

# Data Science for Cybersecurity Hand-on Lab

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# Data Science Tools

# Data Science Tools

- **No-code and low-code**

- Excel (spreadsheets and macros)
- Orange (drag-and-drop machine learning)

- **Code-based tools**

- Python (general language, many data science libraries)
- R (statistics)
- MATLAB (engineering)

- **Many other options**



# The BETH Dataset

# BETH Dataset

**BETH\* is a real cybersecurity dataset published in 2021 as a benchmark for anomaly detection researchers**

- 8 million records, generated by 23 hosts, during 5 discontinuous hours
- Each host includes benign traffic as well as at most one single attack
- Each record is labeled as to whether it is “benign” or “malicious”

\*Highnam, K., Arulkumaran, K., Hanif, Z., & Jennings, N. R. (2021). "BETH dataset: Real Cybersecurity Data for Anomaly Detection Research." ICML 2021 Workshop on Uncertainty and Robustness in Deep Learning.

<http://www.gatsby.ucl.ac.uk/~balaji/udl2021/accepted-papers/UDL2021-paper-033.pdf>

# BETH Dataset (cont.)

## • System log files

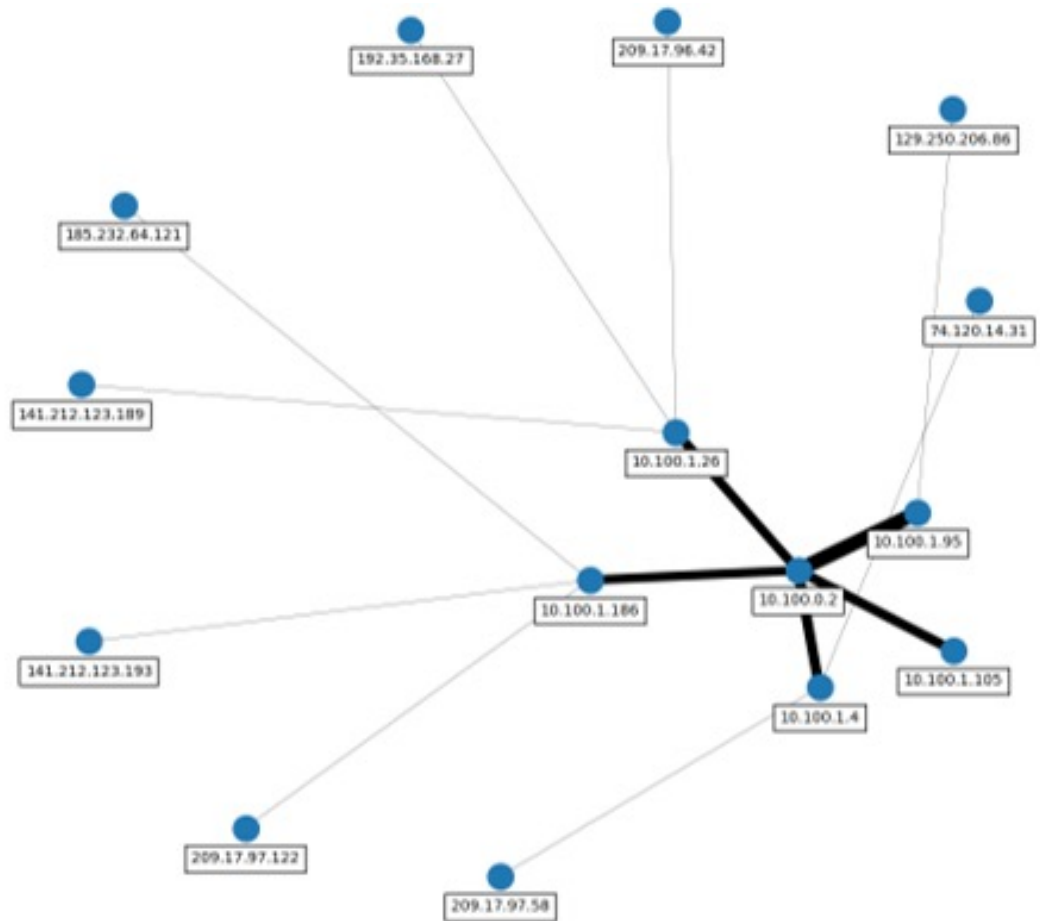
timestamp	processId	threadId	parentProcessId	userId	mountNamespace	processName	hostName	eventId	eventName	sockAddress	argsNum	returnValue	args	sus	evil
129.050634	382	382	1	101	4026532232	systemd-resolve	ip-10-100-1-217	41	socket	[1401591956]	3	15	{{'name': 'dd	0	0
129.051238	379	379	1	100	4026532231	systemd-network	ip-10-100-1-217	41	socket	[1398532280]	3	15	{{'name': 'dd	0	0
129.051434	1	1	0	0	4026531840	systemd	ip-10-100-1-217	1005	security_file_open	[1403628671]	4	0	{{'name': 'pa	0	0
129.051481	1	1	0	0	4026531840	systemd	ip-10-100-1-217	257	openat	[]	4	17	{{'name': 'di	0	0
129.051522	1	1	0	0	4026531840	systemd	ip-10-100-1-217	5	fstat	[1403628671]	2	0	{{'name': 'fd	0	0
129.051635	1	1	0	0	4026531840	systemd	ip-10-100-1-217	3	close	[1403628672]	1	0	{{'name': 'fd	0	0
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...

