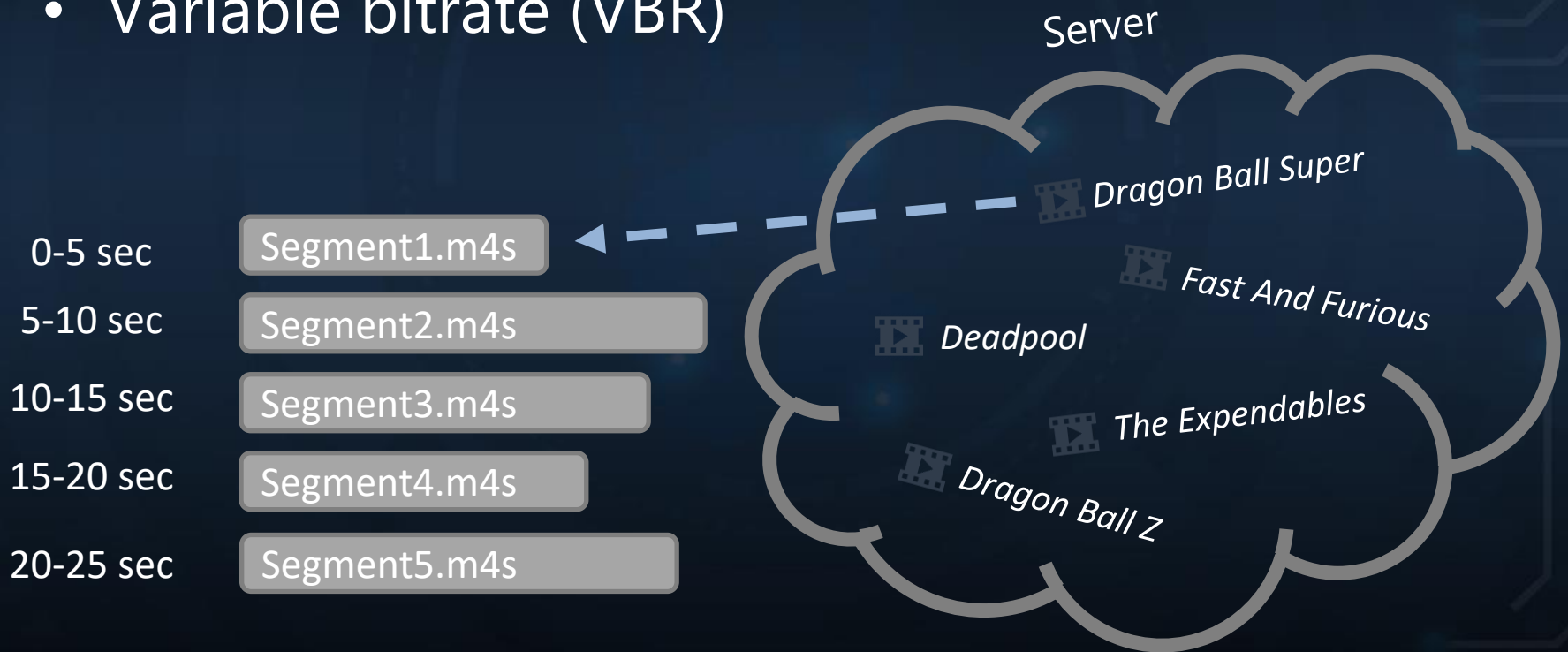
# Initial buffering & on/off bursts



**Where the bursts come from ?**

# MPEG-DASH standard

- Widely adopted by major streaming providers
  - Netflix, YouTube etc.
- Variable bitrate (VBR)





# MPEG-DASH standard

- Adaptive bitrate streaming over HTTP



# VBR Demo

## Iguana vs. Snakes

Scenery, movement, tension rising



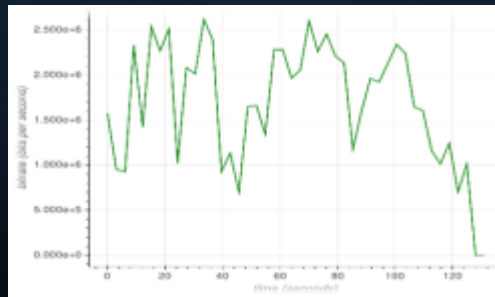
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# MPEG-DASH leak

Content



VBR pattern



Burst sizes



Segments

Segment1.m4s

Segment2.m4s

Segment3.m4s

...

