Network attack detection

Machine Learning course @ SUPSI

Project goal:

Detect network attacks using machine learning techniques

Dataset

Data format: NetFlow Version 9 Flow-Record (by Cisco)

• Two files (total size = 1.01 GB)

• Train set: ~4 mil. packets

• Test set: ~2 mil. packets

14'066 unique hosts

6187 unique hosts

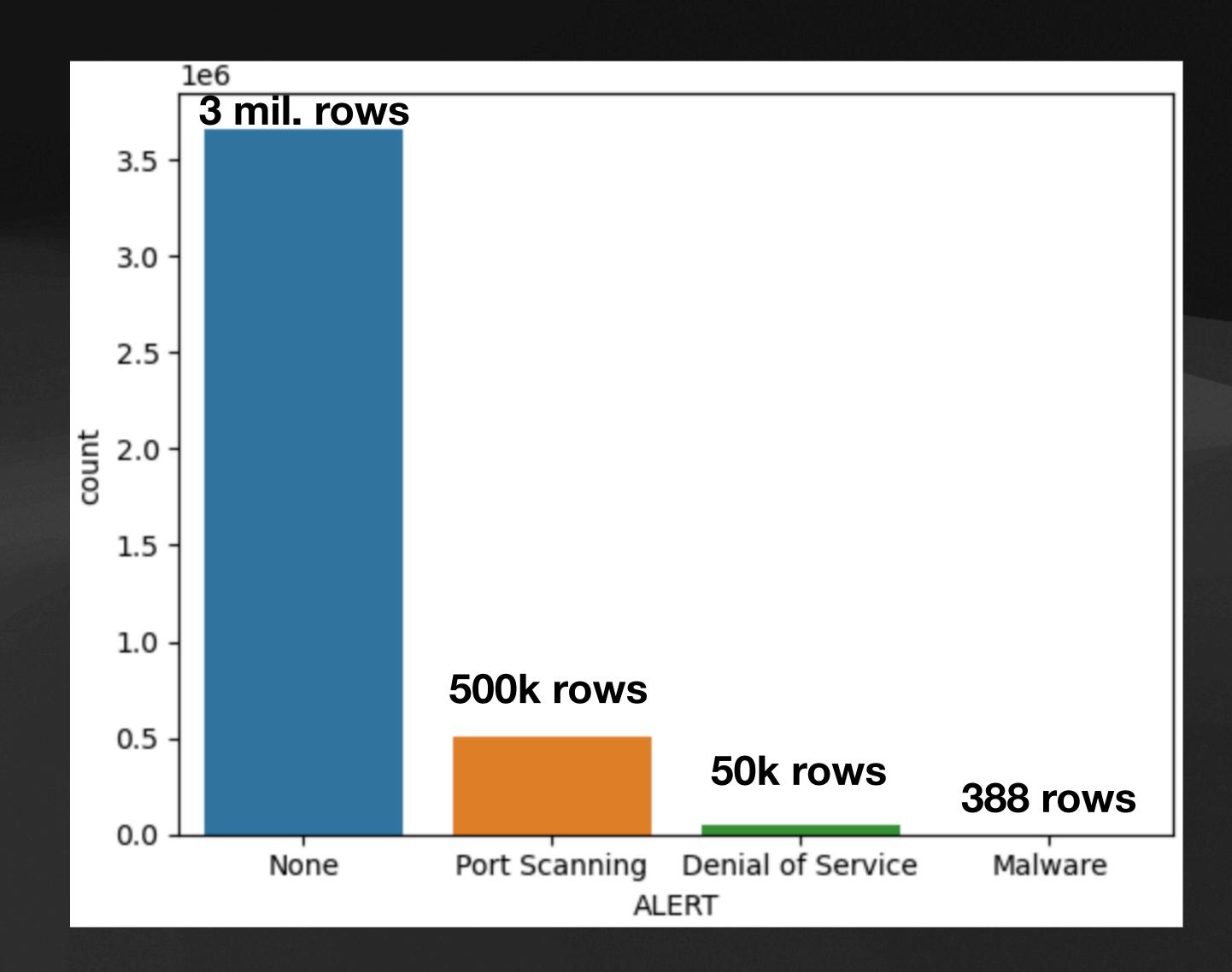
• 32 features:

