

Side-Channel Attacks on Human Secrets

Yossi Oren, BGU

<https://iss.oy.ne.ro>

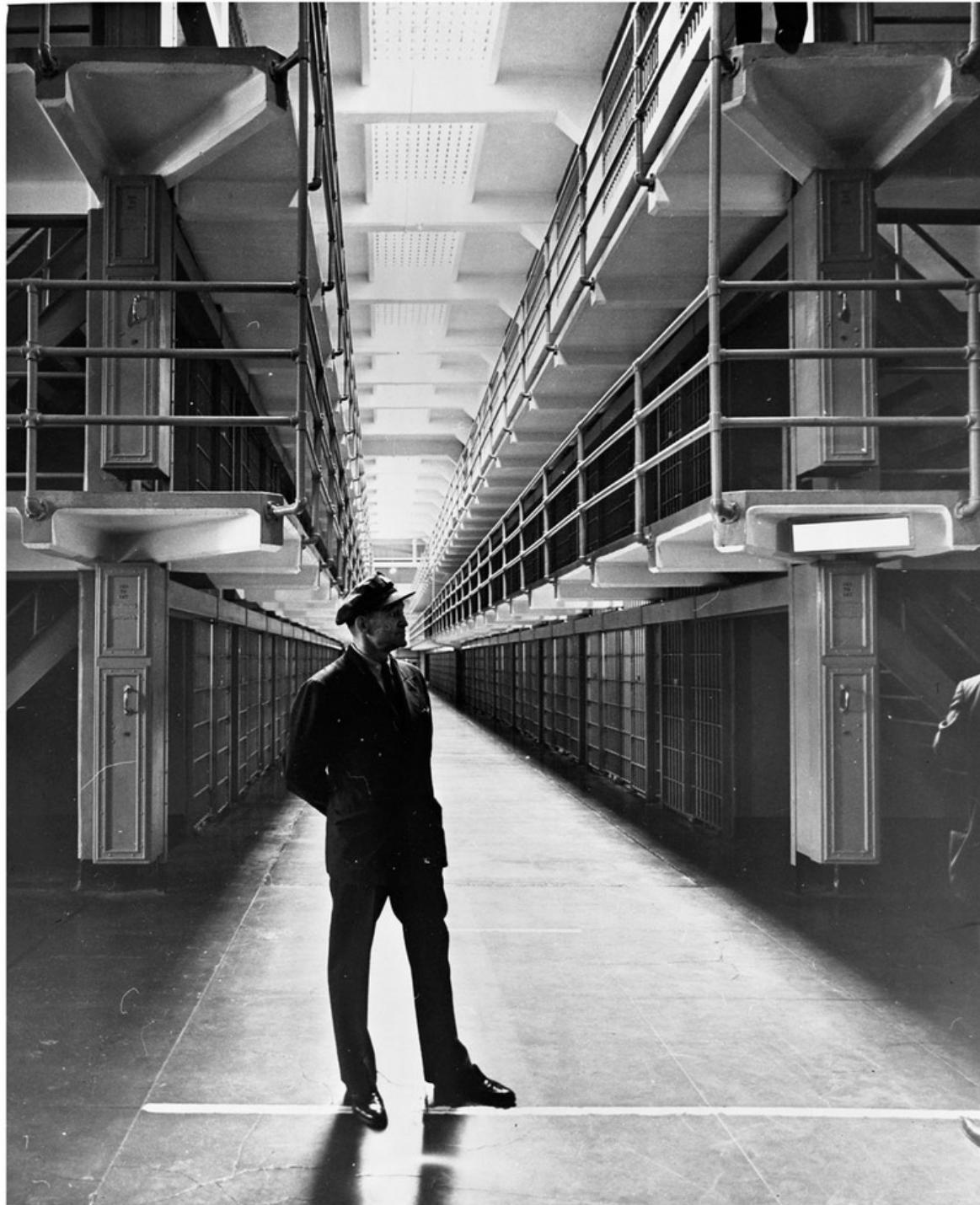
 @yossioren

7th Technion Summer School, 04/Oct/2018