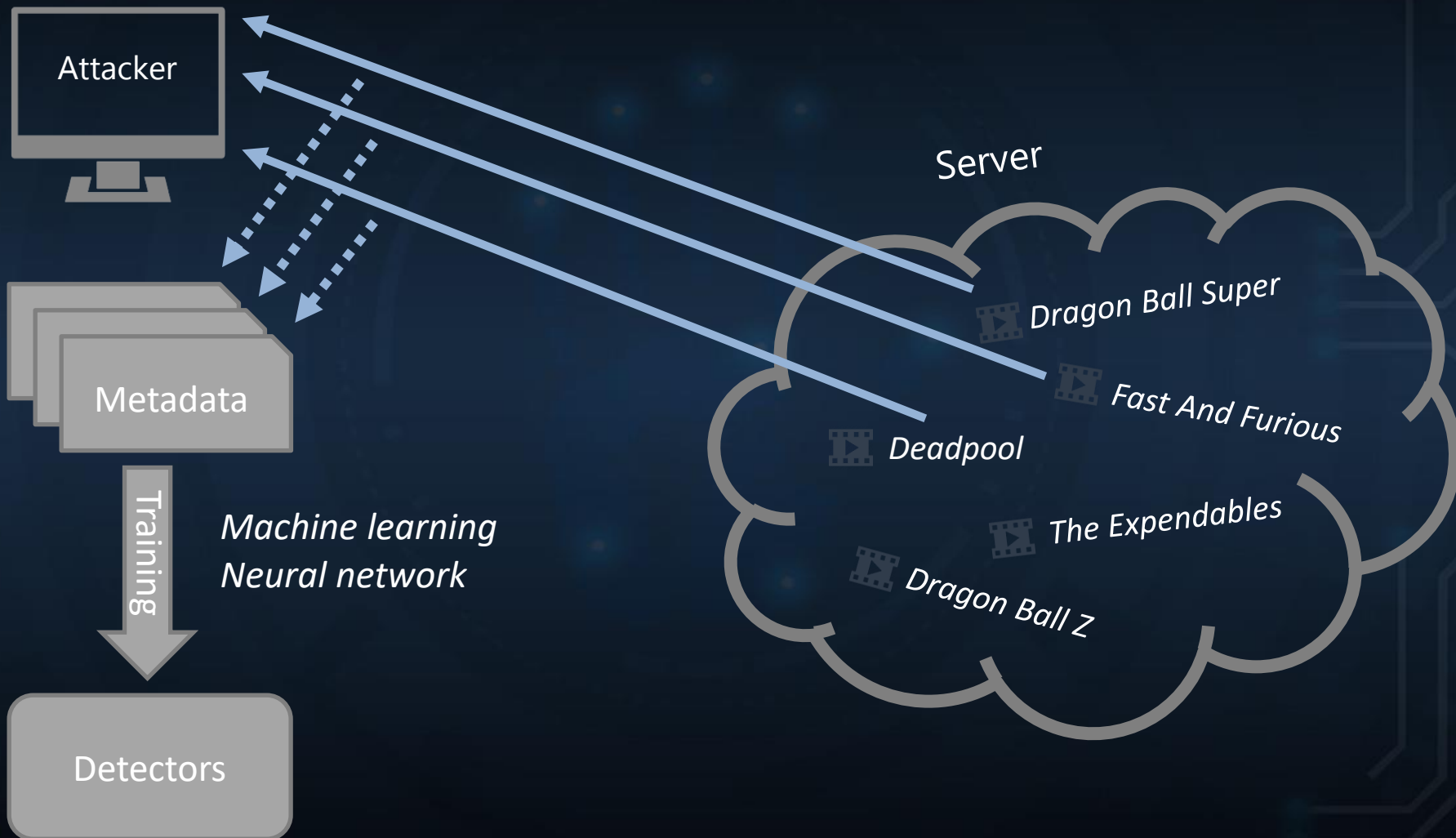
# From a leak to a fingerprint

- Does the pattern of burst (segment) sizes uniquely characterize a title?
- Empirically for the 3500 downloaded from YouTube 20% of them have a uniquely identifying pattern
- Can we learn a title's identifying pattern?
- We can learn a title's identifying pattern because of the pattern consistency

