

BLOX FEST

Infoblox 

The Front Doors Are Locked, But DNS is Wide Open - Preventing Data Exfiltration via DNS

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Topics of Discussion

- What kind of data is stolen and why
- Prevalence of data exfiltration and DNS tunneling
- How data is stolen via DNS tunneling
- Detection strategies
 - Behavior analysis (Patent pending)
 - Machine learning
 - Artificial intelligence
 - Architecture
- Detection results (Conference paper)
- Real-time detection and mitigation solution architecture
- Customer case study (T. J. Short, CISO, Everi)



Stealing Data – Why and What Kind?

PII (Personally Identifiable Information)

Information like social security numbers of employees or customers that cybercriminals can use to steal identity, or sell in the underground market for profit

Regulated Data

Data related to PCI DSS and HIPAA compliance that can be misused

Intellectual Property

Data that can give an organization a competitive advantage

Other Sensitive Information

Credit card numbers, company financials, payroll and emails



Hactivism



Espionage



Financial Profit



DNS and Data Exfiltration

DNS tunneling attacks

let infected endpoints or malicious insiders exfiltrate data.



Attackers have recently used DNS tunneling in cases involving the theft of **millions of accounts**.⁵



46%

of large businesses have experienced DNS exfiltration.⁶

\$3.8 M

Average consolidated cost of a data breach⁷

Goal of Malicious Actors

- Hacktivism
- Espionage
- Financial

Data Targets

- Regulated Data
- PII (personally identifiable information)
- Intellectual property
- Company financials, payroll data

5. SANS Institute paper referencing Ed Skoudis as speaker at RSA Conference, June 2012

6. DNS attacks putting organizations at risk, survey finds, SC Magazine, December 23, 2014

7. Ponemon Institute, 2015 Cost of Data Breach Study

Customer Examples



A large developer of video games had malware inside the network that tried to exfiltrate data via DNS queries using spoofed addresses



A large automaker's main concern is loss of intellectual property that could erode their competitive advantage, and the company is very keen on preventing it from happening via DNS



A large bank failed an audit because of lack of protection for data over DNS



A large insurance company is concerned about liability because they are aware that DNS is not protected

DNS Tunneling

Recursive DNS

Request

www.google.com
www.apple.com
gmail.com

74.125.25.104
23.210.209.236
216.58.192.5

Response

Client



Recursive DNS

Request

7r3ncahnt3s.dnst.com
dulhvl8sfdq6rj.dnst.com
qth9zu6574uk3j.dnst.com
siv9dmlmunfb.dnst.com

acvacv19w1gt79t49w1ctd
3kbel9tszfnjbhiwi3kvauh
NXDomain
ServFail

Response

Compromised Client



Recursive DNS

Request

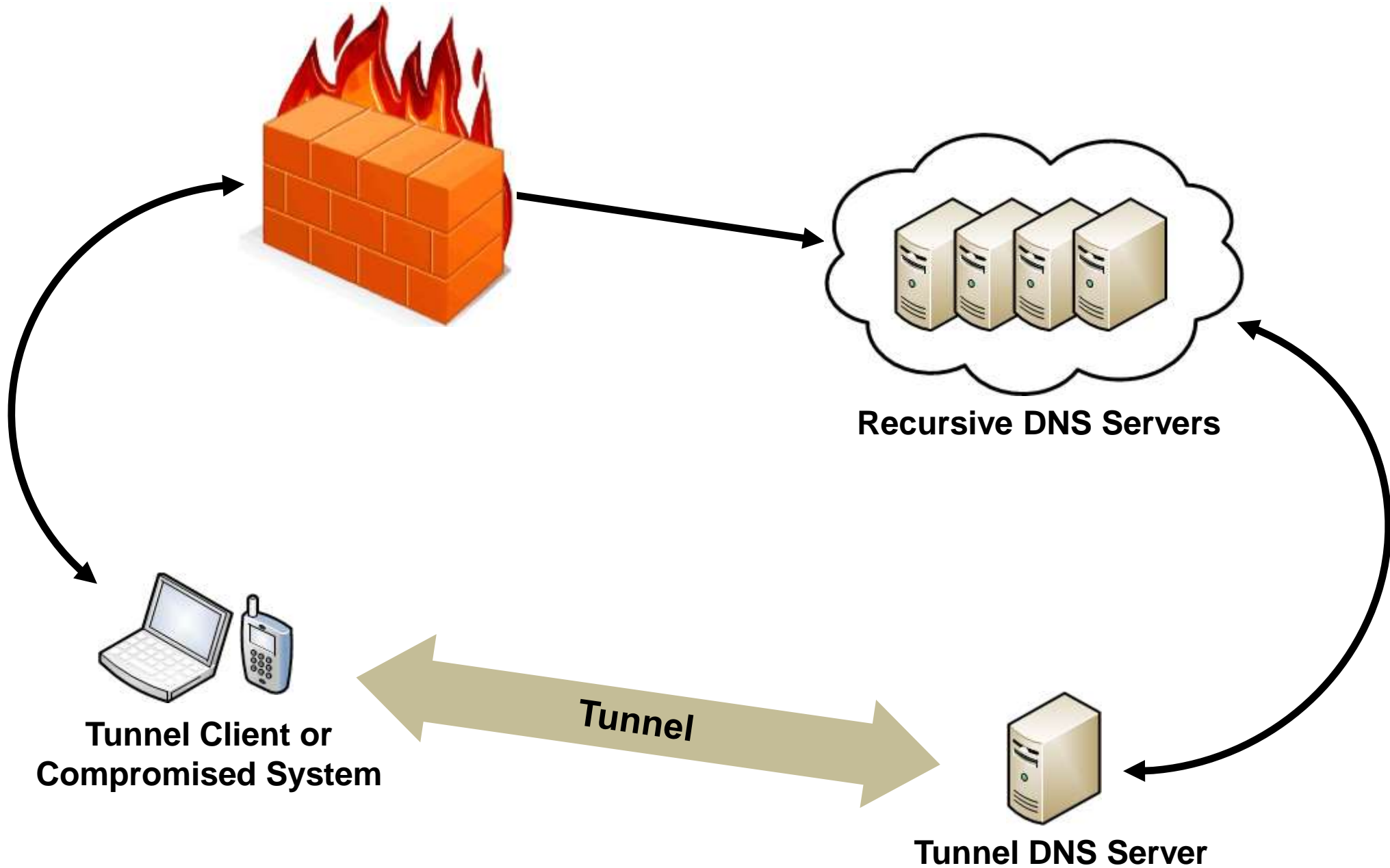
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NXDomain
ServFail

Response

Compromised Client



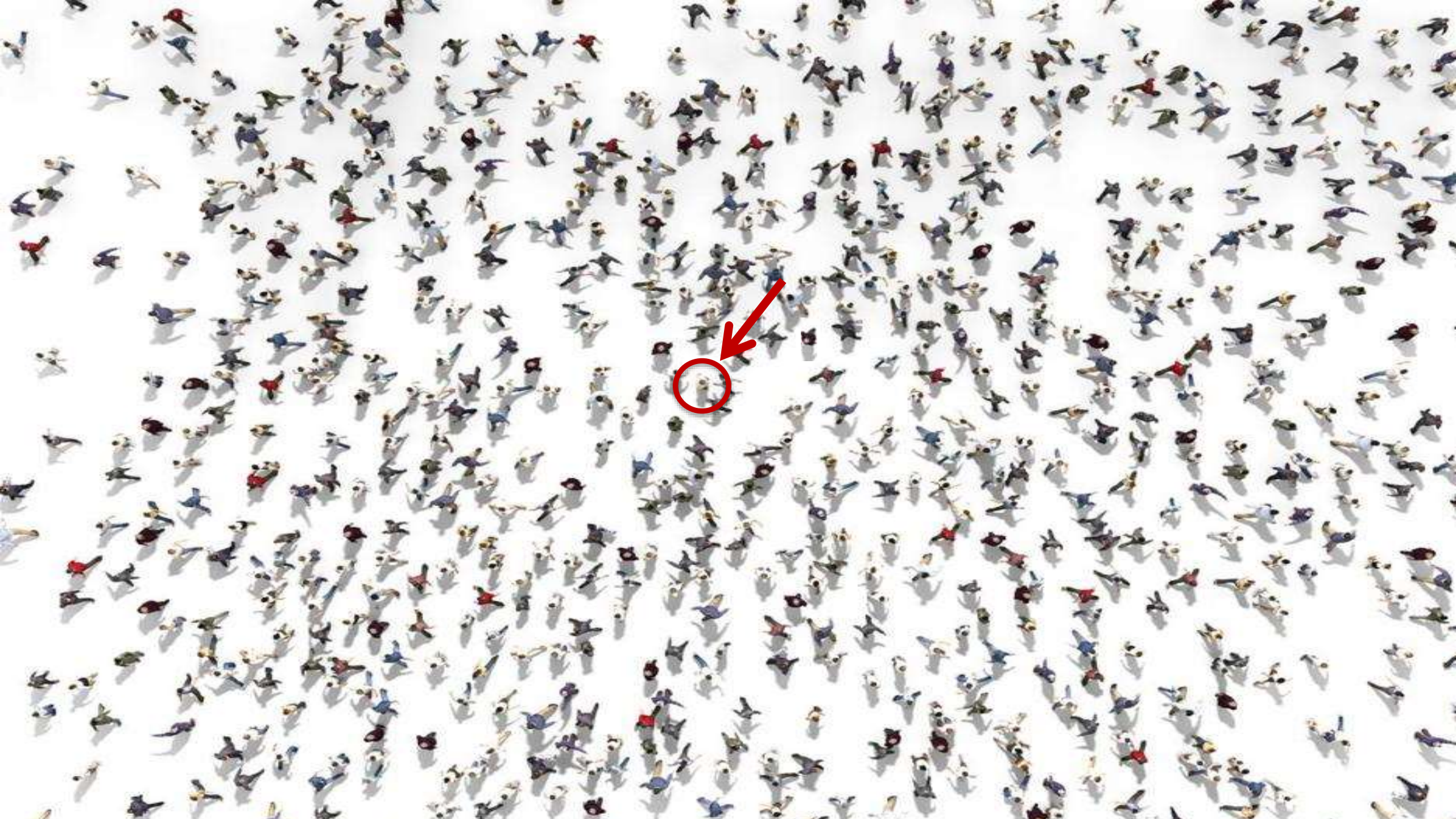


- Not firewalled
 - Data exfiltration
 - Command and control (C&C)
 - Free hotspot
-
- DNS tunneling is a technique
 - Legitimate vs malicious uses
 - No consistent signatures



