



Deflect Labs and Baskerville – Identifying attacks with Machine Learning

Introduction - Why?



The need

 Manual identification and mitigation of (DDoS) attacks on websites is a difficult and timeconsuming task with many challenges

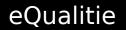
The goal

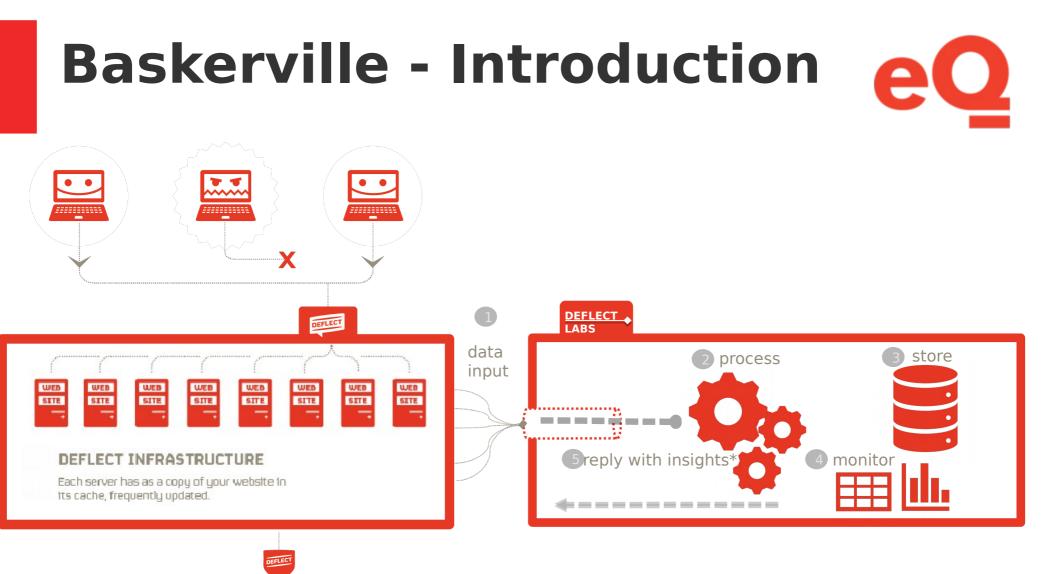
 Create a system to identify the attacks directed to Deflect protected websites as they happen and give the infrastructure the time to respond properly.

Introduction - The challenges



- Be fast enough to make it count
- Be able to adapt to traffic
- Provide actionable info
- Provide reliable predictions
- As with any ML project: not enough labelled data
- Make the system as generic as possible