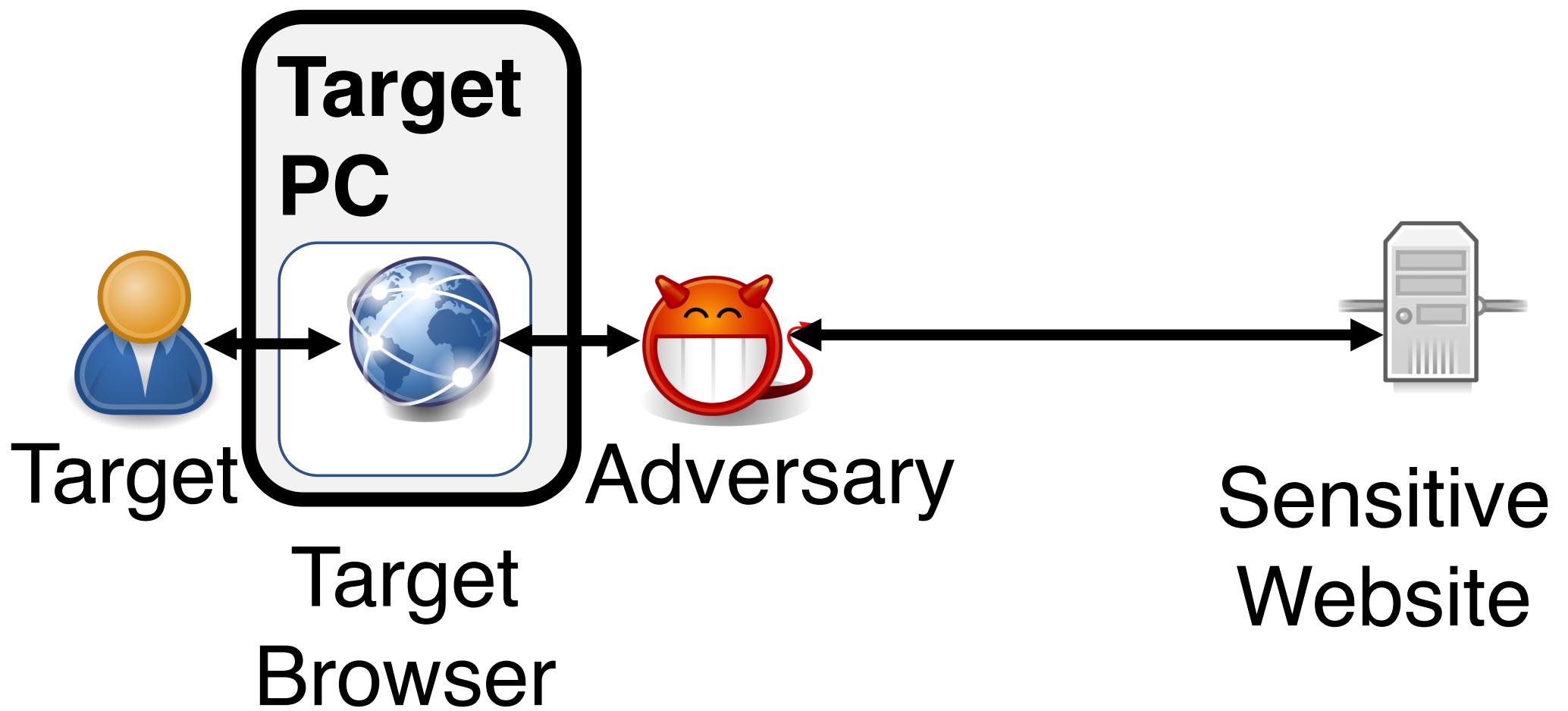
Joint work with Anatoly Shusterman, Lachlan Kang, Yosef Meltser, Yarden Haskal, Prateek Mittal and Yuval Yarom