- Protocol, Source/Dest. Ports, Packet size, etc. (View Cisco documentation)
- ALERT (target) = DoS, Port Scanning, Malware, None

Dataset problems

- Known problems:
 - Highly imbalanced dataset
 - Dataset too big
 - Useless features
 - Only classification

- Solutions:
 - Dataset sampling
 - StatifiedShuffleSplit



Train & Validation sets

(with StratifiedShuffleSplit)

- Dataset sampling
 - Train set = \sim 120k rows
 - Validation set = ~ 60k rows
 - (dev = 3% of the total size)

Same target distribution

```
Train set distribution:
                     0.868508
None
Port Scanning
                     0.119380
Denial of Service
                     0.011983
Malware
                     0.000128
Name: ALERT, dtype: float64
Validation set distribution:
                     0.868490
None
Port Scanning
                     0.119379
Denial of Service
                     0.011973
Malware
                     0.000158
Name: ALERT, dtype: float64
```

Feature selection

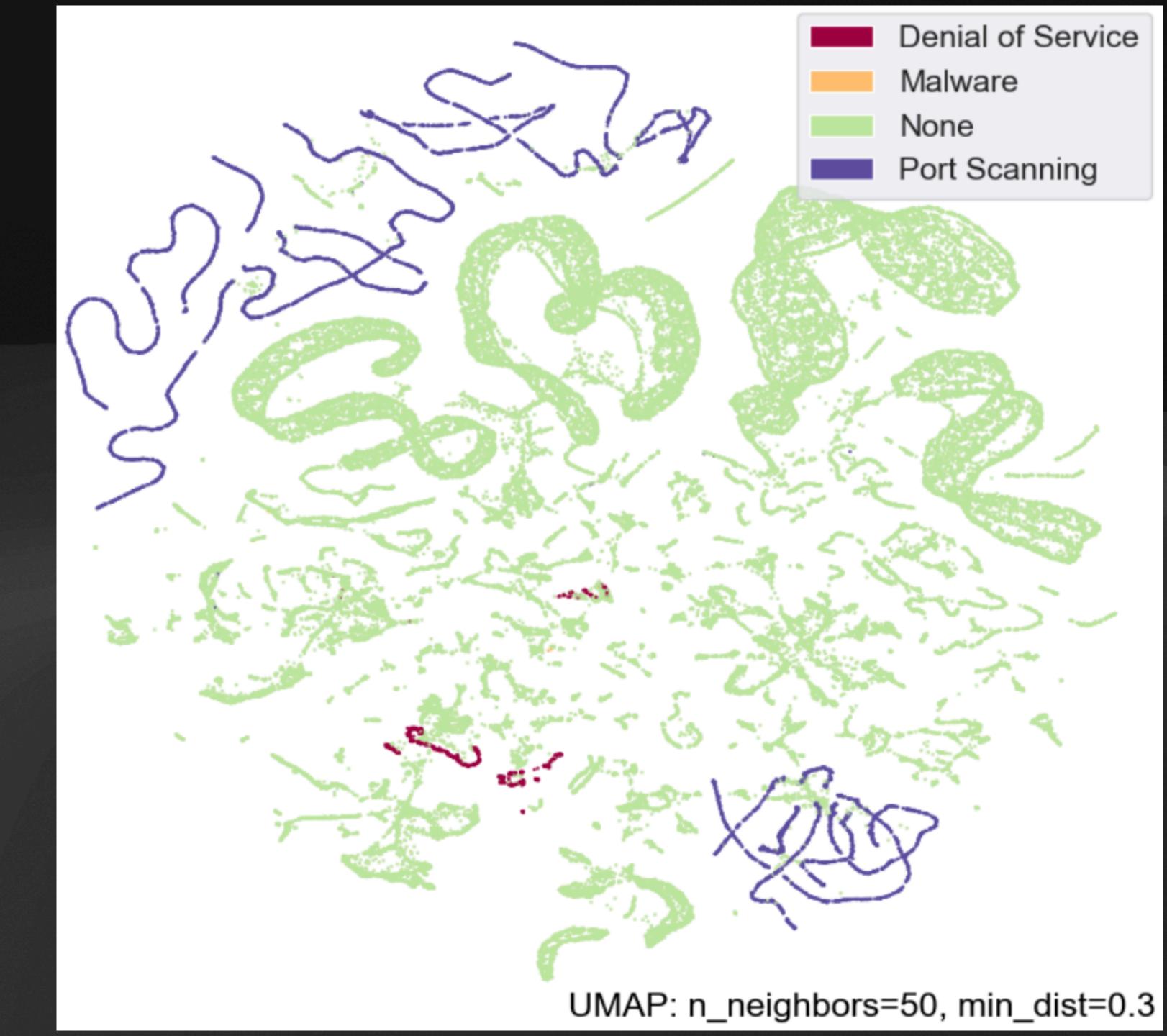
Many features, but which ones are important?