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Baskerville -Architecture

Two main Components

Engine

Main log processing Pipelines

Off-line analysis tools

- Model development
- Visualization
- Investigation

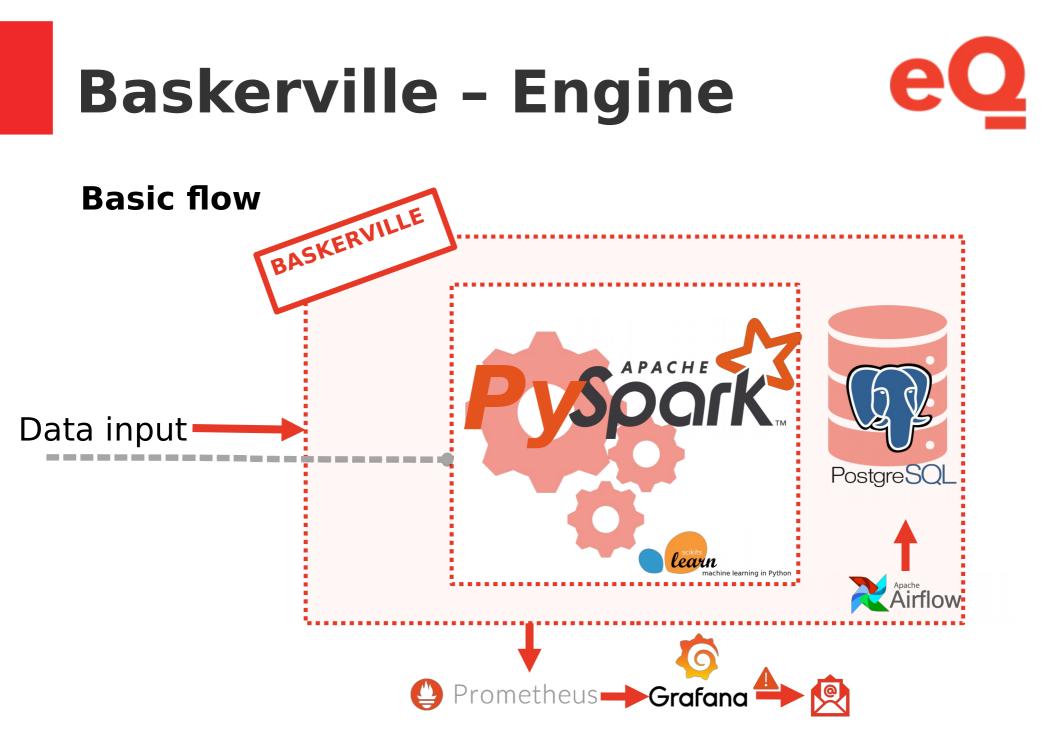
Time-windowed – batch processing



Baskerville - Engine: Pipelines

Pipelines

- Base Pipeline for the basic flow
- Elastic Search Pipeline to process logs from ES
- Raw Logs Pipeline to process log files
- Kafka Pipeline to process logs from Kafka

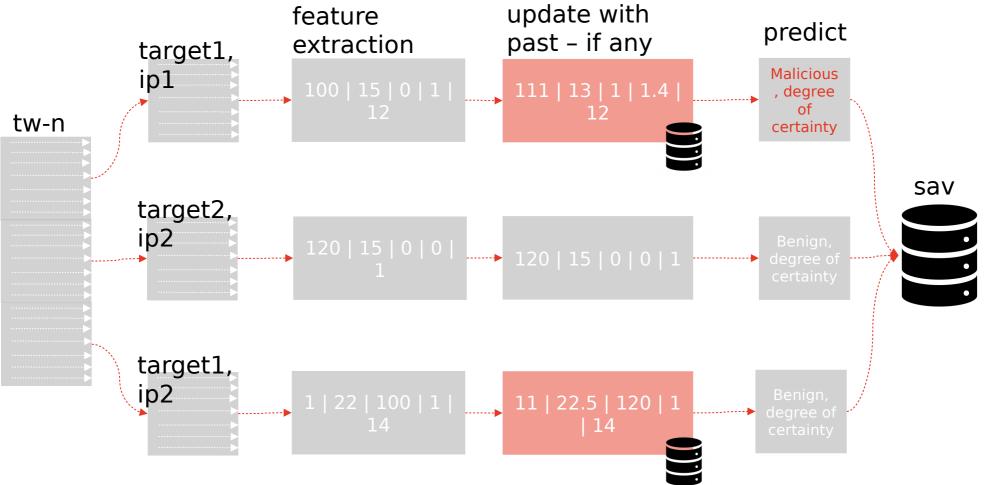


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Baskerville - Engine: Pipelines



Basic flow



Baskerville - Features



Features and structure

- Css to html ratio
- Image to html ratio
- Js to html ratio
- Minutes total
- Path depth average
- Path depth variance
- Payload size average
- Payload size log average
- Request interval average
- Request interval variance
- Request total
- Response 4xx to request ratio
- Top page to request ratio
- Unique path rate
- Unique path to request ratio
- Unique query rate
- Unique query to unique path ratio
- Unique UA rate
- ...