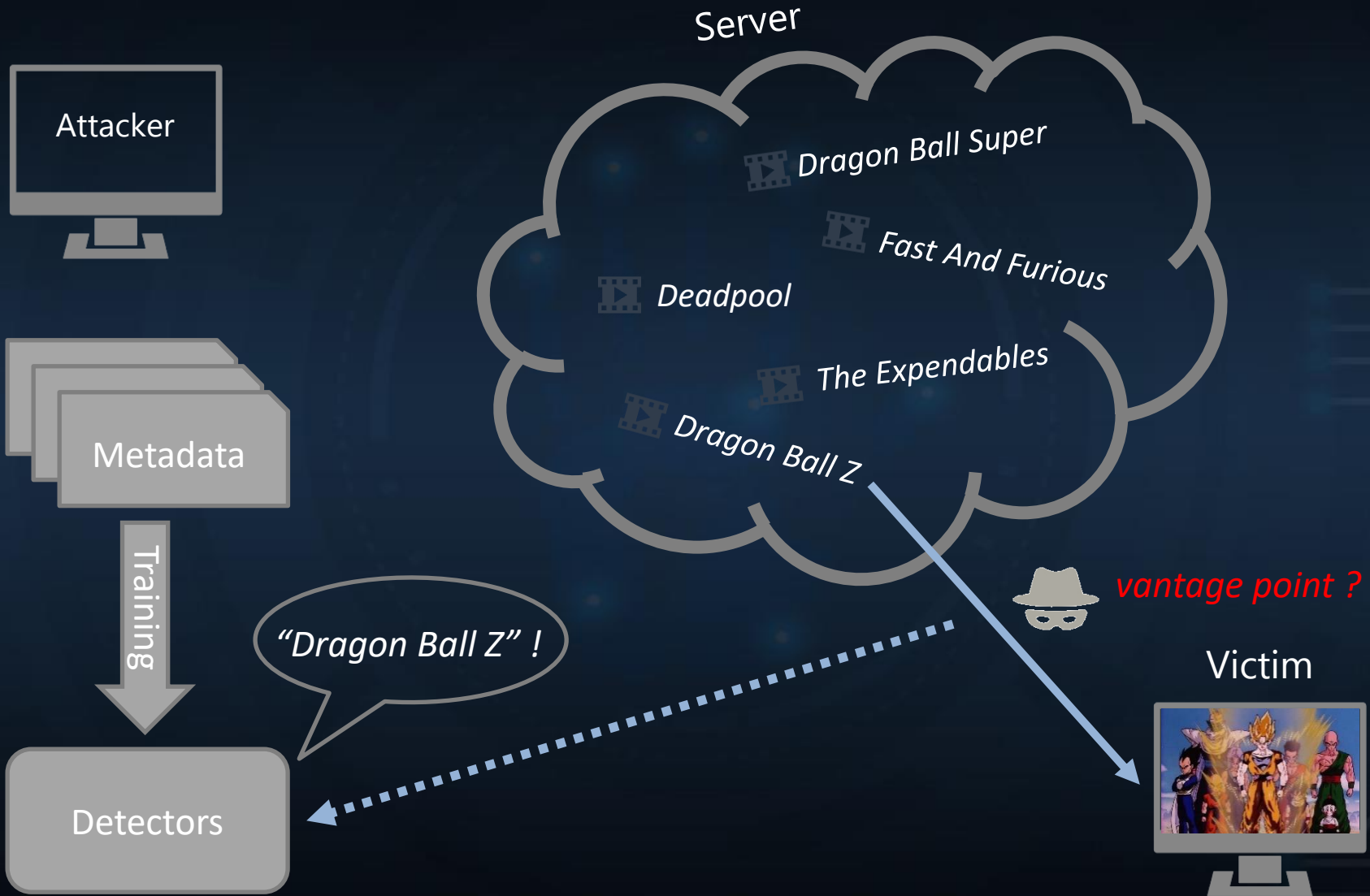


~20% of YouTube titles have fingerprints

# Attack overview



# Attack overview



# Deep Neural Networks

- Very good at learning high-level concepts that human can easily agree on but find it hard to formally express
- Can operate on noisy and coarse measurements
- Agnostic to protocol-specific attributes
- Can learn features other than burst patterns, e.g. arrival patterns of individual packets
- Can use multiple session representations, train on all at once