## • DNS query log files

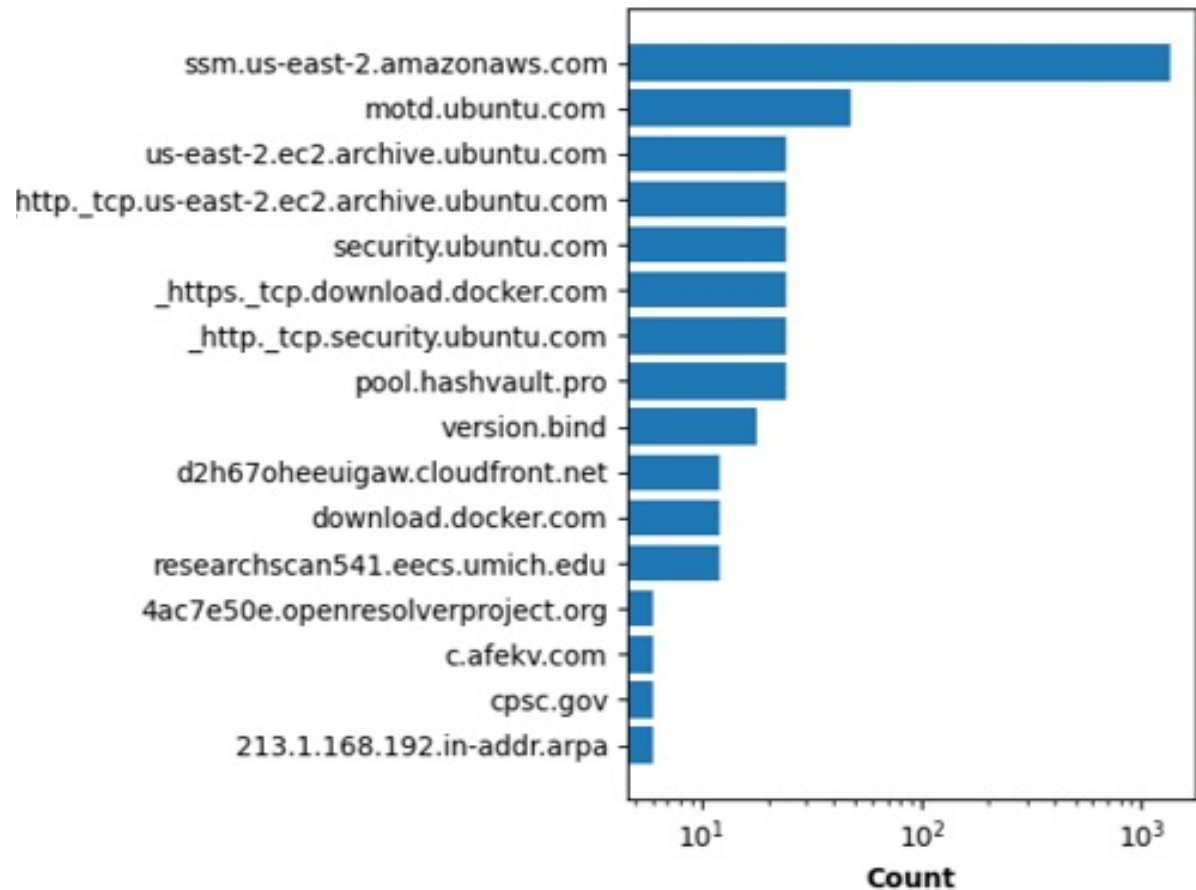
Timestamp	SourceIP	DestinationIP	DnsQuery	DnsAnswer	DnsAnswerTTL	DnsQueryNames	DnsQueryClass	DnsQueryType	NumberOfAnswers	DnsResponseCode	DnsOpCode	SensorId	sus	evil
2021-05-16T17:13:14Z	10.100.1.95	10.100.0.2	ssm.us-east-2.amazonaws.com			ssm.us-east-2.am	['IN']	['A']	0	0	0	ip-10-100-1-95	0	0
2021-05-16T17:13:14Z	10.100.0.2	10.100.1.95	ssm.us-east-2.am[["52.95.19.240"]	['17']		ssm.us-east-2.am	['IN']	['A']	1	0	0	ip-10-100-1-95	0	0
2021-05-16T17:13:14Z	10.100.1.95	10.100.0.2	ssm.us-east-2.amazonaws.com			ssm.us-east-2.am	['IN']	['AAAA']	0	0	0	ip-10-100-1-95	0	0
2021-05-16T17:13:14Z	10.100.0.2	10.100.1.95	ssm.us-east-2.amazonaws.com			ssm.us-east-2.am	['IN']	['AAAA']	0	0	0	ip-10-100-1-95	0	0
2021-05-16T17:13:16Z	10.100.1.186	10.100.0.2	ssm.us-east-2.amazonaws.com			ssm.us-east-2.am	['IN']	['A']	0	0	0	ip-10-100-1-186	0	0
2021-05-16T17:13:16Z	10.100.0.2	10.100.1.186	ssm.us-east-2.am[["52.95.21.209"]	['41']		ssm.us-east-2.am	['IN']	['A']	1	0	0	ip-10-100-1-186	0	0
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...