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Most of them are Updateable Features – they take past into account

compute update

Baskerville - Features



Cache and Feature update - Taking past into account

- Keep track of one week of traffic at any point
- Two level cache short-term (in memory) and long-term (parquet)

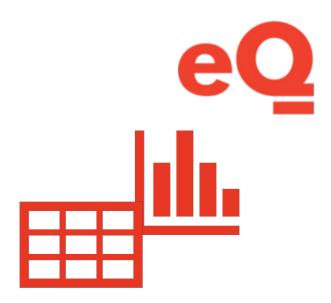
Baskerville - Database

- ORM SQLAlchemy
- Only inserts, no updates except for the offline tools
- Data partitioning: per week of year
- Data archive / retention policy: keep data for one year
- Airflow to coordinate and schedule the database maintenance, create new partitions, detach old ones, archive.



Baskerville -Monitoring

- Prometheus with exporters for:
 - Baskerville itself
 - Spark
 - Postgres
 - Kafka
 - and Prometheus and Grafana of course
- Grafana for visualization

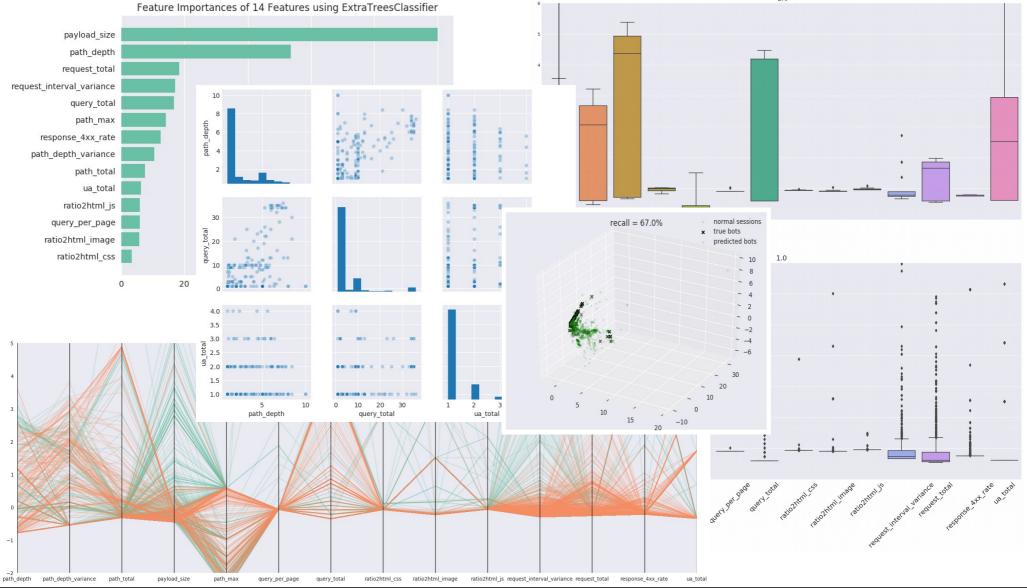


Baskerville – Offline tools Offline Analysis Tools Model development and investigations



- Preprocess historic logs Elastic Search / log files
- Preprocessing: Label training / testing sessions based on MISP database of attacks
- **Training**: Train novelty detection classifier on labelled sessions from historic logs
- Predicting: Classify sessions as malicious / benign using newly trained models
- **Clustering**: Group sessions based on their features to investigate botnets
- Visualization: Produce figures to aid model development and investigations

Baskerville - Offline tools



eQ

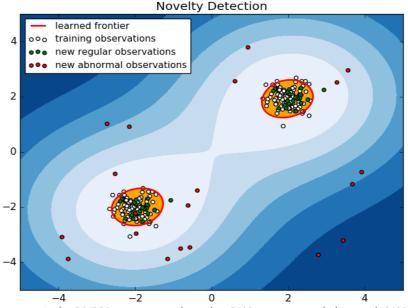
-1.0

Baskerville - Machine Learning



How we use Machine Learning:

- Novelty / Outlier
 detection algorithms One 2
 Class SVM, Isolation Forest
- Lots of normal data
- Train on normal data allowed to include a small amount of abnormal data
- Test on known past attacks
- Cross-reference results with Banjax bans



error train: 21/200 ; errors novel regular: 2/40 ; errors novel abnormal: 1/40 $\,$