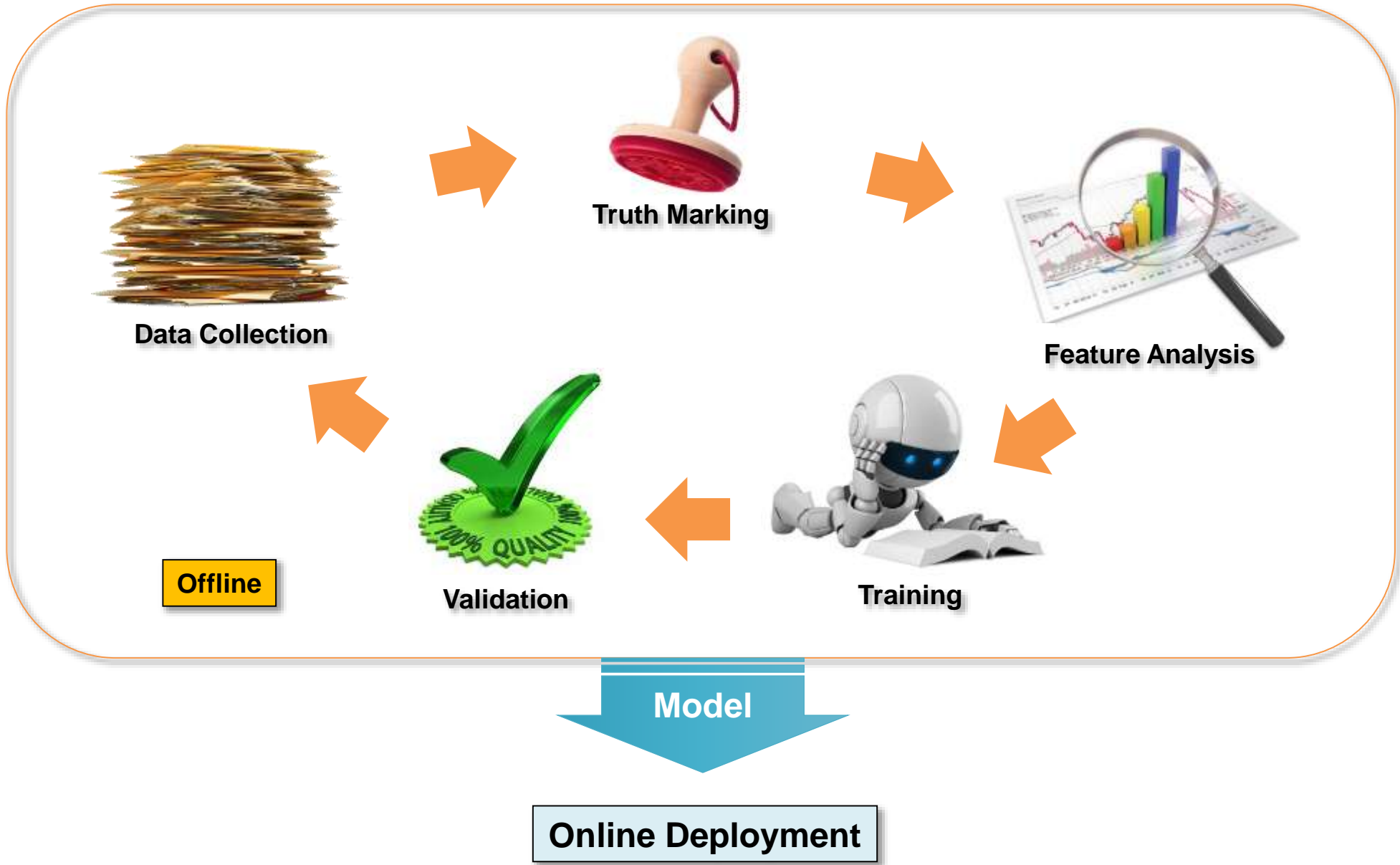
Behavior Analysis*

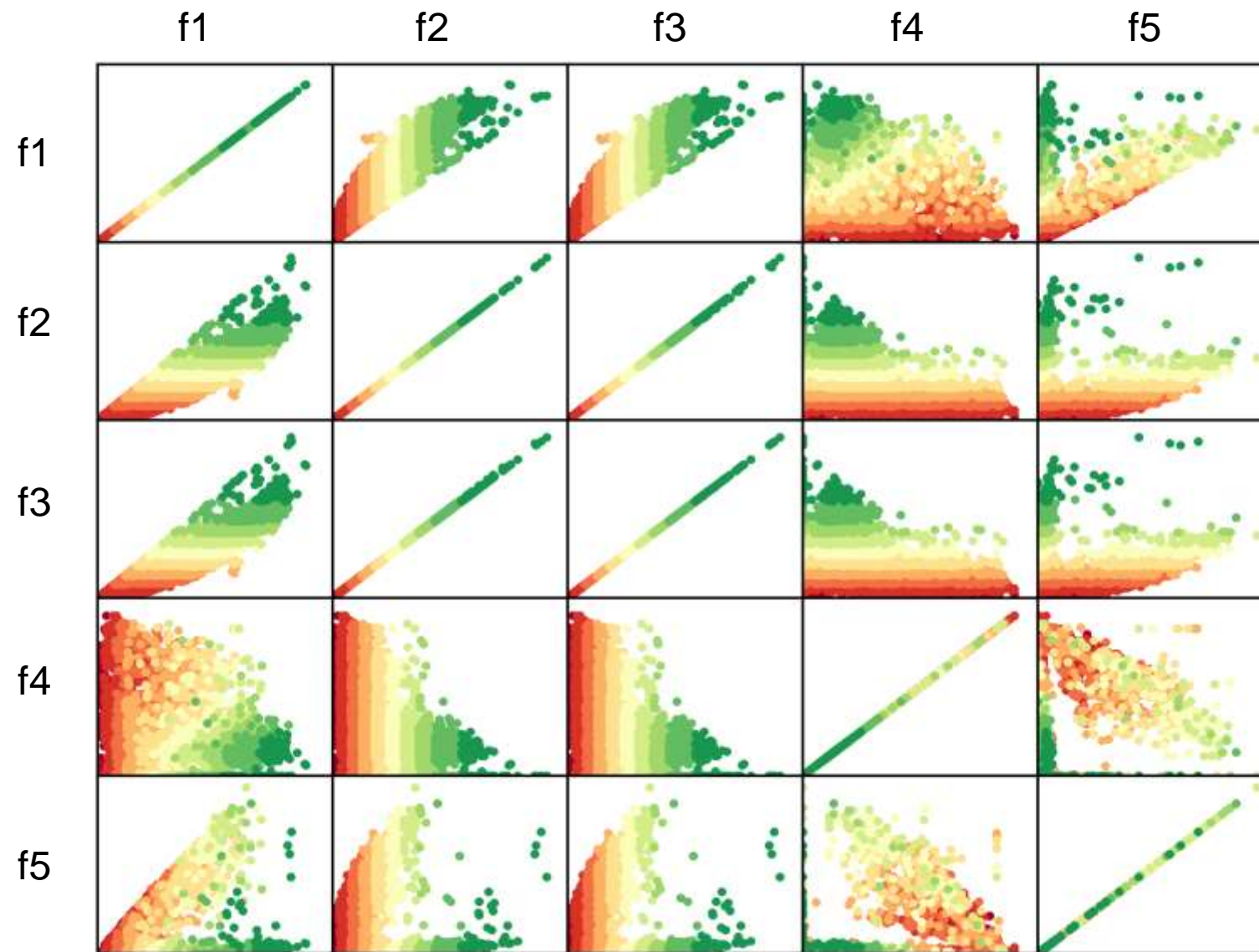
* Patent pending

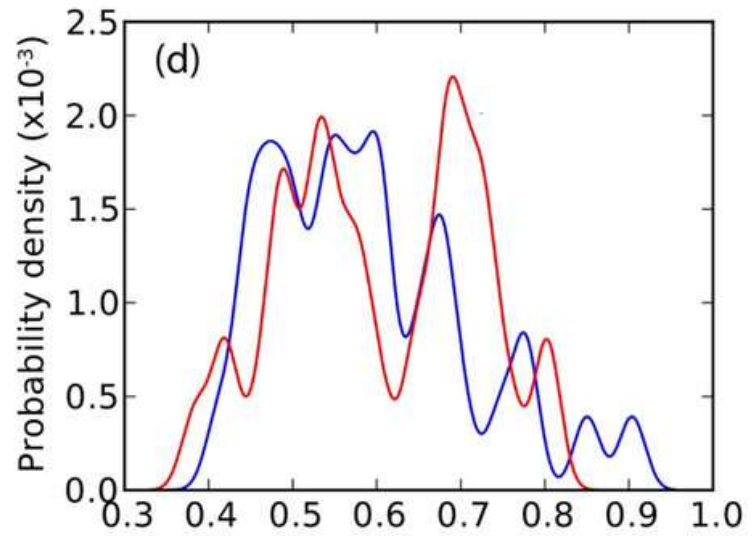
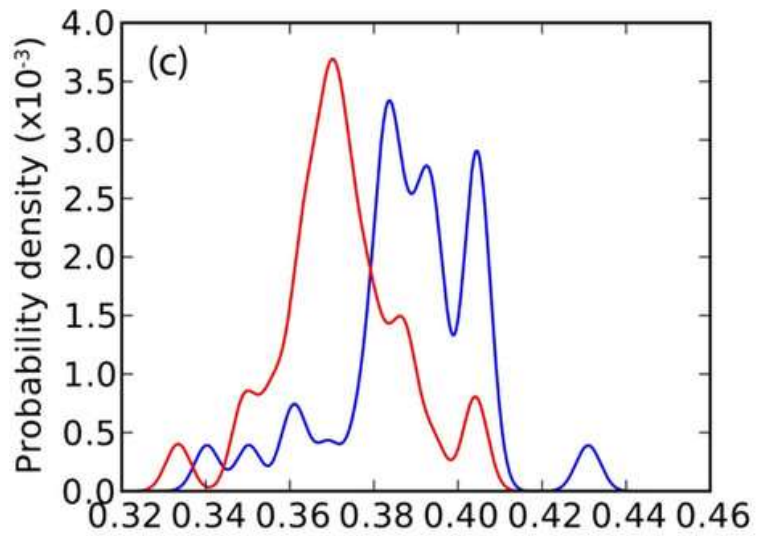
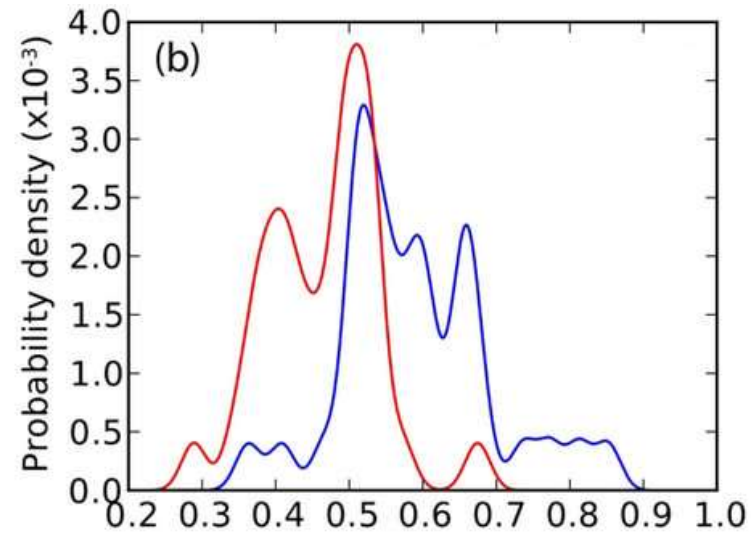
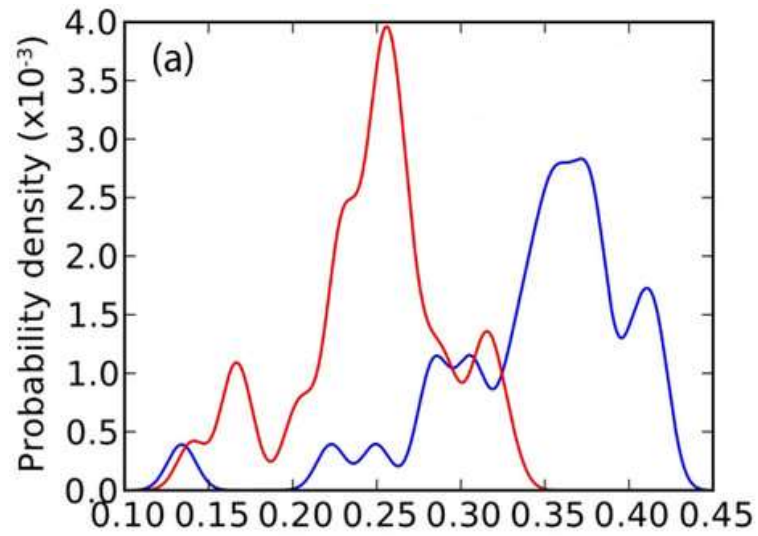