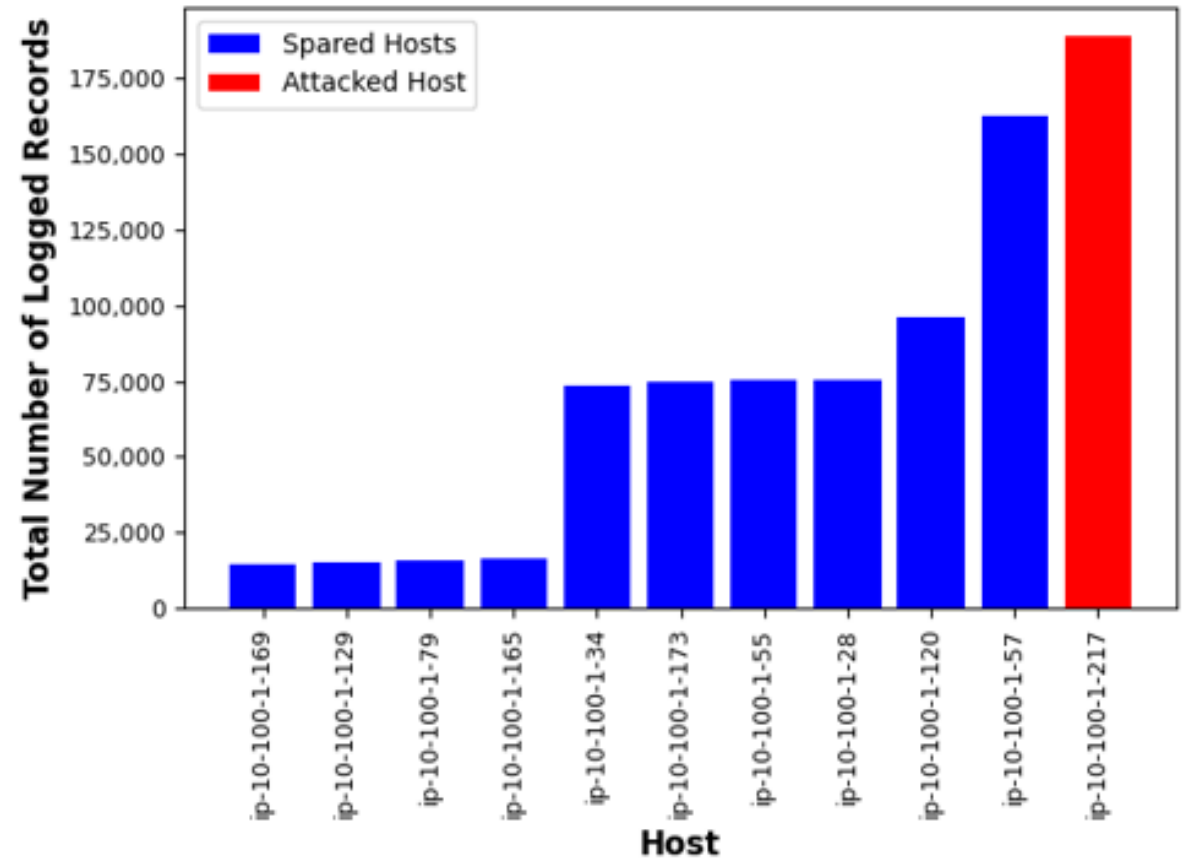
# DNS Query Traffic between IP Addresses