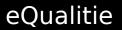
Baskerville - Machine Learning Processing the Attacks against Vietnamese Civil Society

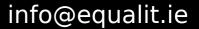
- 8 attacks in total considered
- Processed attack periods and normal traffic (separately) with Baskerville
- Train dataset: normal traffic
- **Test dataset**: a combination of normal and abnormal traffic

Isolation Forest

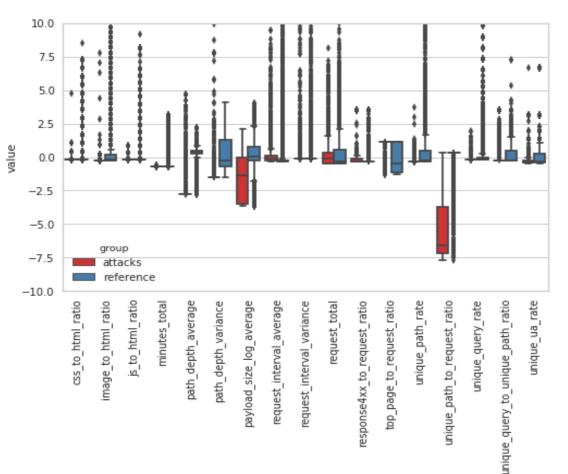
Precisi	0.90
on	
Recall	0.86
F1	0.88
score	

- 90% of the IPs predicted as anomalous by Baskerville were also flagged by Banjax as malicious
- 88% of all the IPs flagged by Banjax as malicious were also identified as anomalous by Baskerville





Baskerville - Machine Learning Processing the Attacks against Vietnamese Civil Society



feature

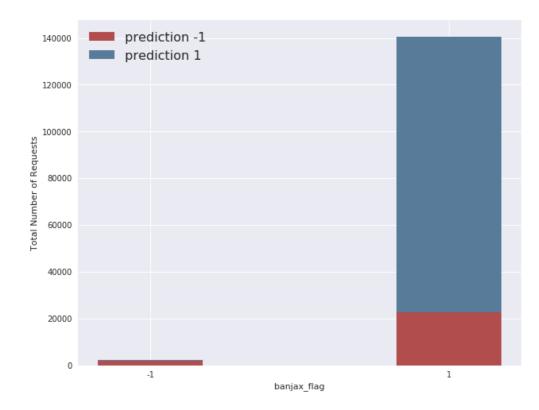
Attack characteristics

- far fewer unique paths requested
- a shorter average path depth
- a smaller variance in the depth of paths requested
- a lower payload size

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Baskerville – Machine Learning Processing the Attacks against Vietnamese Civil Society

The need for a feedback mechanism

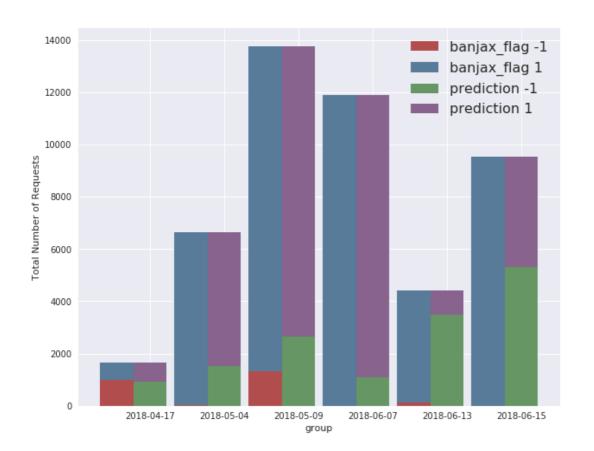


The overlap between the Banjax flag and the Baskerville prediction

-1 indicates malicious+1 indicates benign

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Baskerville – Machine Learning Processing the Attacks against Vietnamese Civil Society The need for a feedback mechanism



Baskerville is picking up more request sets as malicious than Banjax ...

but does this indicate that Baskerville is too sensitive to anomalous behaviour, or that Baskerville is outperforming Banjax?