Machine Learning

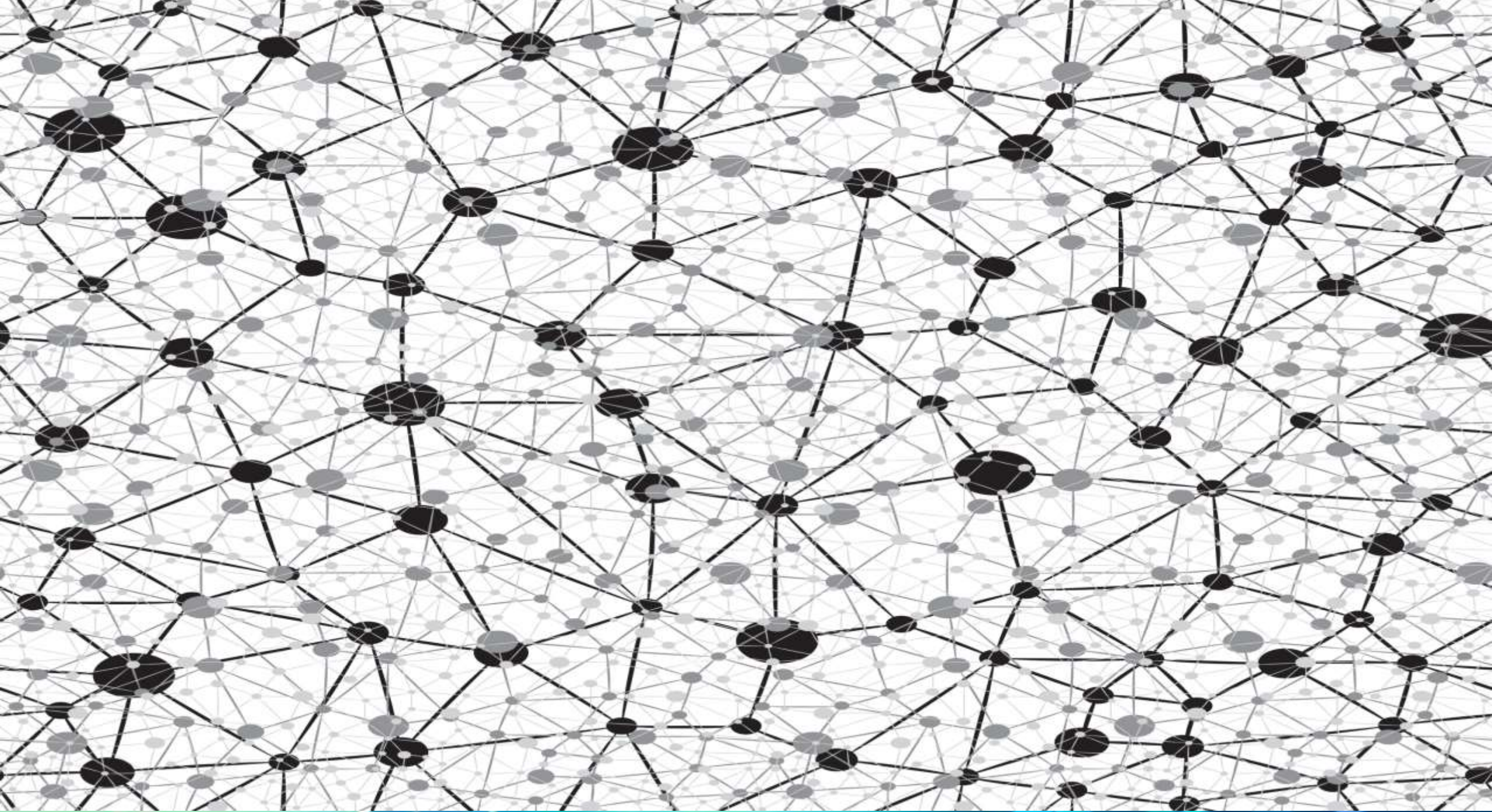


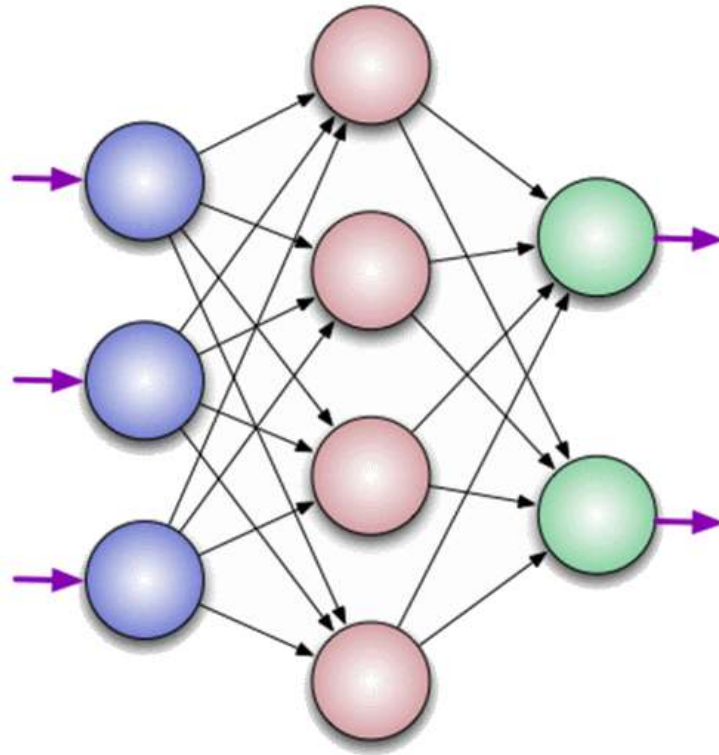
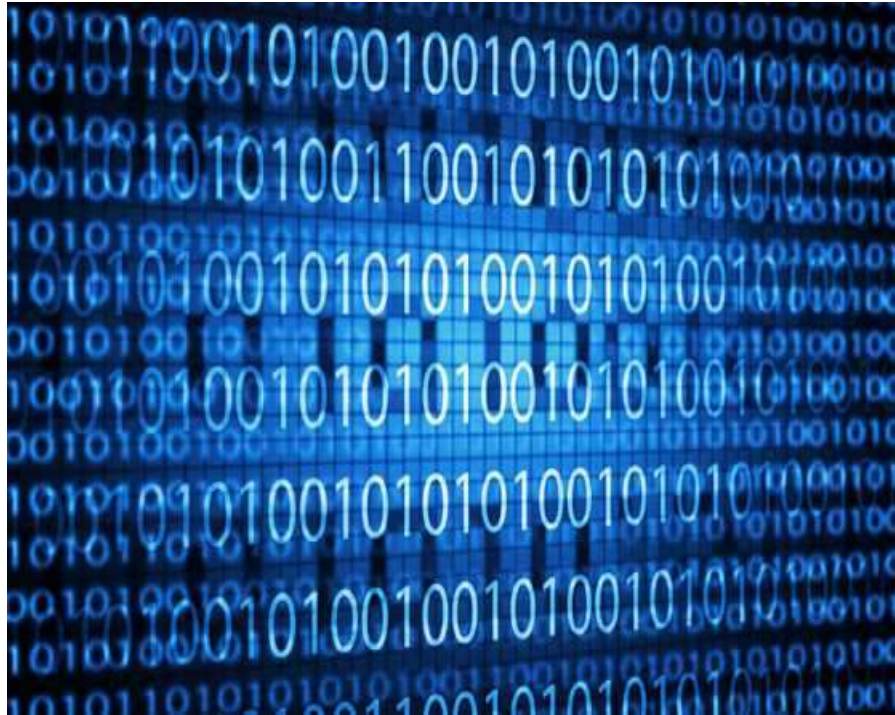