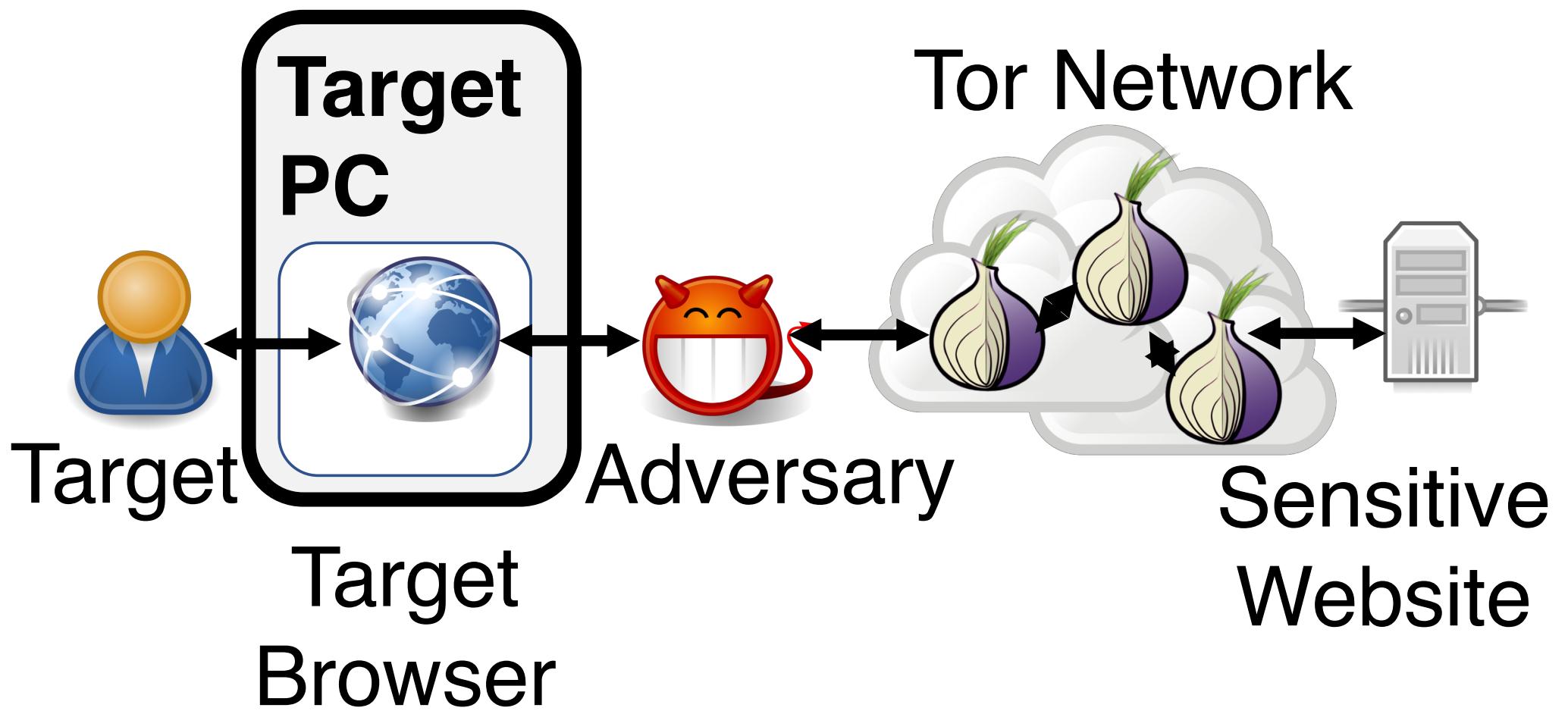
אוניברסיטת בן-גוריון בנגב
Ben-Gurion University of the Negev





Credit: SF Public Library courtesy of Golden Gate NRA, Park Archives, Interpretive Negative Collection, GOGA-2316





