

Prediction of security policies for network

DD  S

Team Introduction

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Company Introduction

IMPERVA

Solutions that protect users
against cyber attacks



**Application
Security**



**Data
Security**



**Network
Security**



**Application
Performance**

6000+ Customers

Understanding our domain

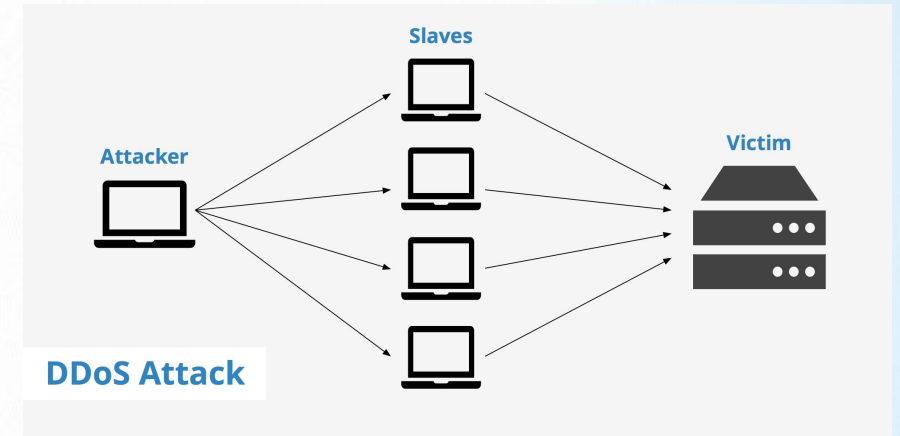
What is DDoS?

A distributed denial of service attack, overwhelming a server, service, or network with more data than it can handle.

IP & IP Range

Start IP: 192.168.0. 1

End IP: 192.168.0. 254



Vectors of attacks:

TCP

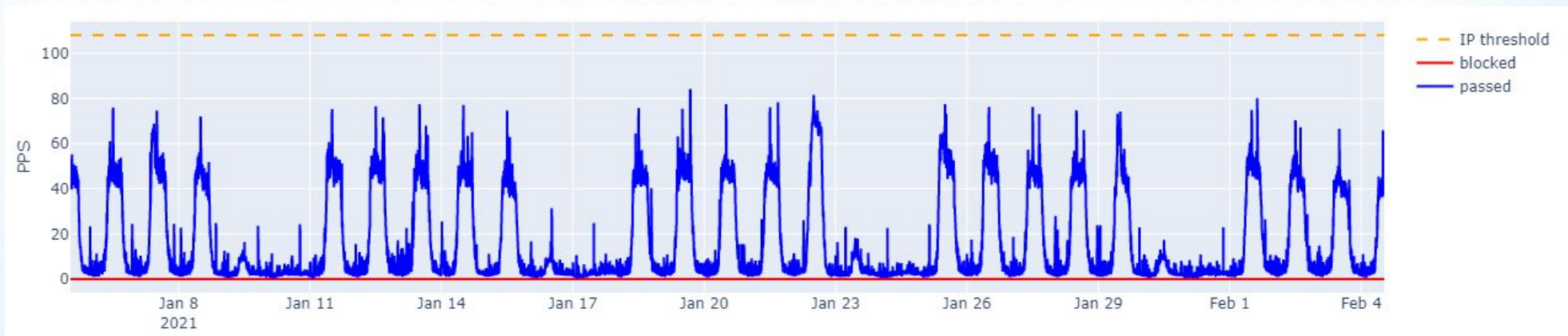
UDP

The largest attack mitigated by Imperva ~ 1,37 Tbs

Understanding our domain

What is a Network security policy?

It is a set of thresholds, each of them activates a set of rules to handle the traffic.



Problem statement

Why do we have to mitigate DDOS attacks?

To ensure business continuity, guarantee uptime and no performance impact

\$300K is the average cost of 1 hour of downtime

Why do we have to update security policies?

“Normal” amount of traffic changes over time

Why do we have to automate creating policies?

It is a human and time consuming task, people make mistakes

Manual creation doesn't scale anymore

Project goals

Main

Create a model for complete network security policy

Project impact

Enable Imperva to set up-to-date, more accurate and personalized security policies

Data overview. Target

2 Metrics:

PPS (packets per second)
BW (MB per second)

13 Vectors:

TCP, UDP, DNS, NTP, SYN
NETFLOW, etc.