Neural Network

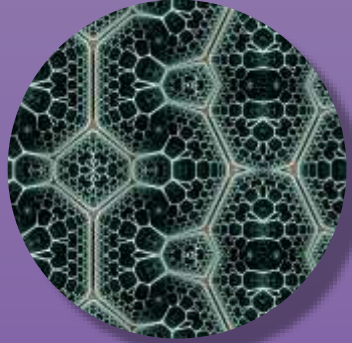






Artificial Intelligence





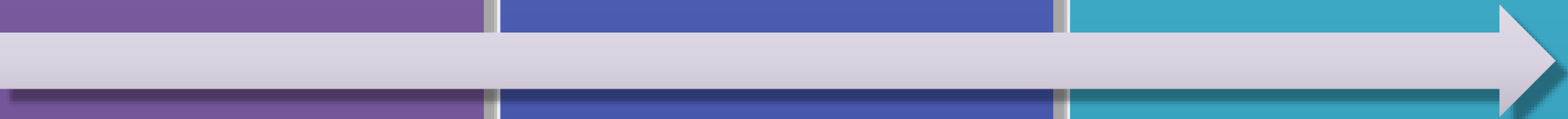
Data
Modeling

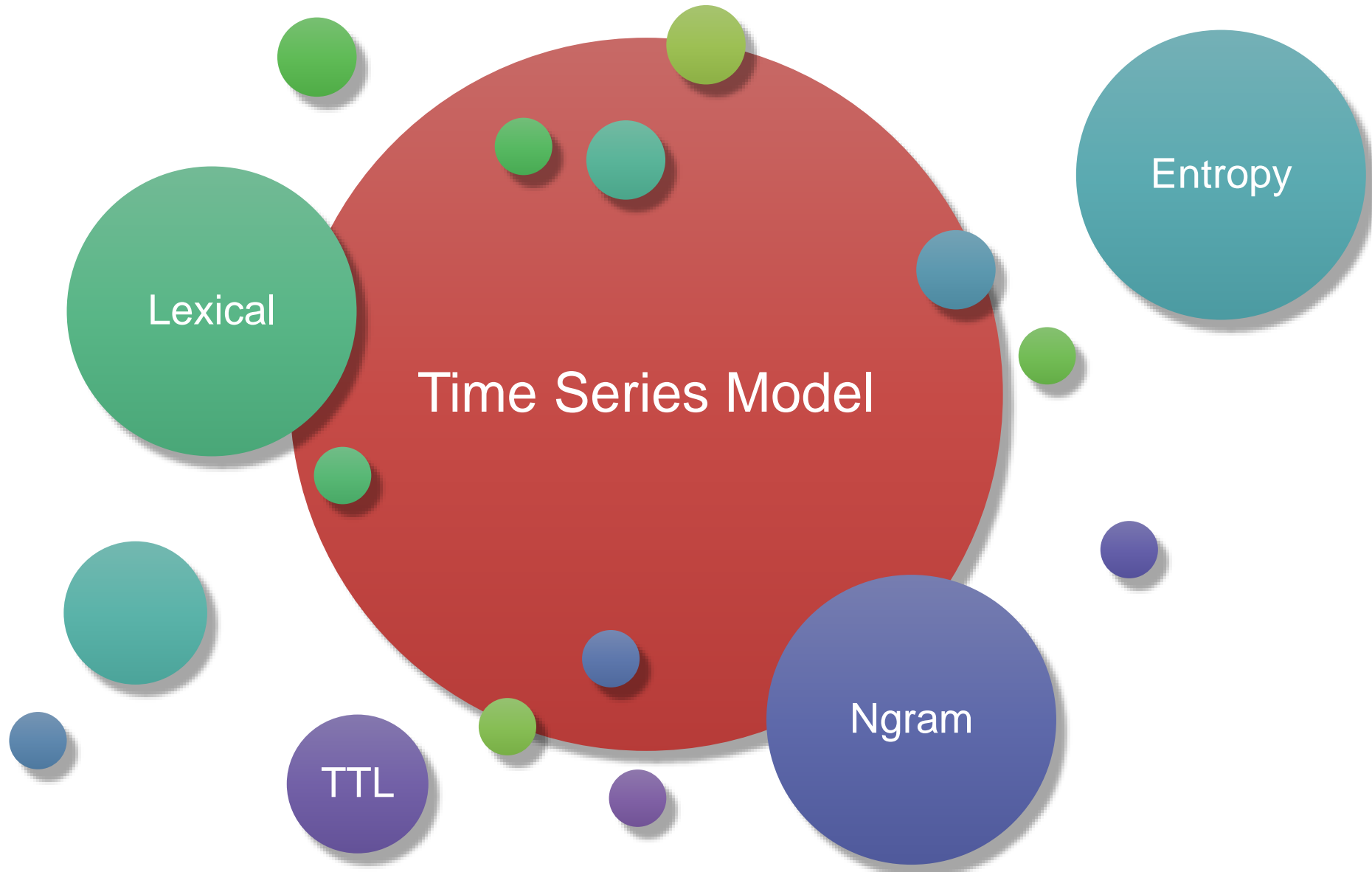


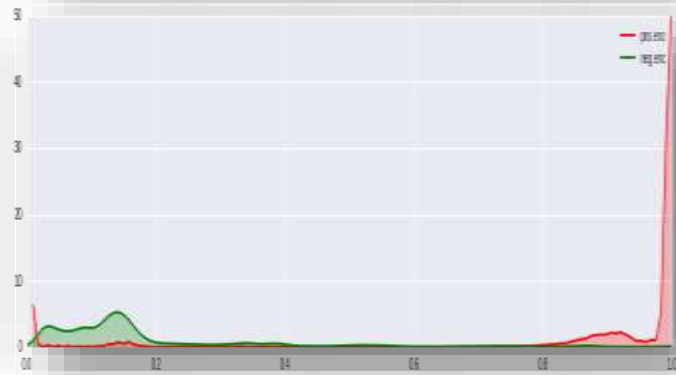
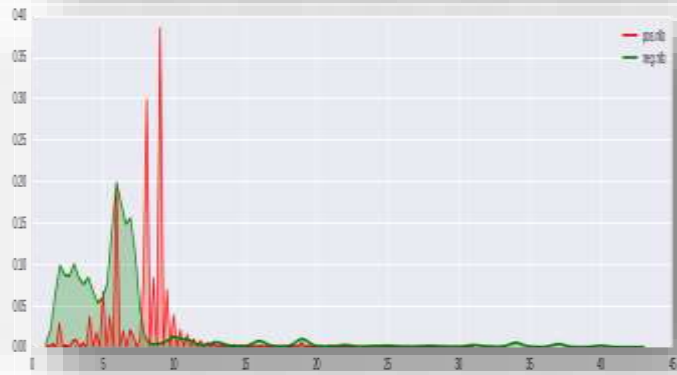
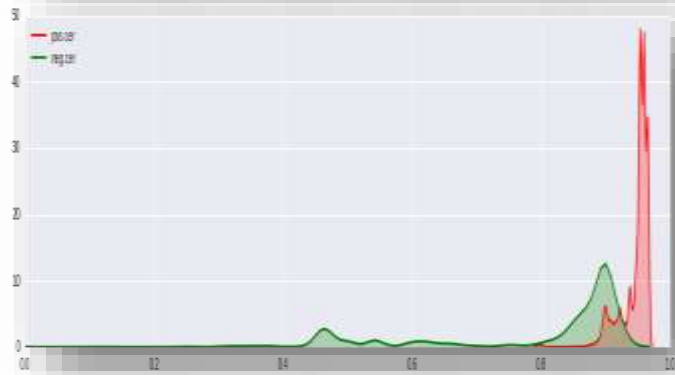
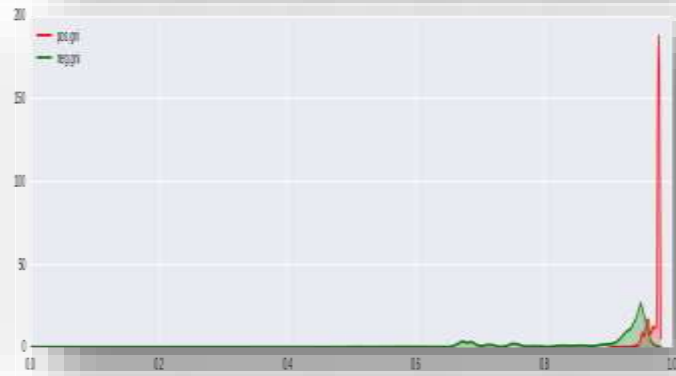
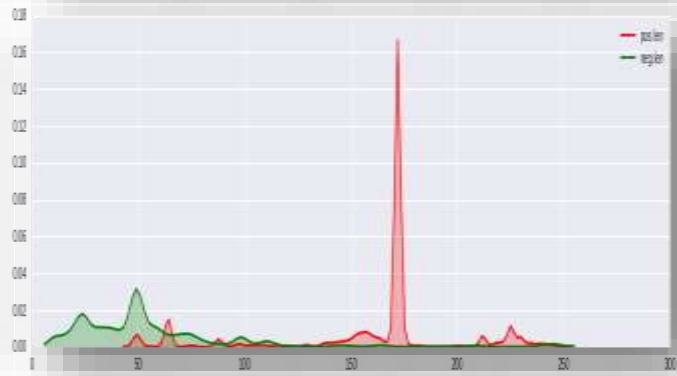
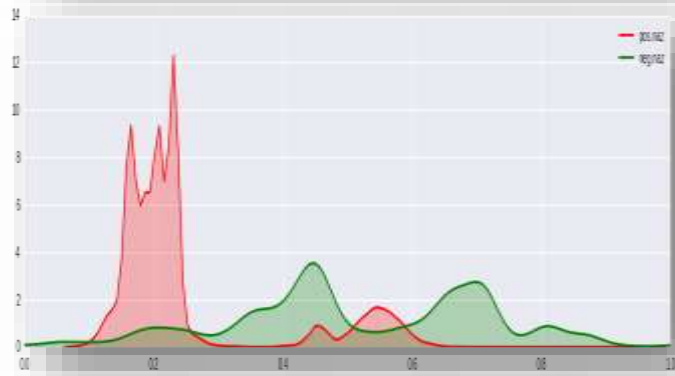
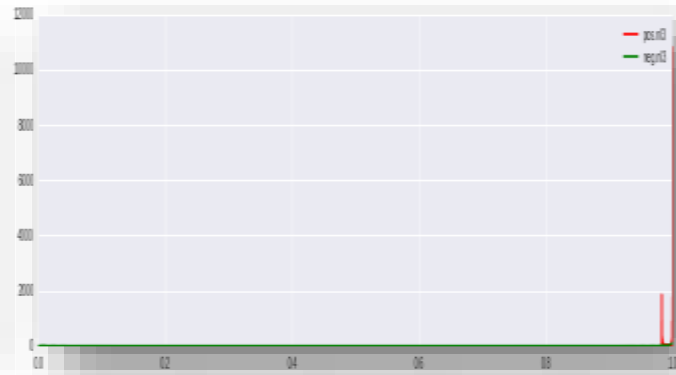
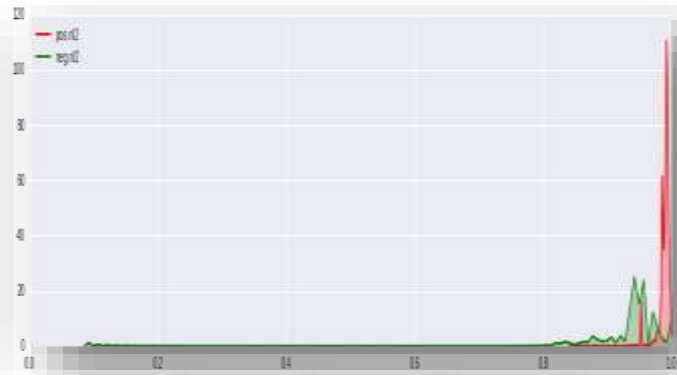
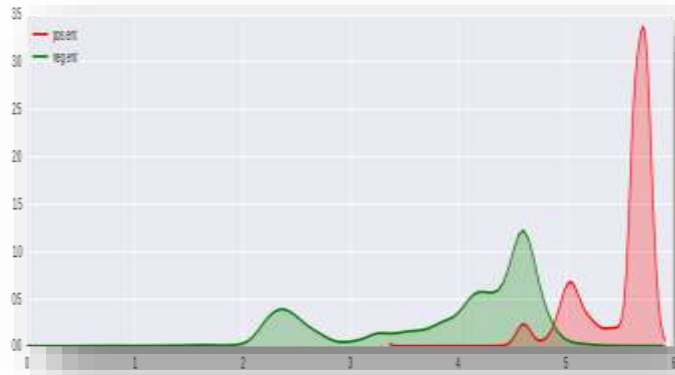
Classification