- Correlation matrix: a complete mess
 - By hand, visual method
 - Any correlation to malware

- Random Forest Regressor:
 - Better results!
 - Dynamic, with threshold (> 0.01)

		importance
IN_BYTES		0.182673
TCP_WIN_MSS_	IN	0.128970
ANOMALY		0.122406
TCP_WIN_MAX_	IN	0.099204
L4_DST_PORT		0.079617
TCP_WIN_MIN_	IN	0.074066
OUT_BYTES		0.039926
TCP_FLAGS		0.036853
TOTAL_FLOWS_	EXP	0.036217
LAST_SWITCHE	D	0.033888
EIOW DIIDATTO	N WILLIECUNDS	A AZAKR1

UMAP visualisation



Classification: KNN + SVC + Bagging

(predictions and evaluation on validation set)

K-Nearest Neighbour

K = 3

	precision	recall	f1-score	support
Denial of Servi	ce 0.94	1.00	0.97	303
Malwa	re 1.00	1.00	1.00	4
No	ne 1.00	1.00	1.00	21978
Port Scanni	ng 0.99	1.00	0.99	3021
accura	су		1.00	25306
macro a	vg 0.98	1.00	0.99	25306
weighted a	vg 1.00	1.00	1.00	25306

Cross validation (5 folds) score: 0.997

Denial of Service	303	0	0	0	- 20000
Malware	0	4	0	0	- 15000
None	18	0	21918	42	- 10000
Port Scan	0	0	0	3021	- 5000
	Denial of Service	Malware	None	Port Scan	- 0

Support Vector Classificator (SVC) Pipeline: PCA + SVC

PCA (10 components) - SVC (kernel=RBF, C=100 and Gamma=0.1)

	precision	recall	f1-score	support
Denial of Service	0.99	1.00	1.00	303
Malware	1.00	0.75	0.86	4
None	1.00	1.00	1.00	21978
Port Scanning	0.99	1.00	1.00	3021
accuracy			1.00	25306
macro avg	1.00	0.94	0.96	25306
weighted avg	1.00	1.00	1.00	25306

Voting classifier

BaggingClassifier + 30x SVC estimators

	precision	recall	f1-score	support
Denial of Service	0.53	1.00	0.70	303
Malware	1.00	1.00	1.00	4
None	1.00	0.98	0.99	21978
Port Scanning	0.97	1.00	0.98	3021
accuracy			0.99	25306
macro avg	0.87	1.00	0.92	25306
weighted avg	0.99	0.99	0.99	25306

Why this precision?

Sources

- Cisco NetFlow v9 format
 - https://www.cisco.com/en/US/technologies/tk648/tk362/technologies_white_paper09186a00800a3db9.html
- Dataset source
 - https://www.kaggle.com/datasets/ashtcoder/network-data-schema-in-the-netflow-v9-format
- SciKit learn StratifiedShuffleSplit
 - https://scikit-learn.org/stable/modules/generated/sklearn.model_selection.StratifiedShuffleSplit.html
- SciKit learn BaggingClassifier
 - https://scikit-learn.org/stable/modules/generated/sklearn.ensemble.BaggingClassifier.html