Website Fingerprinting

Automated Website Fingerprinting through Deep Learning

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Abstract—Several studies have shown that the network traffic that is generated by a visit to a website over Tor reveals information specific to the website through the timing and sizes of network packets. By capturing traffic traces between users and their Tor entry guard, a network eavesdropper can leverage this meta-data to reveal which website Tor users are visiting. The success of such attacks heavily depends on the particular set of traffic features that are used to construct the fingerprint. Typically, these features are manually engineered and, as such, any change introduced to the Tor network can render these carefully constructed features ineffective. In this paper, we show that an adversary can automate the feature engineering process, and thus automatically deanonymize Tor traffic by applying our novel method based on deep learning. We collect a dataset comprised of more than three million network traces, which is the largest dataset of web traffic ever used for website fingerprinting, and find that the performance achieved by our deep learning approaches is comparable to known methods which include various research efforts spanning over multiple years. The obtained success rate exceeds 96% for a closed world of 100 websites and 94% for our biggest open world of 900 classes. In our open world evaluation, the most performant deep learning model is 2% more accurate than the state-of-the-art attack. Furthermore, we show that the implicit features automatically learned by our approach are far more resilient to dynamic changes of web content over time. We conclude that the ability to automatically construct the most relevant traffic features and perform accurate traffic recognition makes our deep learning based approach an efficient, flexible and robust technique for website fingerprinting.

I. INTRODUCTION

Tor (Tor) is a communication tool that provides privacy of its users. It is an actively developed

never the origin and destination of a communication at the same time. Tor's architecture thus prevents ISPs and local network observers from identifying the websites users visit. As a result of previous research on Tor privacy, a serious side-channel of Tor network traffic was revealed that allowed a local adversary to infer which websites were visited by a particular user [14]. The identifying information leaks from the communication's meta-data, more precisely, from the directions and sizes of encrypted network packets. As this side-channel information is often unique for a specific website, it can be leveraged to form a unique fingerprint, thus allowing network eavesdroppers to reveal which website was visited based on the traffic that it generated.

The feasibility of Website Fingerprinting (WF) attacks on Tor was assessed in a series of studies [25], [31], [19], [24], [32]. In the related works, the attack is treated as a classification problem. This problem is solved by, first, manually engineering features of traffic traces and then classifying these features with state-of-practice machine learning algorithms. Proposed approaches have been shown to achieve a classification accuracy of 91-96% correctly recognized websites [30], [24], [113] in a set of 100 websites with 100 traces per website. Their works show that finding distinctive features is essential for accurate recognition of websites. Moreover, this tasks can be costly for the adversary as he has to keep up with changes introduced in the network protocol [4], [20], [9]. The WF research community thus far has not investigated the success of an attacker who automates the feature extraction step for classification. This is the key problem that we address in this work.

An essential step of traditional machine learning is feature engineering. Feature engineering is a manual process, based on domain knowledge, to find a representation of raw data that are most relevant to the task to be even more

- Collect Labeled Network Traces
- Extract Features
- Train Classifier (classical/deep)
- Classify Unknown Network Traces

How is WF Evaluated?