8 Thresholds:

IP Level
IP Range level

Complete policy - 208 values

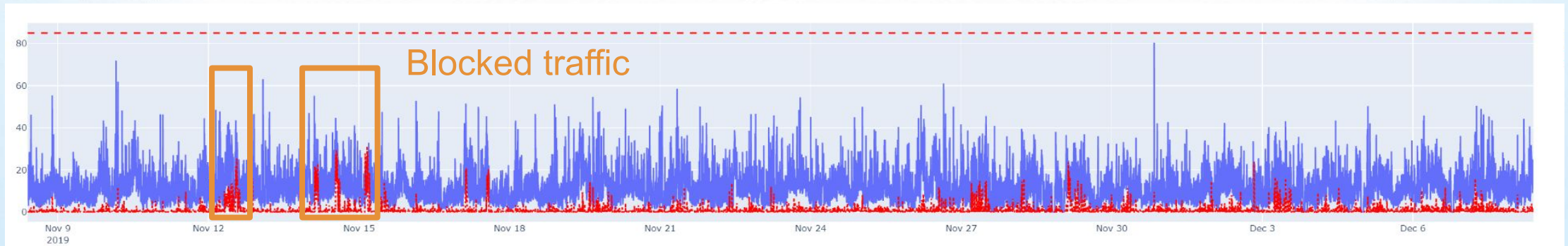
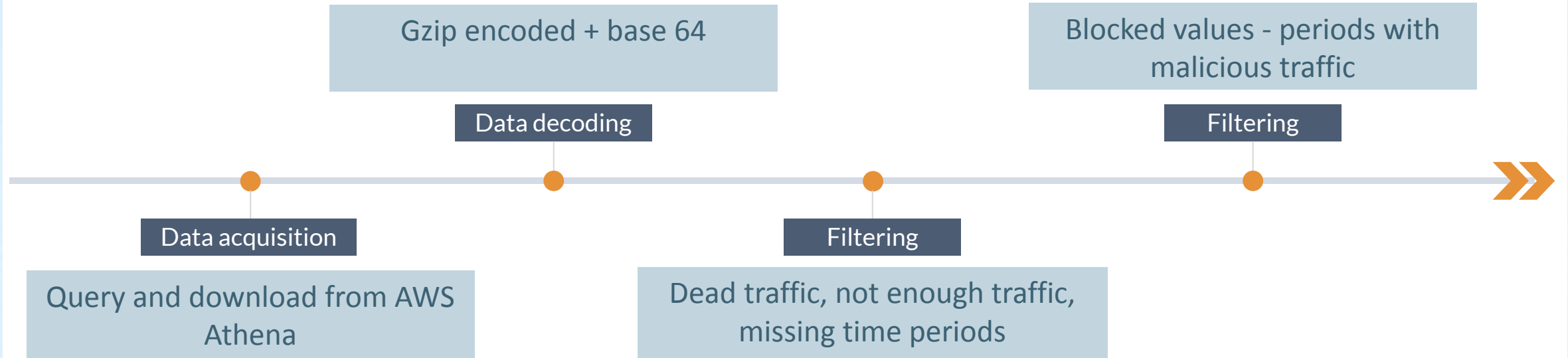


Narrowed down to TCP - PPS

Data overview. Input

Data format	Parquet files	Encoded data
Size	133 000 samples	71 GB
Time series sample	1 minute intervals	max mean 30 days per sample
Metadata	Comments	Reasons