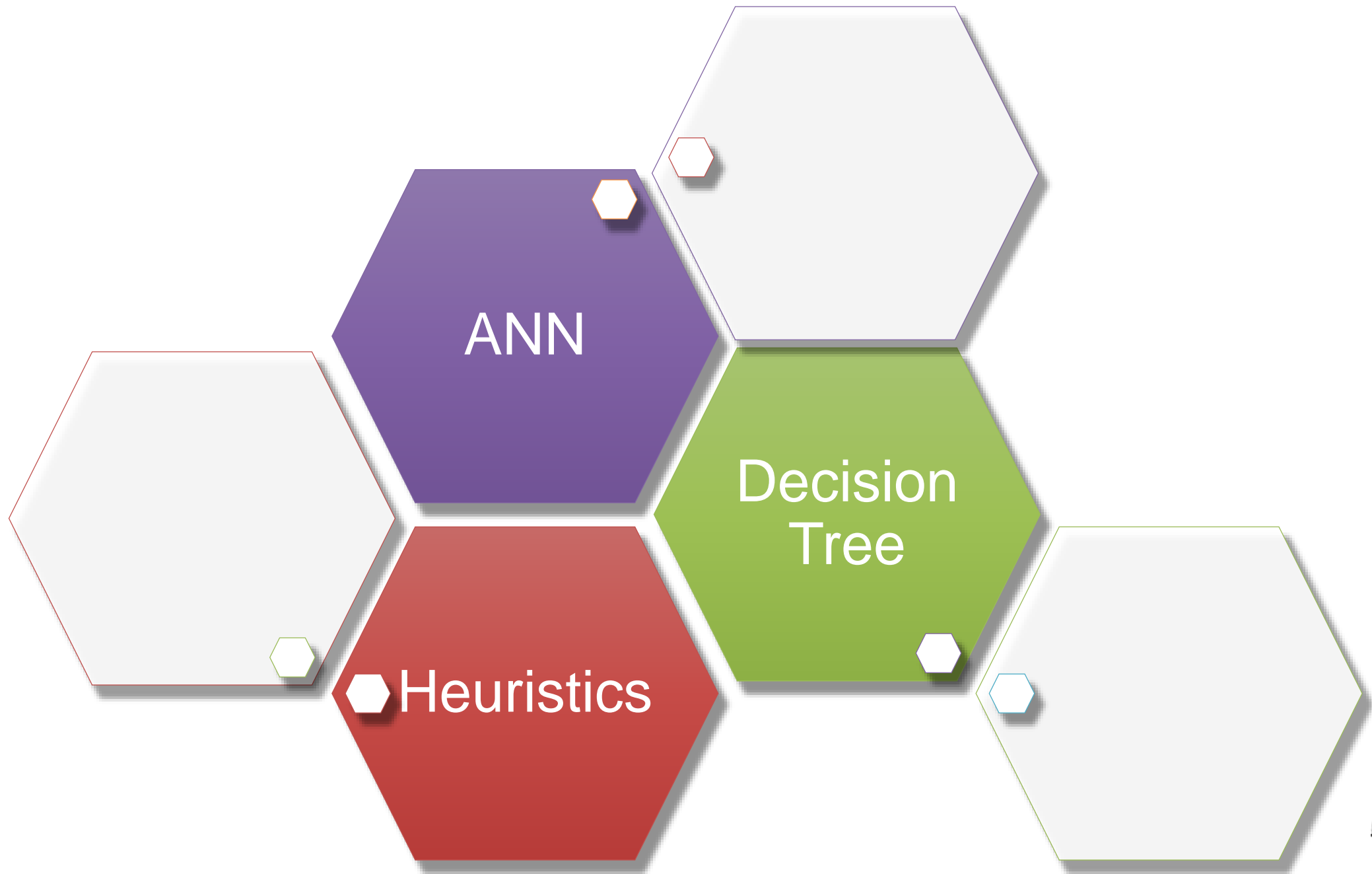


Artificial
Intelligence









Time Series Model

TSM

Type

T1

T2

T3

Behavior

B11

B12

B13

B21

B22

B31

Class

C111

C112

C121

C131

C211

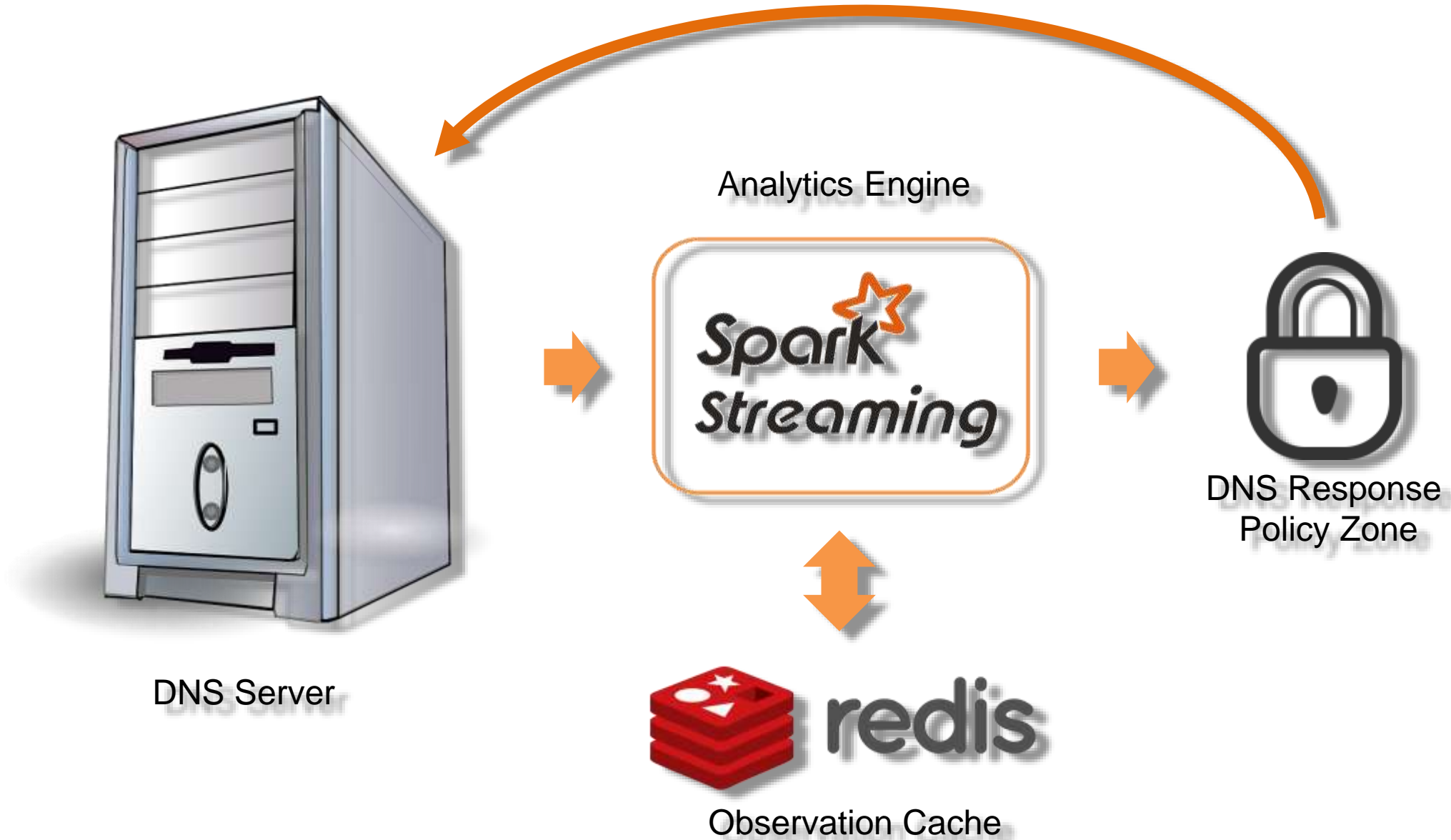
C212

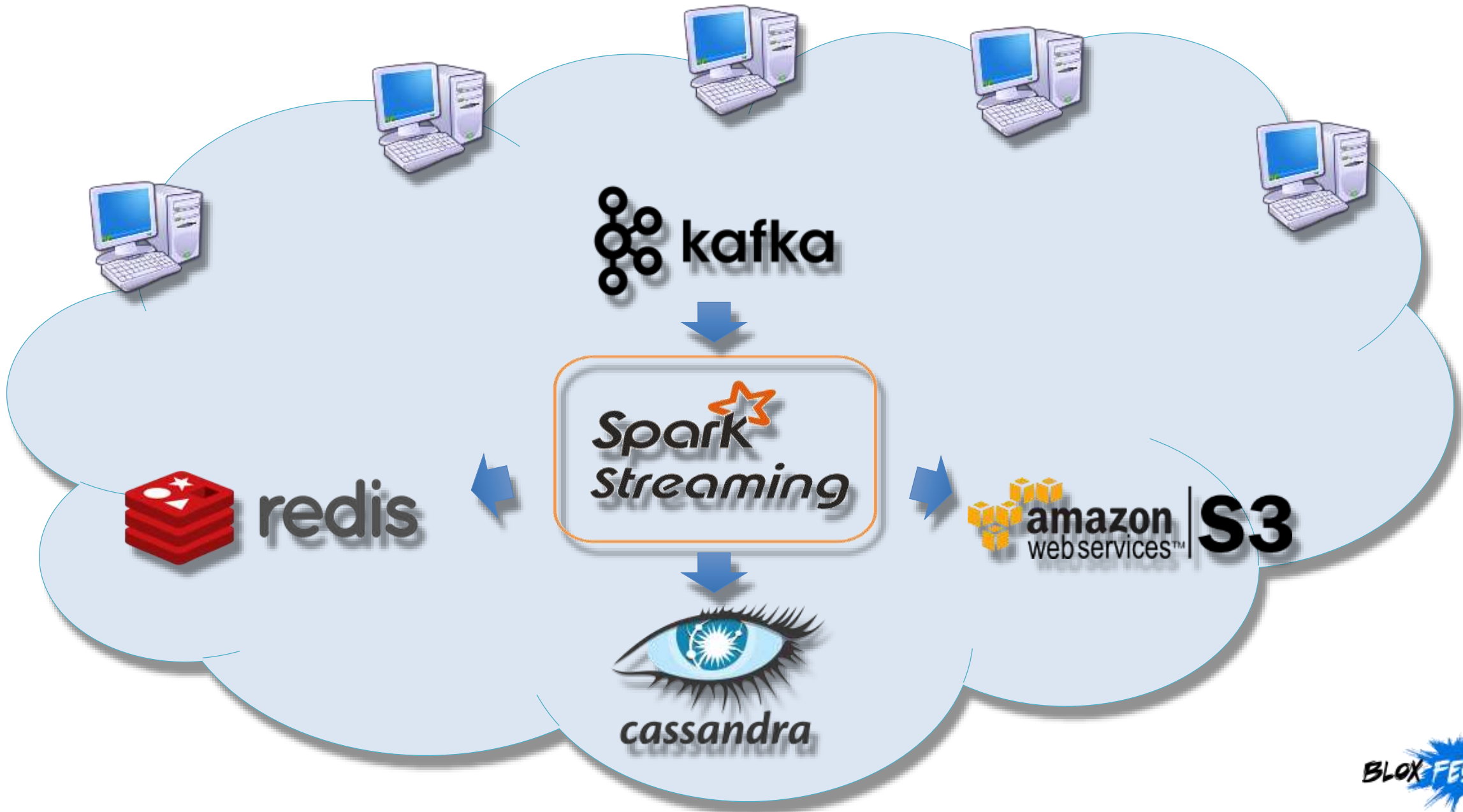
C221

C311



Big Data Architecture





Detection Results

Behavior Analysis based DNS Tunneling Detection and Classification with Big Data Technologies

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Keywords: Behaviour Analysis, Time Series, Big Data Analytics, DNS Security, Data Exfiltration, Anomaly Detection, Classification.

Abstract: Domain Name System (DNS) is ubiquitous in any network. DNS tunnelling is a technique to transfer data, convey messages or conduct TCP activities over DNS protocol that is typically not blocked or watched by security enforcement such as firewalls. As a technique, it can be utilized in many malicious ways which can compromise the security of a network by the activities of data exfiltration, cyber-espionage, and command and control. On the other side, it can also be used by legitimate users. The traditional methods may not be able to distinguish between legitimate and malicious uses even if they can detect the DNS tunnelling activities. We propose a behaviour analysis based method that can not only detect the DNS tunnelling, but also classify the activities in order to catch and block the malicious tunnelling traffic. The proposed method can achieve the scale of real-time detection on fast and large DNS data with the use of big data technologies in offline training and online detection systems.

1 INTRODUCTION