# Deep Neural Networks



100 titles  
98.5% accuracy



18 titles + 3500 sessions  
of different other titles  
99.5% accuracy



10 titles  
92.5% accuracy



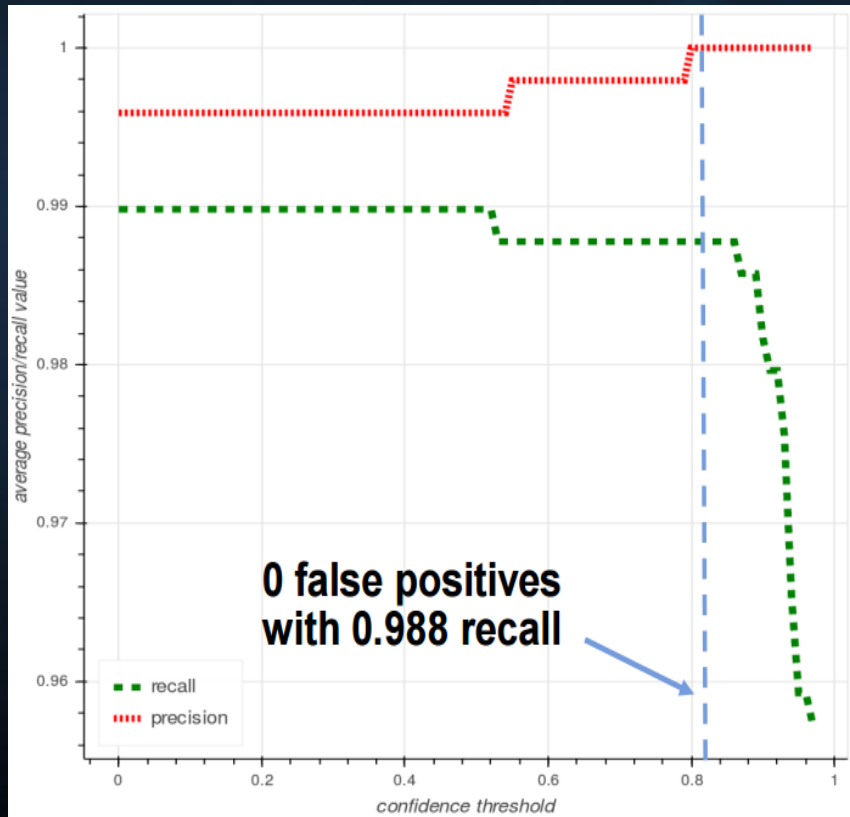
10 titles  
98.6% accuracy