# DNS Query Volume by Domain

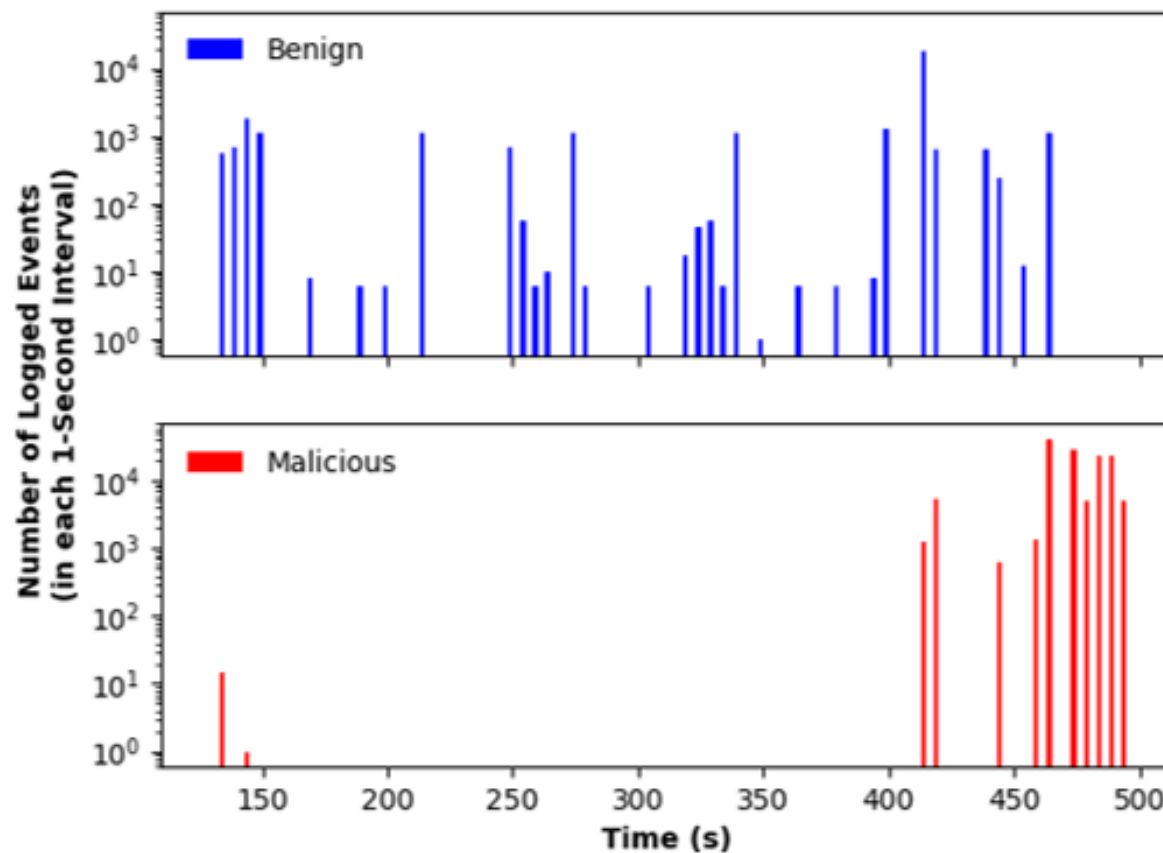


# Logged Events by Host

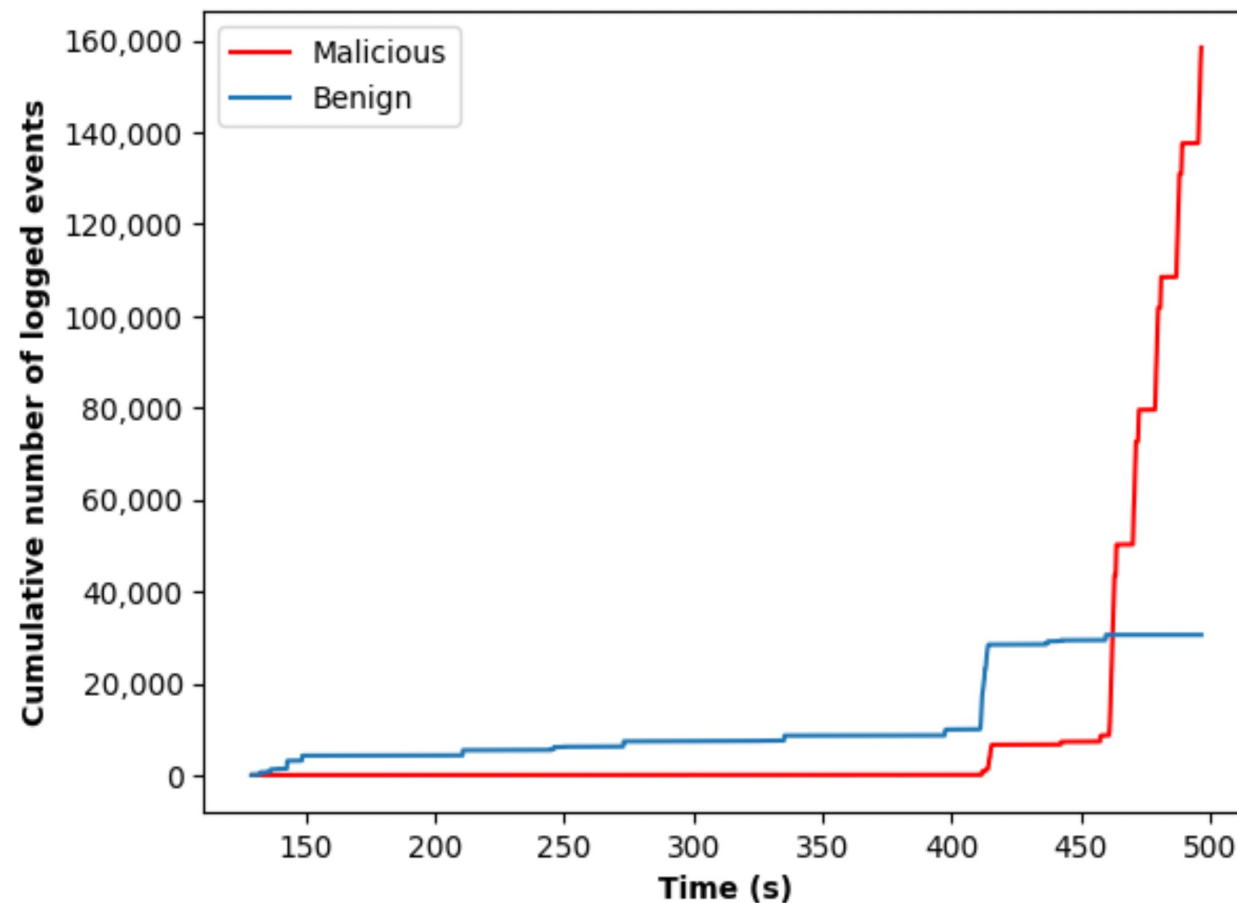




# Logged Events in Time on the Attacked Host



# Logged Events in Time on the Attacked Host



# Lab Environment Setup



# Environment Options

- **Option 1: Google Colab (recommended)**
  - Write and execute code in a browser
  - No installation required
  - Requires gmail account
  - <https://colab.research.google.com>
- **Option 2: Student-preferred python environment**