Data overview. Preprocessing



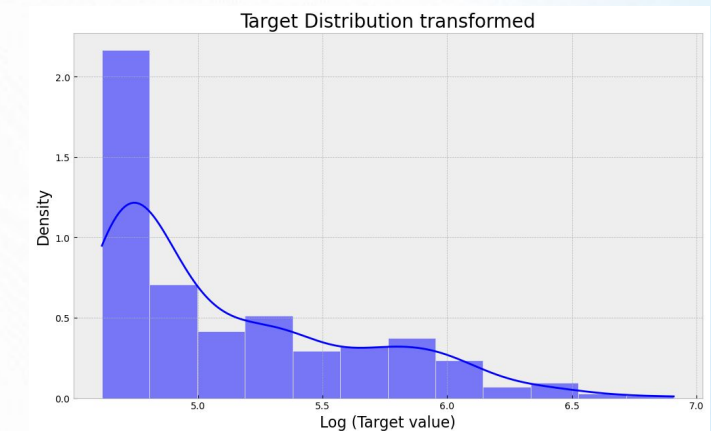
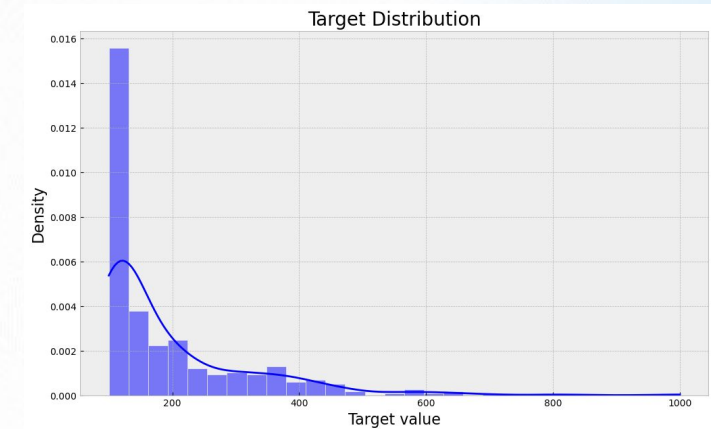
Our approach. Filtering and transformation

Filtering
samples by:

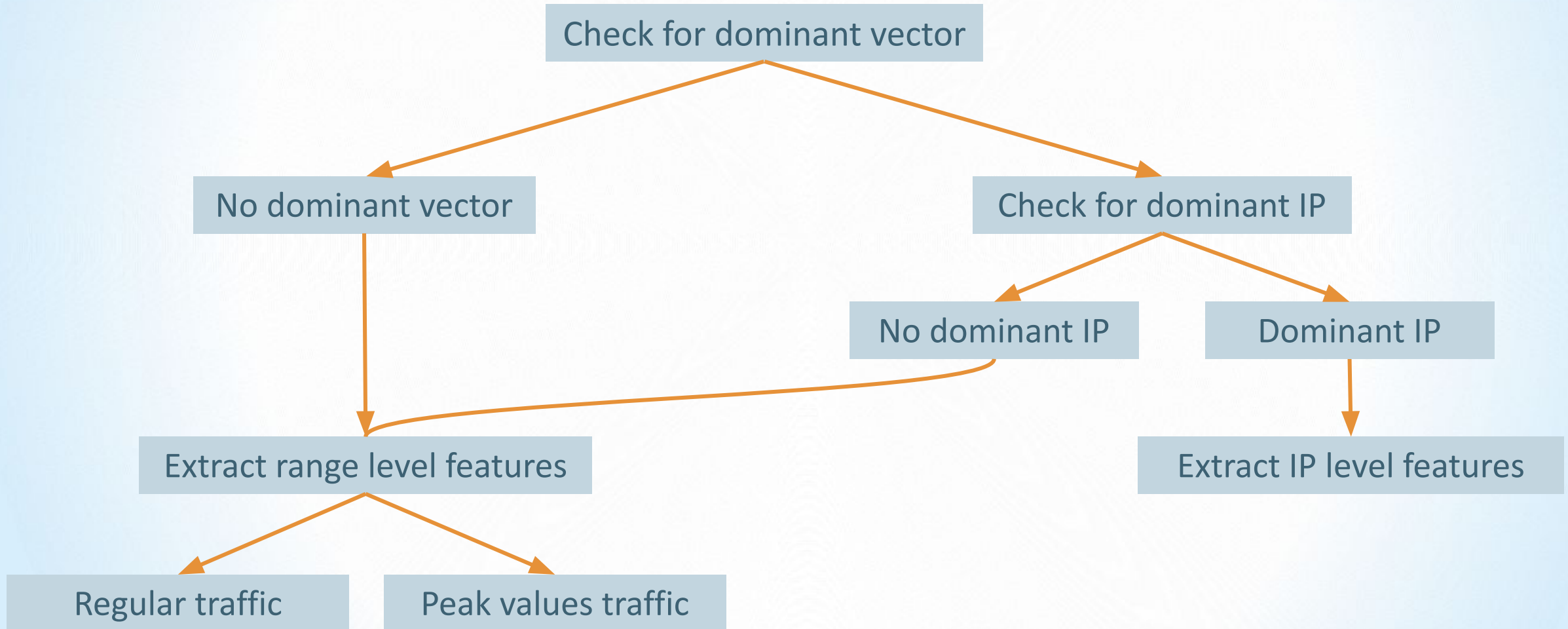
- Target values - constant/ default values, particular ranges
- Comments/ reasons
- Irrelevant policies/ edge cases