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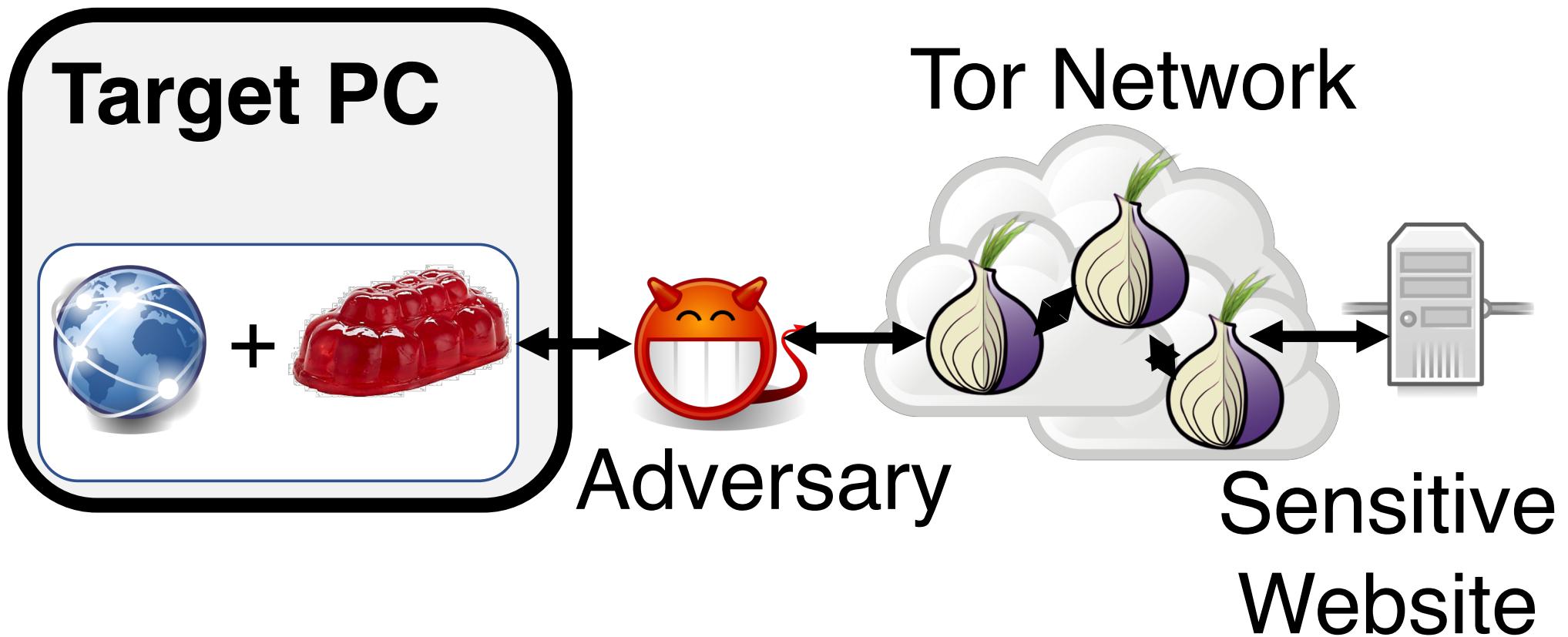
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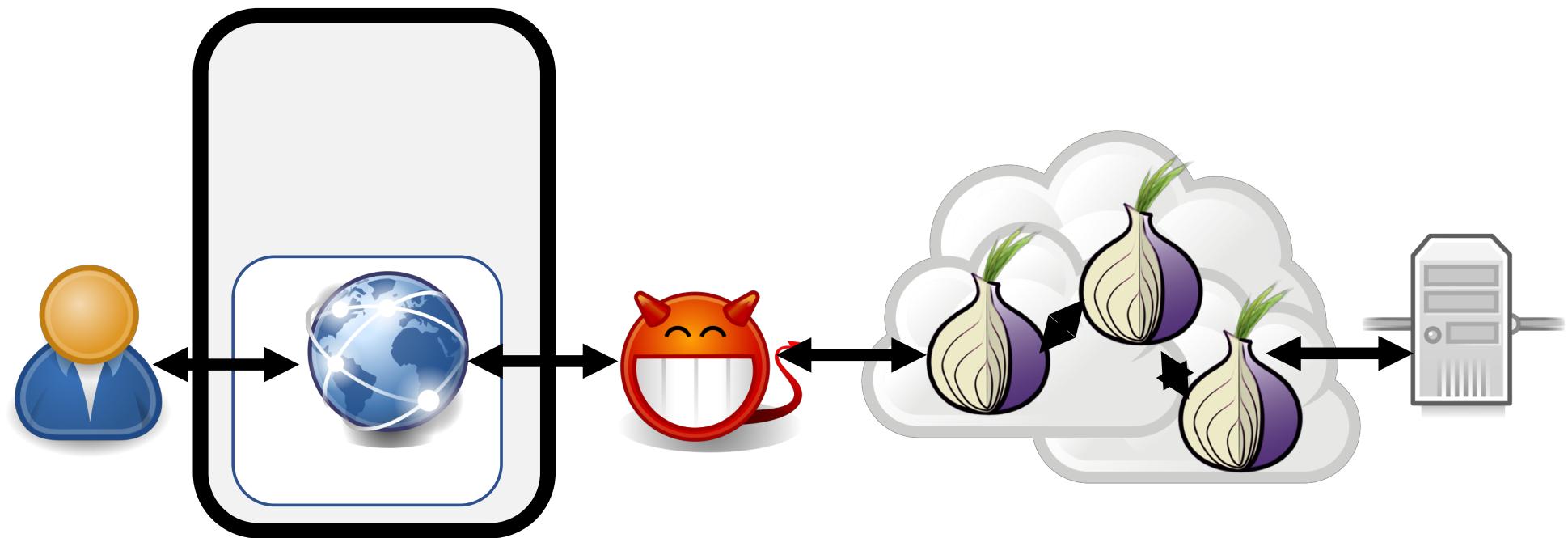
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- Main metric is accuracy
- Closed World vs Open World
- Base rate is important!
- Network based WF has >90% accuracy