Read the report comparing human analysis vs baskerville

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Baskerville - Machine Learning

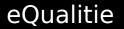


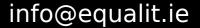
What about Bias?

Perhaps there is a chance that people with slower internet connections display different than average browsing behaviour. So we need to be **careful feature selection** and **a good training dataset** with lots of examples of browsing.

To conclude

 So in general we are doing well, but there is a long way to go to properly evaluate the system and make sure we have low false positives and false negatives





Baskerville – How do I use it?



- Just like any other Python project, download, pip install it, configure it and run it
- Docker compose for Baskerville itself and the peripheral components like Postgres, Kafka, Prometheus, Grafana and various exporters
- Script for setting up a stand-alone single node Spark cluster set up (kubernetes spark integration is still experimental in spark 2.4.0)

Baskerville - Extensibility eQ & Use cases

From a user's perspective:

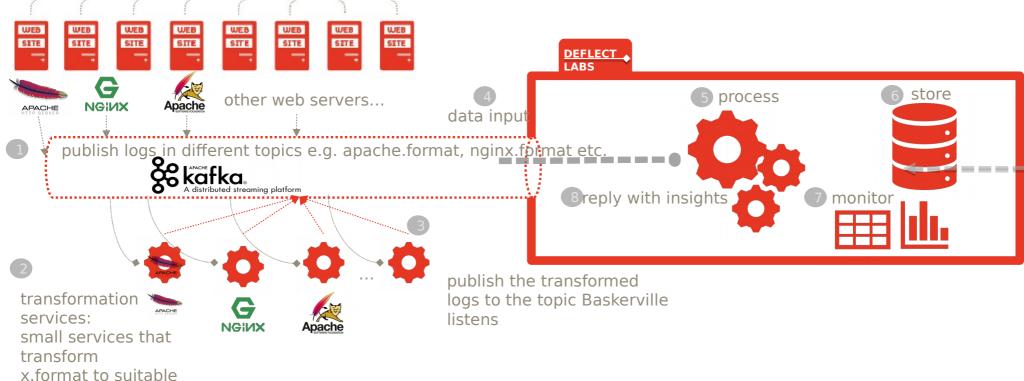
- You can set it up as a stand-alone analysis engine to process and analyze old logs in various forms and formats:
 - text files
 - Elastic Search
 - through a queue / adapter
- Train models on own logs

From a developer's perspective:

- Easy to enable / disable features or create new ones
- Easy to extend the pipelines or add new

What is Baskerville -Use cases





Baskerville input

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Baskerville - Current state eQ

Dev deployment

- Consuming about **10%** of average daily traffic (~3M requests)
- Multiplied by 10 on the Baskerville end to simulate the actual traffic
- Running on a **standalone single node spark cluster** with 1 worker, 4 executors with 2 cores and 6GB RAM each.
- **Performance tuning**: Java GC, Spark parameters, worker numbers, database tuning ...
- About **30M requests per day**
- Processing the data in a 2-minute time window within ~30sec about ¼ of the time window

Model Development

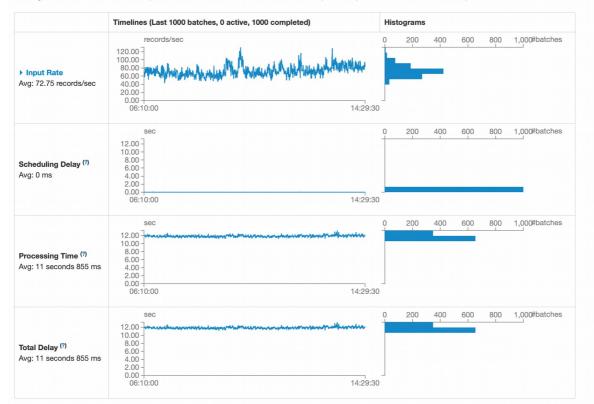
- Feature selection, hyper-parameter optimization (Grid search)
- Training / Testing gathering datasets and attacks

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Baskerville - Current state eQ