Transformation:

- Target distribution has long and light tail - used log-transform



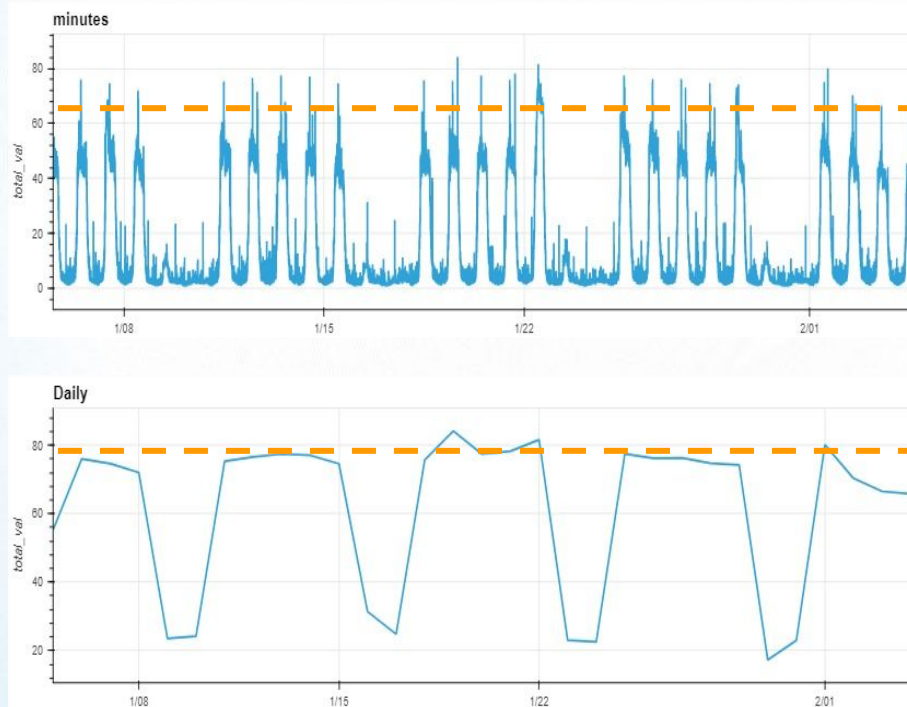
Our approach. Feature extraction



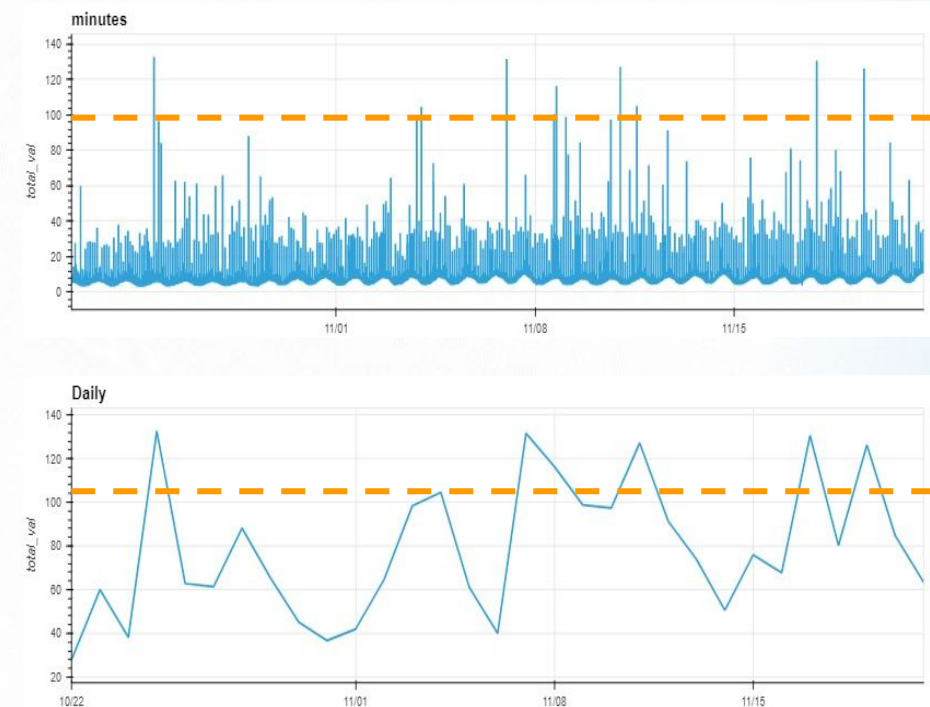
Our approach. Feature extraction

Patterns of traffic:

“Regular”



“Irregular”

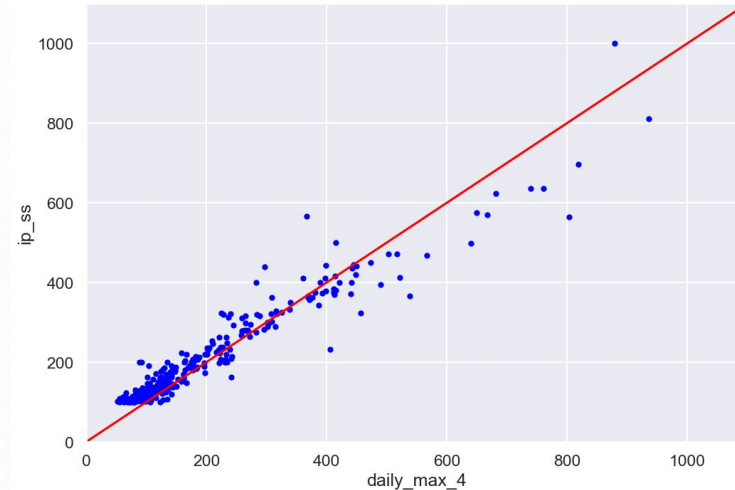


Daily Max 90th quantile — — — — —

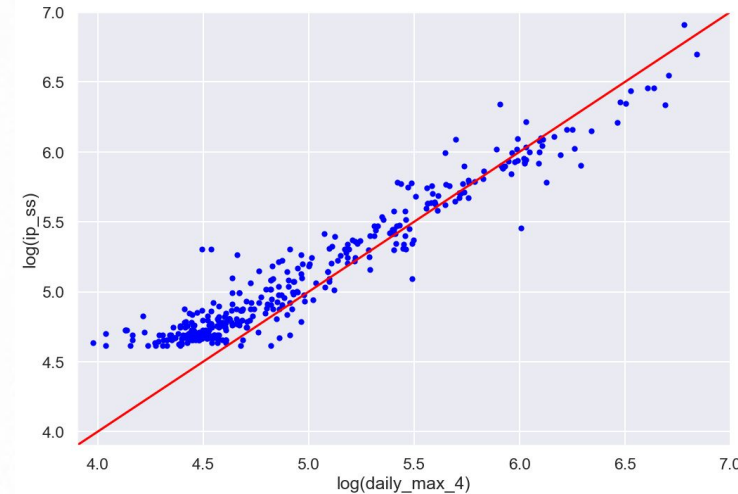
Our approach. Features

	ip_ss
ip_ss	1.00
daily_max_q90	0.97
daily_max_4	0.97
daily_max_3	0.96
daily_max_2	0.93
daily_max_q70	0.91
daily_max_1	0.78
daily_mean_q90	0.73
daily_q90_q90	0.73
daily_mean_q70	0.70
daily_q90_q70	0.69
daily_median_q90	0.69
daily_median_q70	0.67

Initial



Log-transformed



Our approach. Training

Scalers:

No Scaling
MinMaxScaler
StandardScaler
RobustScaler

Linear models:

Linear Regression
Ridge
Lasso
SVR

Grid search:

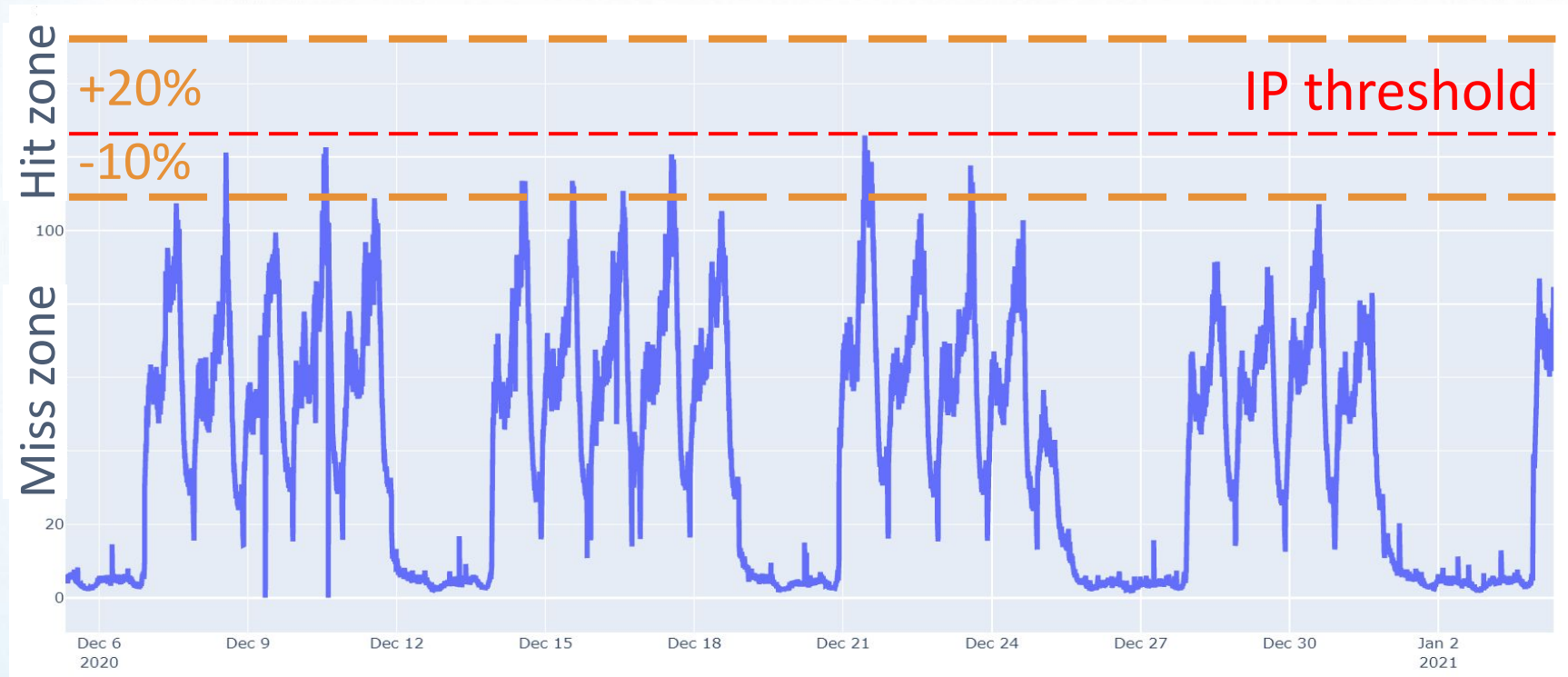
model__alpha
model__kernel
model__C



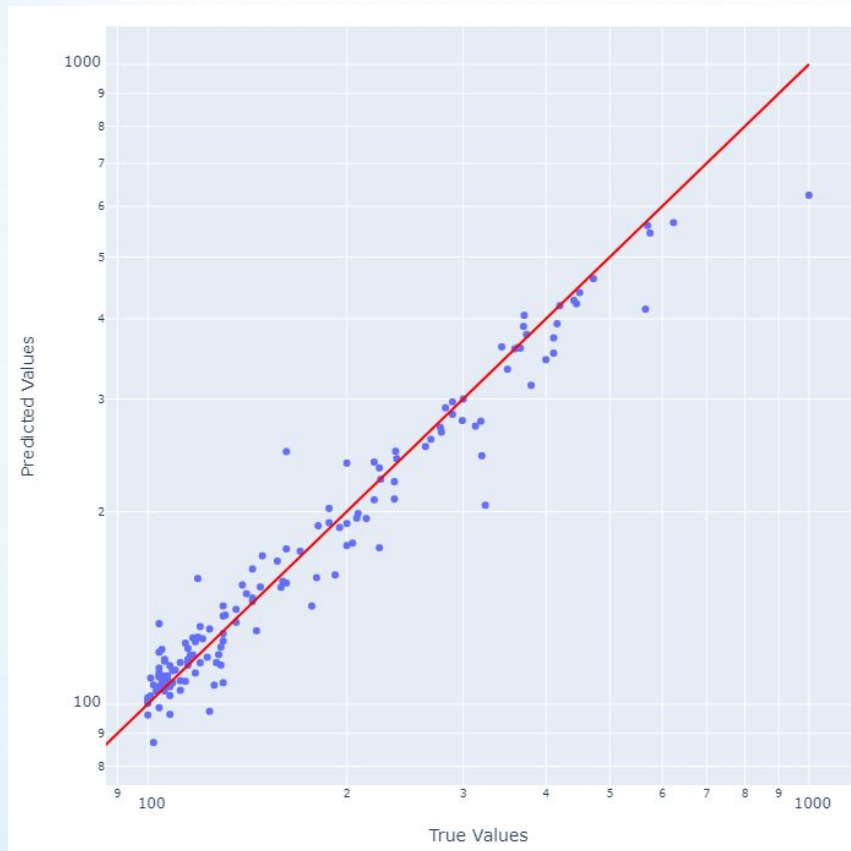
Tree-based models:

Random forest
XGBoost
CatBoost

Metrics and evaluation



Project results



MinMaxScaler + SVR + symmetric threshold
accuracy: 0.83 ± 0.05