Domain Name System (DNS) that mainly services a domain name resolution to IP addresses on UDP is a service ubiquitous in every network. Because DNS is not intended for data transfer, people can overlook it as a threat for malicious communications or for data exfiltration. Most networks, public or private, do not firewall DNS traffic which creates security vulnerability. Tunnelling data over DNS or TCP over DNS is a technique that can be used as a way to circumvent access and security policies in firewalled networks. A typical example is to illegally browse the web through public hotspot while free service is not provided. There are many free software tools available for people of interest to setup a DNS tunnelling system quickly. One of the most popular tools is Iodine (Iodine). The fact that information bypasses a network first line security mechanism makes DNS tunnelling very attractive also in contexts other than free web browsing. Such examples include command and control and data exfiltration in cyber-espionage attacks in which it is fundamental for an attacker to have an available but inconspicuous communication channel.

DNS tunnelling works by encapsulating data into DNS packets. Typically, the tunnel client

encapsulates the data to be sent in a query for a specific domain name. The DNS resolver treats the tunnel traffic as a regular request by starting the lookup process for the requested domain name, possibly recursively consulting other DNS resolvers, as shown in Figure 1. At the end of this operation, the request is processed by the tunnel server. The server retrieves the encapsulated data and responds to DNS queries by enclosing tunnel data in the answer section of the DNS response message.