# Tuning for precision

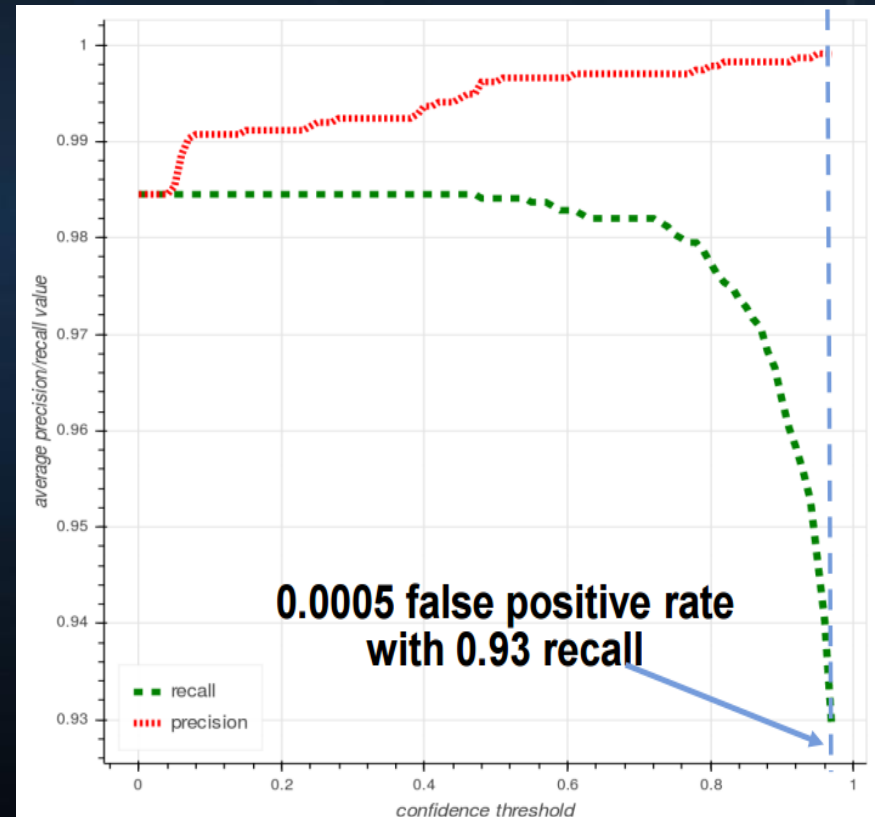
## YouTube

*feature: total burst size*



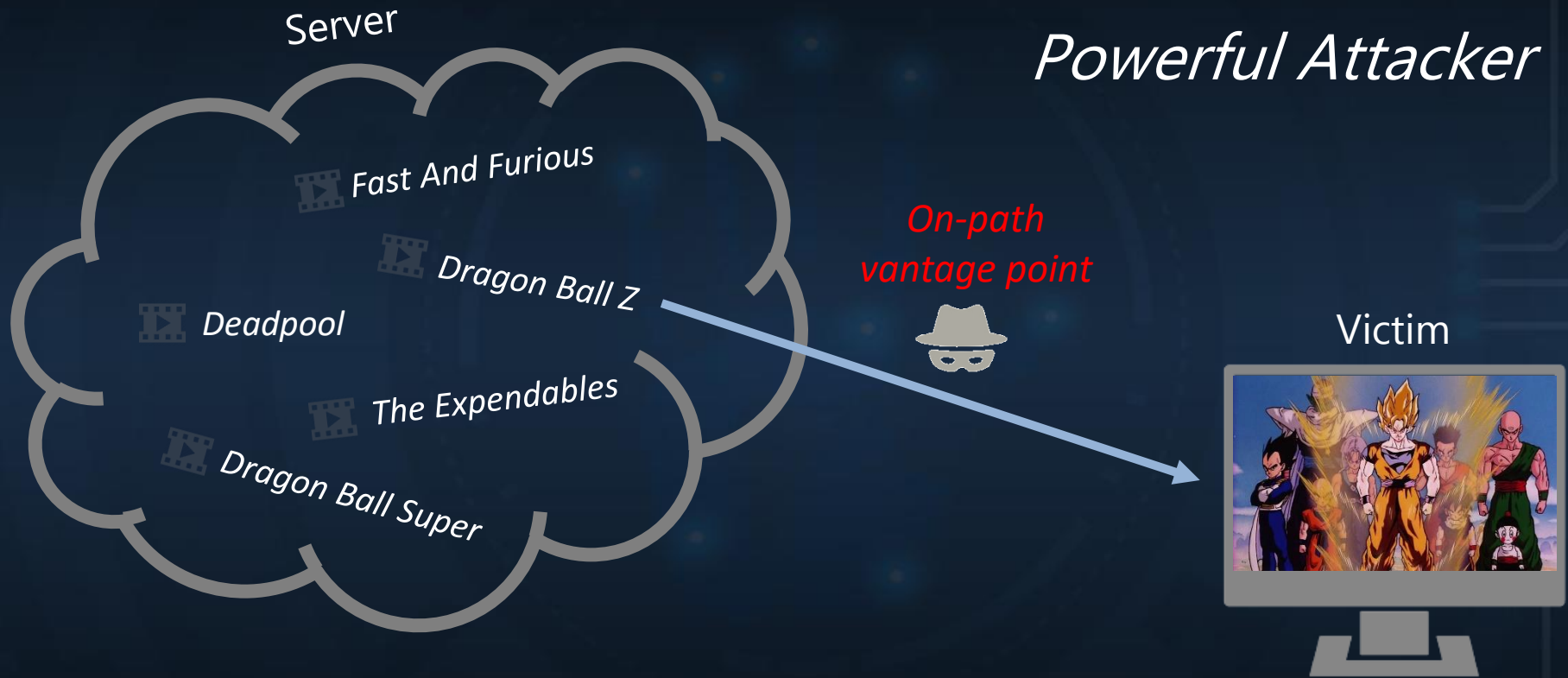
## Netflix

*feature: total burst size*



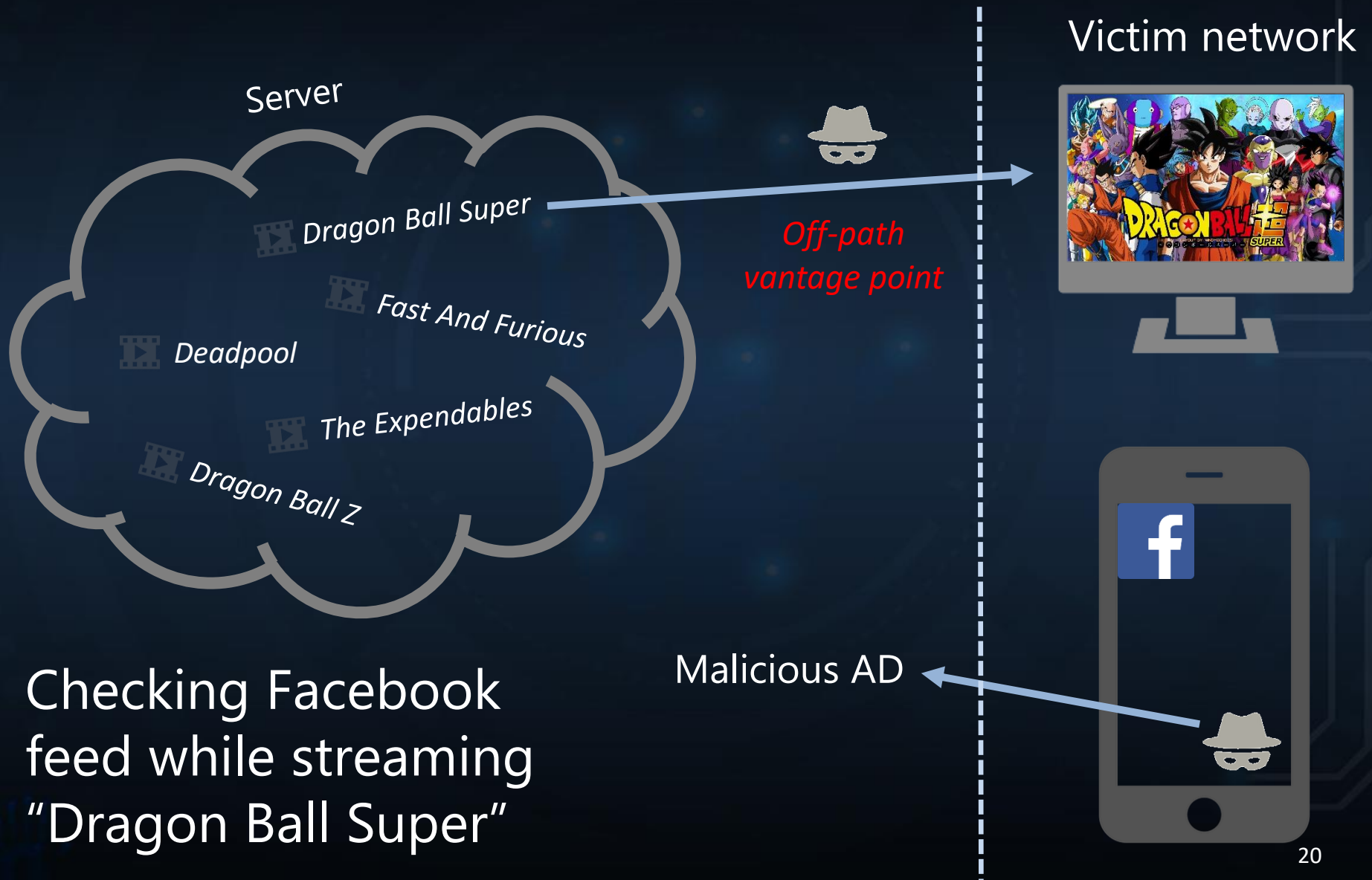
So what is the **vantage point** ?