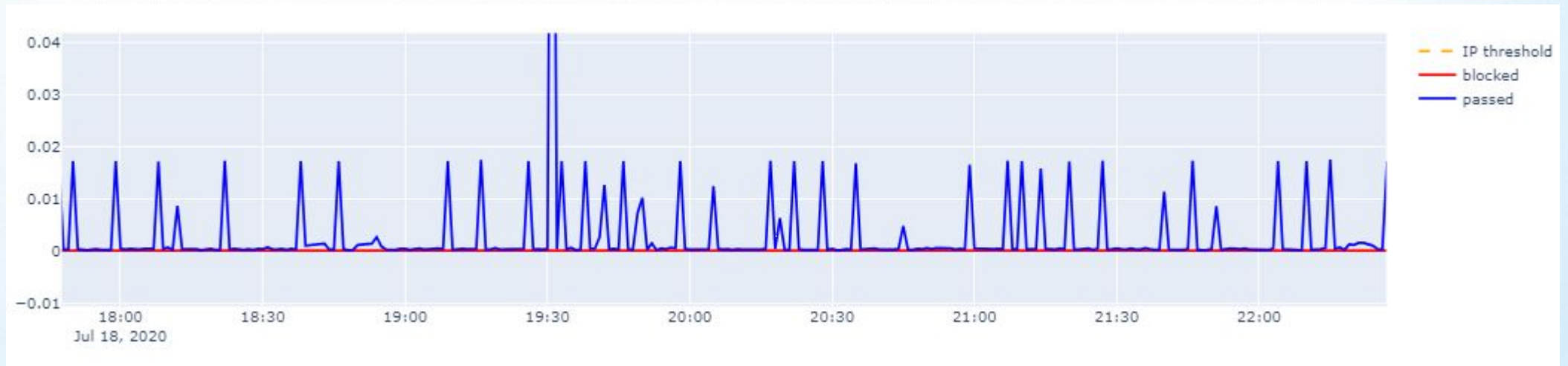
MinMaxScaler + SVR + asymmetric threshold
accuracy: 0.85 ± 0.03

Expected accuracy was ~0.9

Future work

Advanced goal:

Detect port scanning operations and handle them, so the model for security policies won't be affected



THANK YOU!

СПАСИБО!

TODA RABA!