Traffic Moulding Defenses against WF

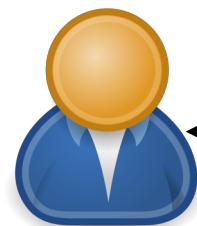






Adversary

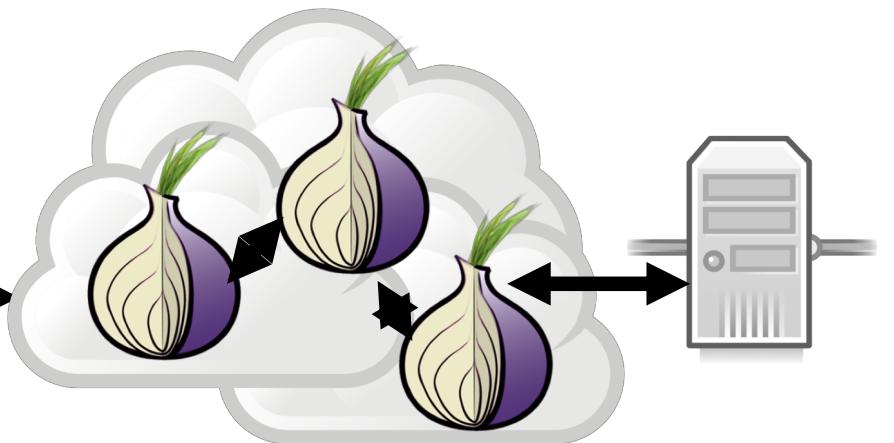
**Architectural
Boundary**



Target

Target
Browser

Tor Network



Sensitive
Website

forum.antichat.ru

FORUMS MEMBERS RECENT POSTS LOG IN

Раскрутка сайта: *Выход в ТОП-10 Я и G, поднятие тИЦ , 3000+ отзывов*

ПРЕДЛАГАЮ ВИРТУАЛЬНЫЕ СЕРВЕРЫ
ПОД РАЗЛИЧНЫЕ ЦЕЛИ И ПРОЕКТЫ.

Финансовые задачи/Социальные сети Покупка, Продажа, Обмен Трафик, инсталлы, загрузки - Покупка, продажа Search...

Продажа качественных загрузок

Discussion in 'Трафик, инсталлы, загрузки - Покупка, продажа' started by sasagiant, 2 Mar 2017.

2 Mar 2017 #1



sasagiant
New Member

Joined: 22 Feb 2017
Messages: 3
Likes Received: 0
Reputations: 0

Доброго времени суток:
Представляю вашему вниманию сервис по продаже инсталлов(кроме РУ и СНГ)!!!!
Доступны большие объемы.
Просьба уточнять цены и доступное количество в личке.
Интересны оптовые закупки и прогруз на постоянной основе.
Все средства для прогруза мы предоставляем сами,exe/dll.
Старт в течении 5-15 минут.