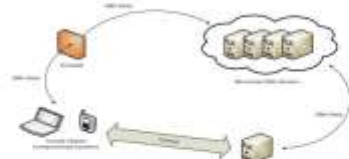


Figure 1: DNS tunnelling setup.

Although most DNS tunnelling techniques use TXT type queries in DNS that can maximize the payload in response packets, there are implementations that make use of DNS query types other than TXT such as A, AAAA, CNAME, NS,

Classification with Big Data Technologies

are, Gini index is another way to measure the diversity of the data that is defined as

$$G = 1 - \sum D^2(x).$$

The Gini index feature, Gini index, is bounded within a range

Error

To measure the diversity of a data set, we use the Gini index. Like the Gini index, this feature is also bounded. The definition is as follows.

$$G = 1 - \max\{D(x)\}.$$

Labels

The feature is the number of QDN payload named as $n_{i,j}$ and malicious payloads.

The output of a neural network is the output of a neural network of the above features as input, described in the next section.



Real-time Detection and Classification with Big Data Technologies

The data in this paper is collected from Farsight. It receives passive DNS from a large number of contributors worldwide, mainly in the US. Sample filtering logics such as DNS type, series length, and whitelisting, a set of extracted and reviewed by security analysts with labelling. About 2000 samples are used for training and testing a tree classifier that is used to minimize the false positive rate.

REAL-TIME DETECTION

The classifiers that were trained in offline system are deployed in an online real-time detection system (Farnham, 2014) that is designated to deal with streaming data. In an enterprise, the throughput can be up to 1-3 million per second. The throughput can reach second in a cloud based deployment. The horizontal scalability is one of the key factors in design.

In Figure 6, the incoming stream is processed in real-time with Storm or Spark and is inserted into the observation cache. The extracted features that are indexed by address and SLD. The observation cache is in-memory layer and an on-disk layer. The use is dependent on the data size. The cache can be triggered by event or scheduled by cron job.



Figure 3: Real-time detection system architecture.

FEATURES AND CONCLUSION

The DNS data collected from Farsight in 2013 at a rate of 1.8B/day is used in this process. In total, 126K tunnels are detected. A tunnel is defined as from one IP address to one unique destination. Table 1 shows the summary of the detected tunnels. About 70% detected tunnels are

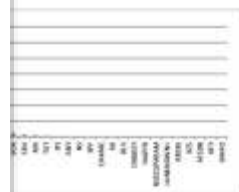


Figure 4: DNS query distribution by query type.

FEATURE EXTRACTION AND CLASSIFICATION

DNS traffic typically has very small payloads. The reason many approaches detect tunnels is on payload size (Farnham, 2013). However, when space and bandwidth are limited, more and more legitimate users are using domain names. Since the main goal of DNS tunnelling technique is to convey data in a way as efficient as possible, the tunnel in a way as efficient as possible. The tunnel in a way as efficient as possible. The tunnel in a way as efficient as possible.

Payload

There are several types of DNS queries. A tunnel will typically carry outbound payloads. The payloads are carried in many different ways. The payloads are carried in many different ways. The payloads are carried in many different ways. For example, the payload is encoded in the query type, the payload is encoded in the query type, the payload is encoded in the query type.

Common Features for Inbound and Outbound

The features common for both inbound and outbound traffic are extracted. Figure 3 shows the feature analysis results for inbound and outbound traffic.

a resolver or DNS server from the normal client IP address. The information availability is limited into an observation window. A TTL pre-set to remove old data points. It also has a capacity to hit the capacity though the TTL criterion. This is to avoid and reserve the storage space to be recycled. Applying the criteria of the messages within the window that is denoted as a 2-

$$f(x)$$

is the k^{th} feature on the i^{th} inbound payloads. The features are the basic features on the series that can

$$f(x)$$

is the collection of features across the time series. The copy values on effective payloads are calculated, the fact that the payload of change as much as the series.

CONCLUSION

The classification. In the first tier, the focus is on identifying encoded or is for tunnel detection.

ACKNOWLEDGMENT

The classifiers are designed and indicating if a payload is inbound and outbound traffic. The classification of the classifiers is files with truth labelled by independent sets of classifiers have a single feature and each uses a defined as follows.

$$f(x) = w_k + w_0$$

more malicious traffic. This makes it a focus for future research.

of DNS tunneling, network (SPF) for email, version 1, RFC6440. Semi-supervised the flux domain in the 10th IEEE/ACM Computing and Machine Learning (ICML) or DNS (EDNS0), RFC2671. Detecting covert channels in the 11th ACM Communications

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(Garelli, L., 2008, 'Across network the 4th IEEE/ACM Communications, Beijing

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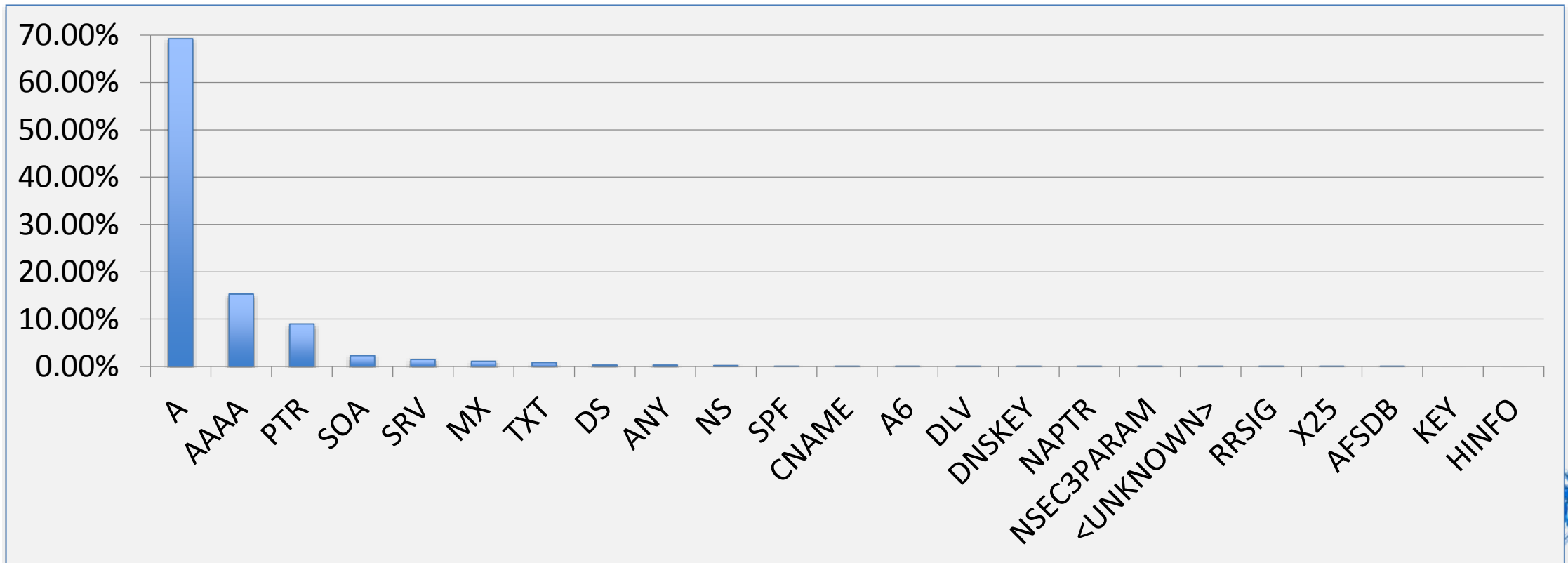
(Garelli, L., 2008, 'Flow-based flow Management Lecture Notes in Computer Science, 124-135. http://www.isc.