# Scenario I: on-path attack

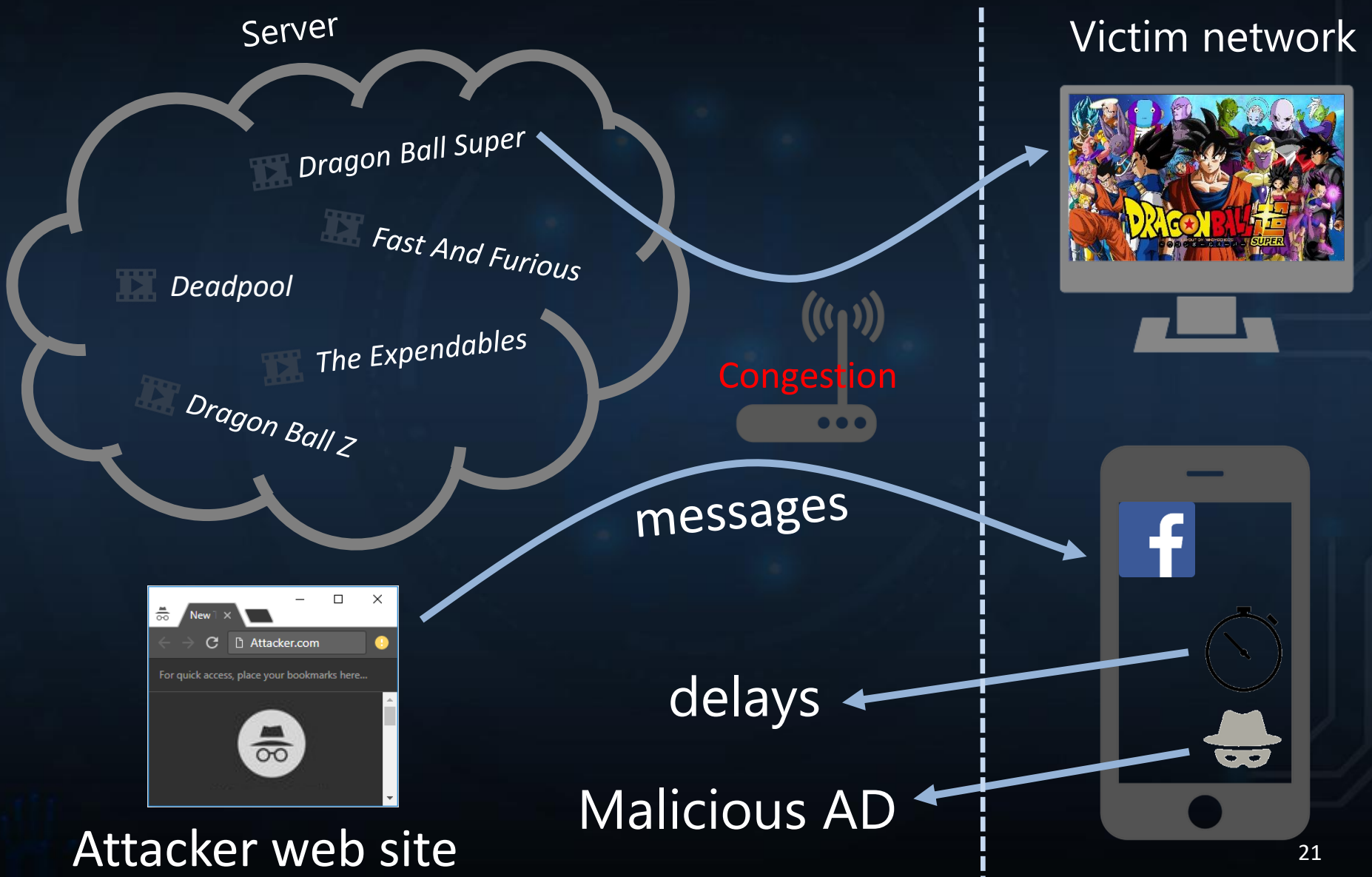


Wi-Fi access points, proxies, routers,  
enterprise or national network censors, ISPs