Streaming Statistics

Running batches of 30 seconds for 1 day 16 hours 29 minutes since 2019/01/30 22:00:07 (4859 completed batches, 9692125 records)



Active Batches (0)

- Completed Batches (last 1000 out of 4859)

Batch Time	Records	Scheduling Delay ^(?)	Processing Time ^(?)	Total Delay ^(?)
2019/02/01 14:29:30	2481 records	1 ms	12 s	12 s
2019/02/01 14:29:00	2728 records	0 ms	12 s	12 s
2019/02/01 14:28:30	2842 records	0 ms	12 s	12 s
2019/02/01 14:28:00	2545 records	0 ms	12 s	12 s
2019/02/01 14:27:30	2272 records	0 ms	12 s	12 s

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Baskerville: The next steps

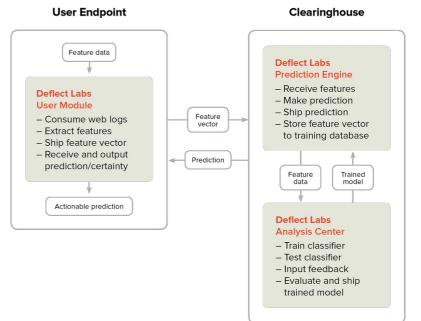


- Deployment within April Probation period
- Baskerville Dashboard
- On-going model development
- Feedback mechanism
- Release and Open Source by the end of Q2
- Create the challenge / ban communication component
- Work on k8s deployment for the spark cluster*

Baskerville - The next steps

Further down the road: Information Sharing and Analysis Center (ISAC)

- Clients run part of Baskerville: the processing engine
- Prediction with the degree of confidence is served by ISAC
- Clients chose to use or not the prediction, e.g. ban or serve a challenge to the IP with the potentially malicious intent



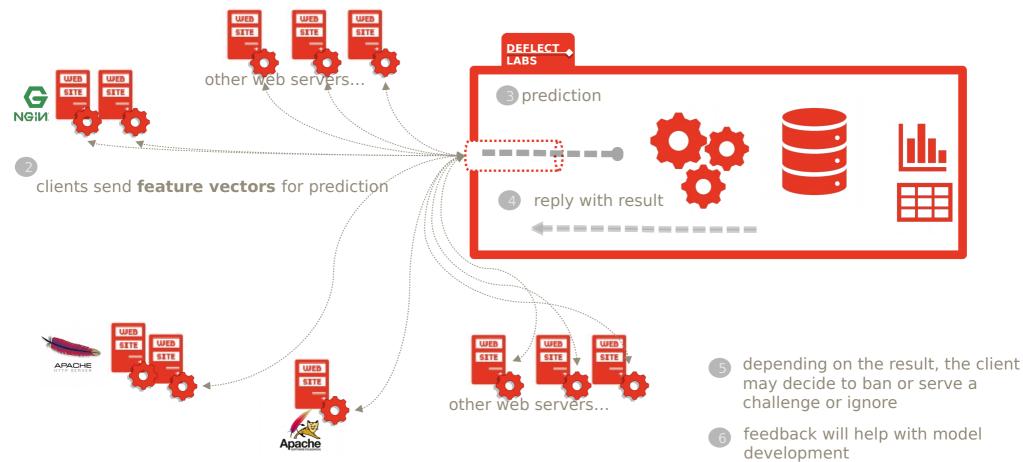
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Baskerville: The next steps



Further down the road: ISAC

processing takes place at each client



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Baskerville: The next steps



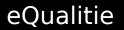
Further down the road: ISAC

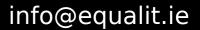
Pros

- No sensitive data sharing just the feature vectors are enough
- With a **feedback mechanism** we will be able to expand the training dataset and improve the model

Cons & Challenges

- The size of the infrastructure to be fast enough
- Convert part of Baskerville the processing engine to a "plugin" that can be used on the client side





Any Questions?



Website: https://docs.deflect.ca

GitHub: https://deflectca.github.io

- Twitter: @equalitie
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- Public launch: Fall 2019



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