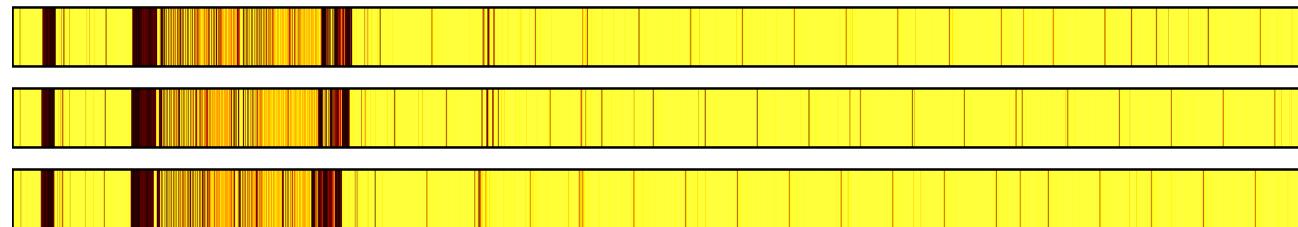
Информация по потокам :
Тематика : миксовая
Происхождение трафа : Бирж +спам
Работаем с лоадера , мин заказ 500
доступно . микс мира Канада и юса
Цена : микс мира 100USD за 1к .(цена на объемы обговариваем отдельно)
Наши контакты : alen.sgor@exploit.im

тема на других форумах ..
<https://fuckav.ru/showthread.php?t=32345&cdn=1>

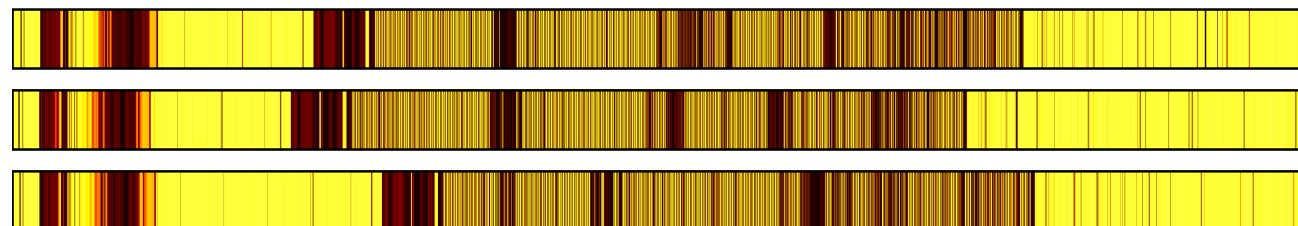
Last edited: 17 Jun 2018

Memorygrams

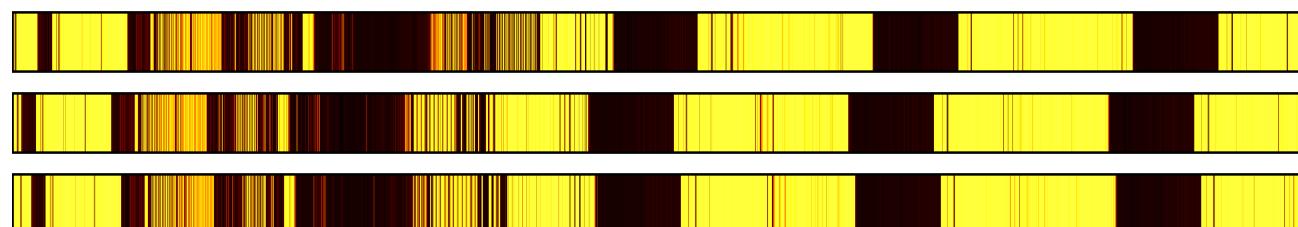
Wikipedia



Github



Oracle



Cache-Based WF

- Collect Labeled Memorygrams
- Extract Features
- Train Classifier (classical/deep)
- Classify Unknown Memorygrams
- >90% accuracy

Cache-based vs Net-based WF

Cache beats Net	Net beats Cache
Resists net countermeasures	Can be detected by victim
Robust to response caching	Depends on hardware config
Works across NICs	
Lighter attack model	

Conclusions

- Side-channel attacks can attack **human secrets**, not just **cryptographic secrets**
- Specifically, cache-based website fingerprinting is feasible and very dangerous to user privacy
- What other secrets can we attack?
- What kind of countermeasures apply here?
- Why aren't you joining my lab?

<https://iss.oy.ne.ro/Lab>