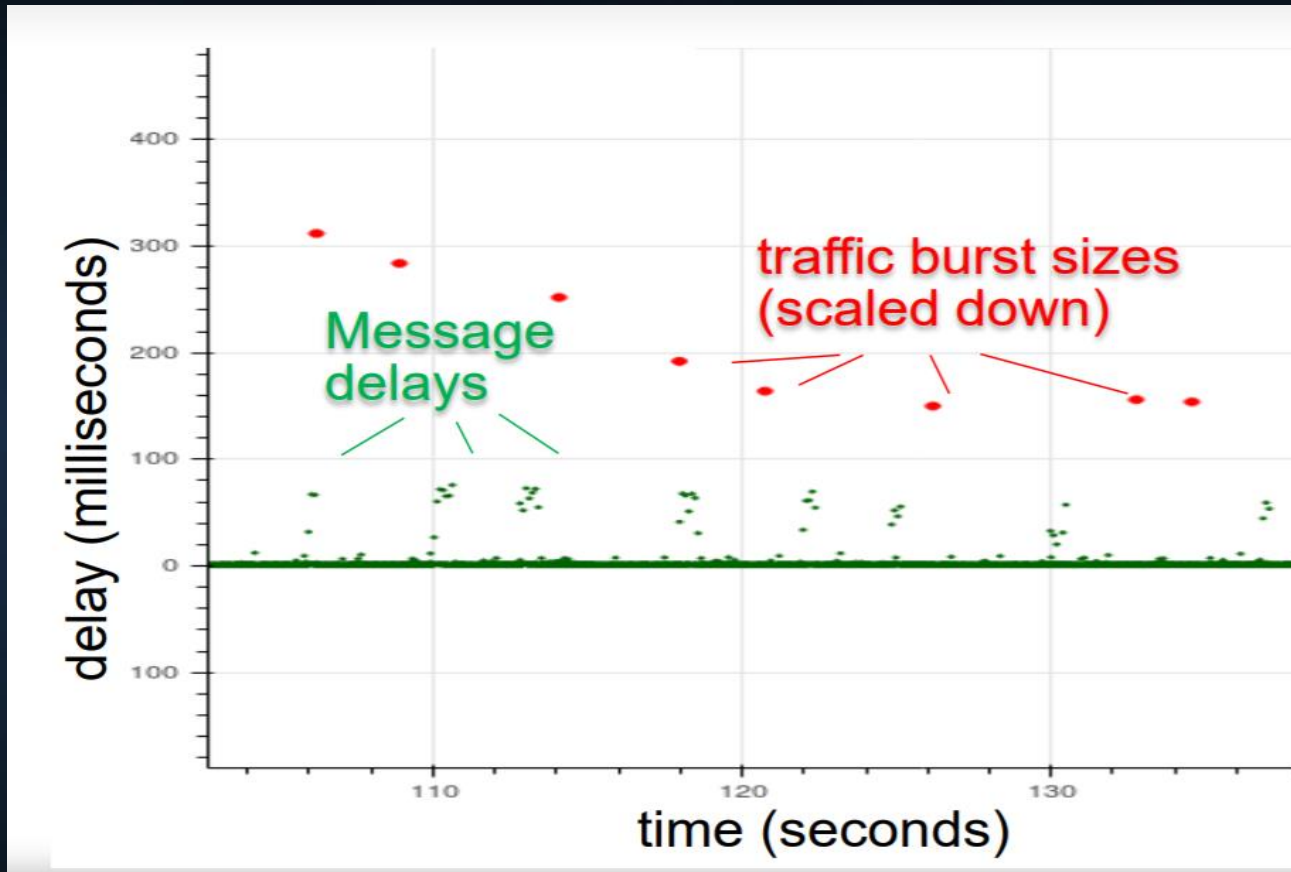
# Scenario II: off-path attack



# Cross-device attack

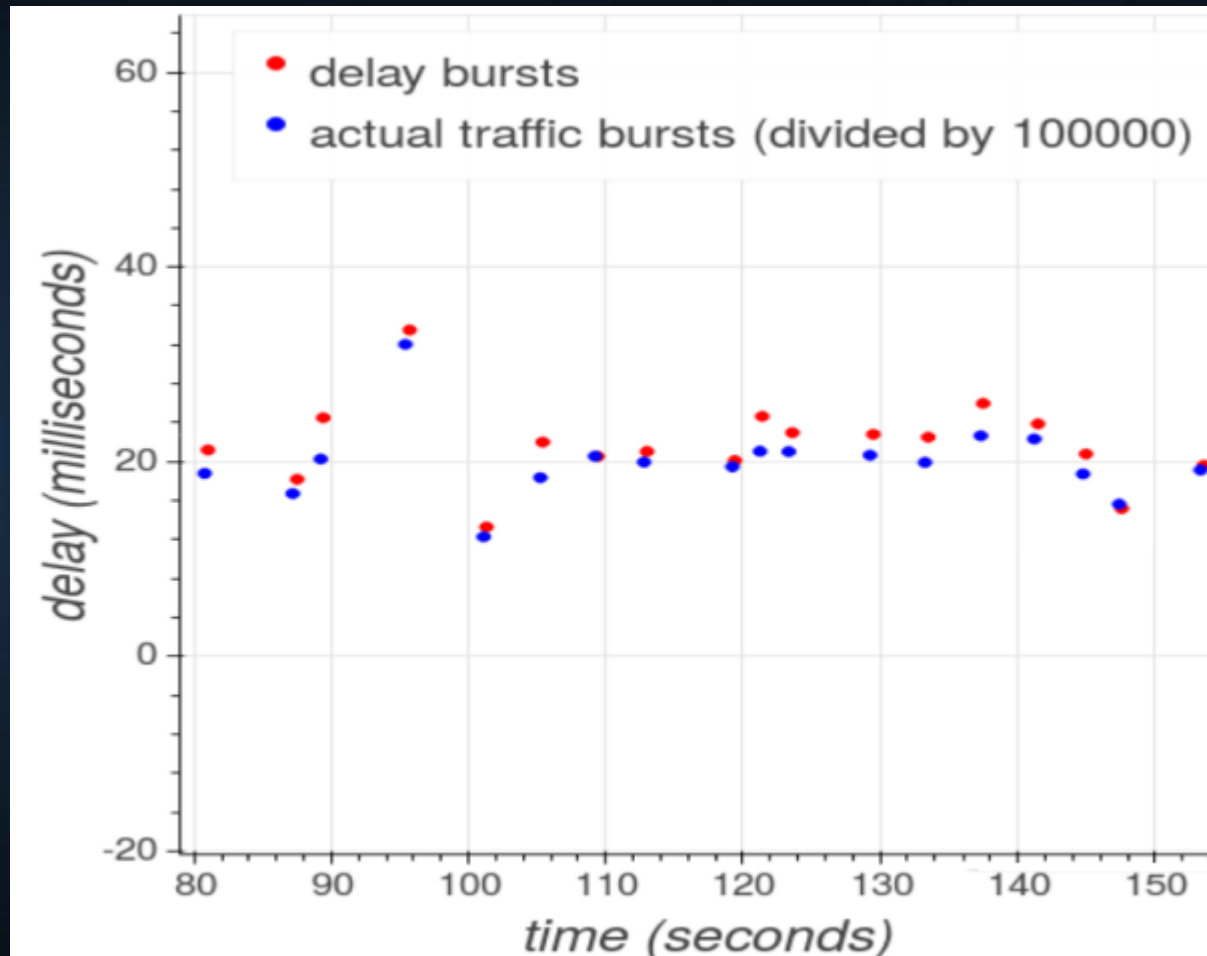


# Delay-bursts



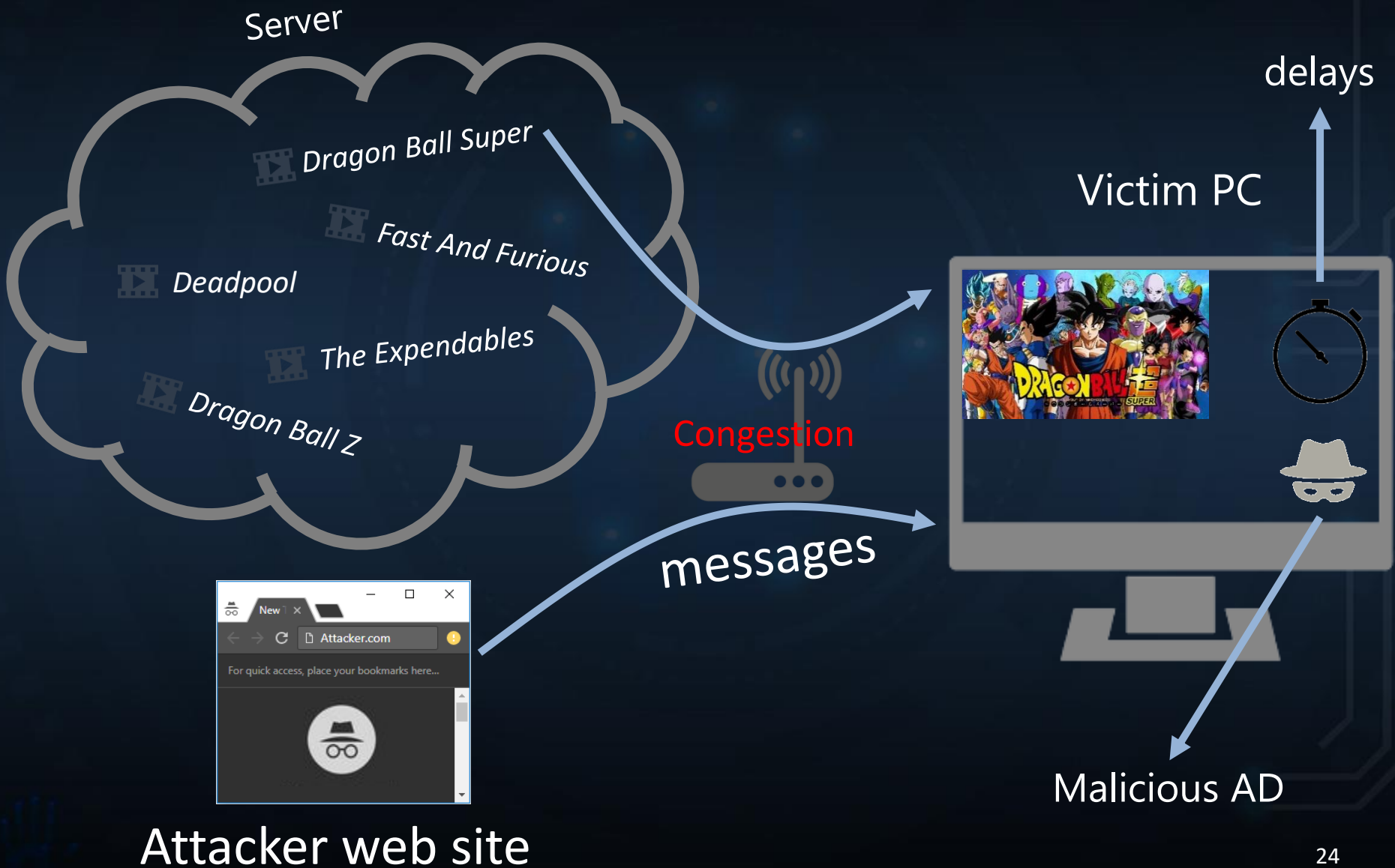
For each traffic burst, compute aggregate delay induced  
Use resulting time-series as input to neural network

# Delay-bursts VS. Traffic-bursts



**Delay-bursts**  
time series: the  
delays induced  
by traffic bursts

# Cross-site attack





# Mitigating the MPEG-DASH leak

- Modern streaming traffic characteristics
  - Title bitrate pattern unique when sampled at few-seconds granularity
  - Fetching at segment granularity (every few seconds)
- Maximizes quality of experience (QoE), server load, and network bandwidth utilization
- However, **information leakage is intrinsic...**

# Conclusions

- Leakage of information about video content via network traffic patterns is prevalent in modern streaming protocols and popular services
- Detectors are tuned for high accuracy and effective in an “open-world” setting
- It can be used by on-path adversaries such as ISPs to spy on their users
- It can be used by off-path adversary who merely serves a Web page to identify videos being streamed by the user

# Thank You !

- Further information and the paper:  
<https://beautyburst.github.io/>

*“Everything has a fingerprint,  
and so do encrypted streams”*

Any Questions?