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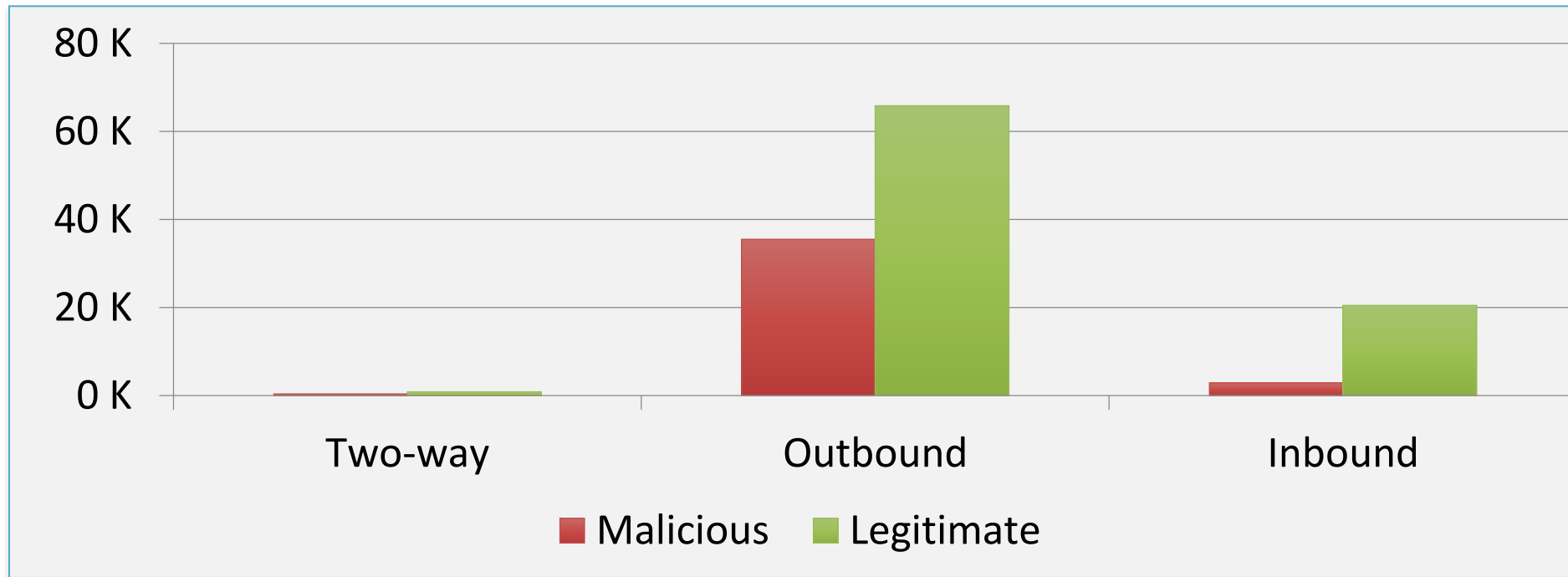
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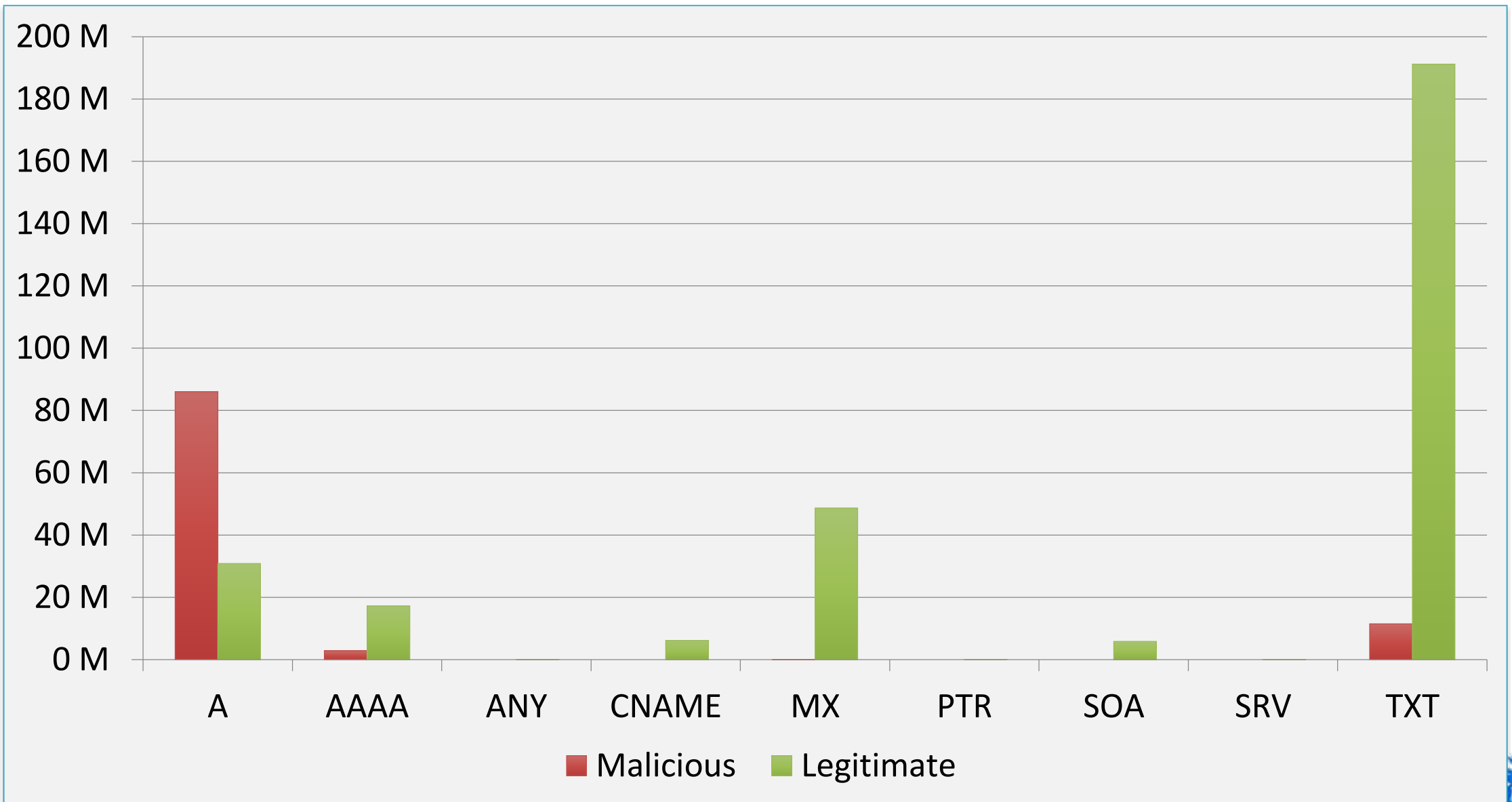


- Farsight (ISC)
- 2012.12 – 2013.08
- 1.8 billion per day
- 150 TB

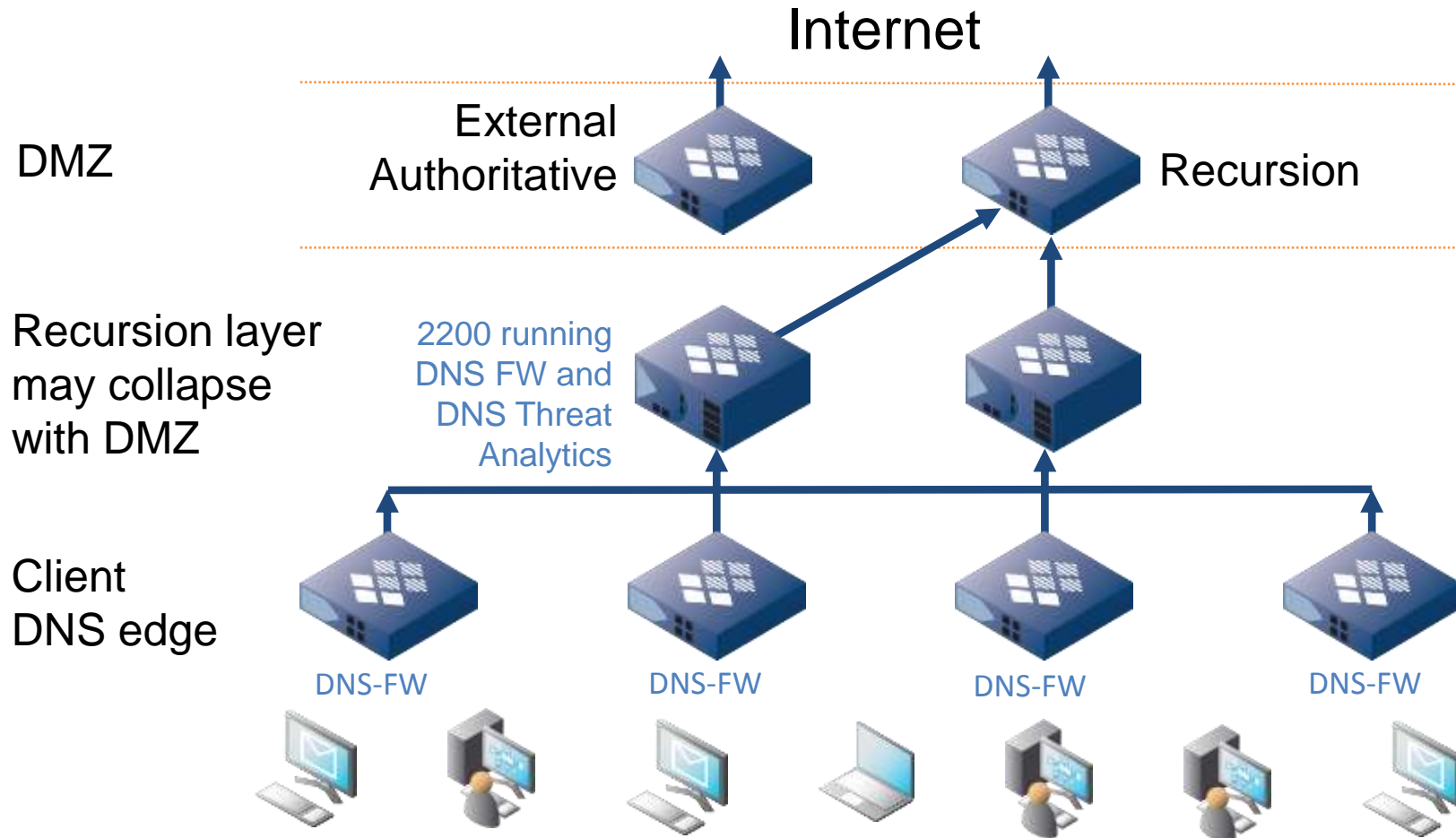


| | Malicious | Legitimate | All |
|----------|-----------|------------|---------|
| Two-way | 356 | 869 | 1,225 |
| Outbound | 35,478 | 65,820 | 101,298 |
| Inbound | 2,845 | 20,504 | 23,349 |
| Total | 38,678 | 87,193 | 125,871 |





DNS Threat Analytics in Recursion Layer



- Central detection of tunnels and data exfiltration
- Scaling of enforcement to all edge Grid members once destinations are on RPZ list
- Pinpoint infected systems at edge
- Lower platforms can be deployed at edge

Customer Case Study:

*T. J. Short, CISO and VP of Infrastructure,
Everi Holdings, Inc.*

“The attackers are getting smarter every day. They’re getting new tools, new ideas, new concepts. So we have to have defenses that are leading edge, that can change, adapt, and update very quickly. Infoblox Internal DNS Security does that.”



Questions?