# Tour of Google Colab Functionality

- Add and execute code cells
- Add formatted text cells
- Import data

## Download Lab Files from GitHub

<https://github.com/CDS-Team/InfoSecWorld24>



# Hands-on Exercise

## Install and Import Libraries

```
!pip install umap-learn  
import umap
```

## Read .csv into Pandas Dataframe

```
df = pandas.read_csv('data.csv')  
df.dtypes
```



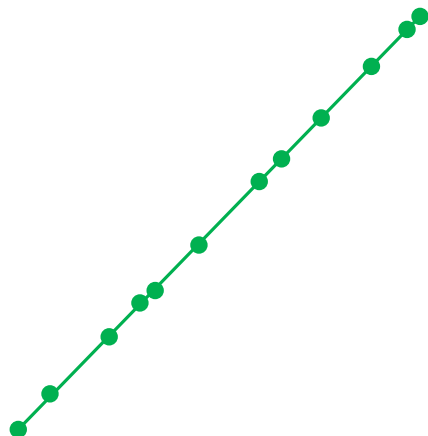
## View First 5 Records

```
df.head()
```

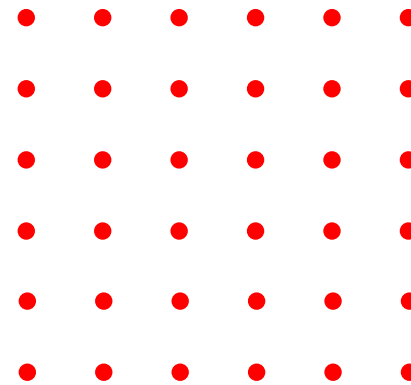
## Histograms of the Raw Data

```
df.hist()  
plt.tight_layout()
```

## Cyber-Specific Challenges (cont.)



Underlying Function



Underlying Function?

***Cyber data often limit us to basic statistics and challenge the modeling of relationships***

## Histograms of the Engineered Features

```
df_eng, X, y = preprocess(df)
df_eng.hist()
plt.tight_layout()
```



## Correlations Plot

```
correlations = df_eng.corr()  
seaborn.heatmap(correlations, cmap='vlag')  
plt.show()
```

# UMAP Dimensionality Reduction

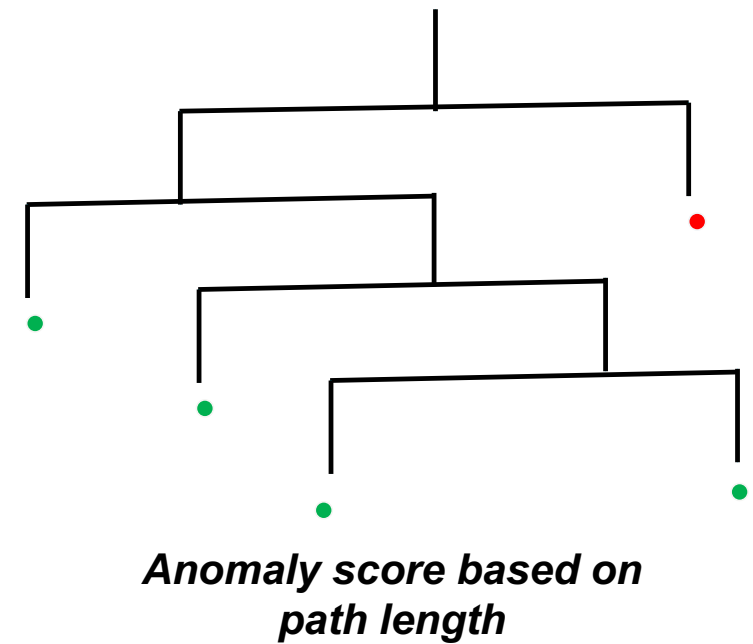
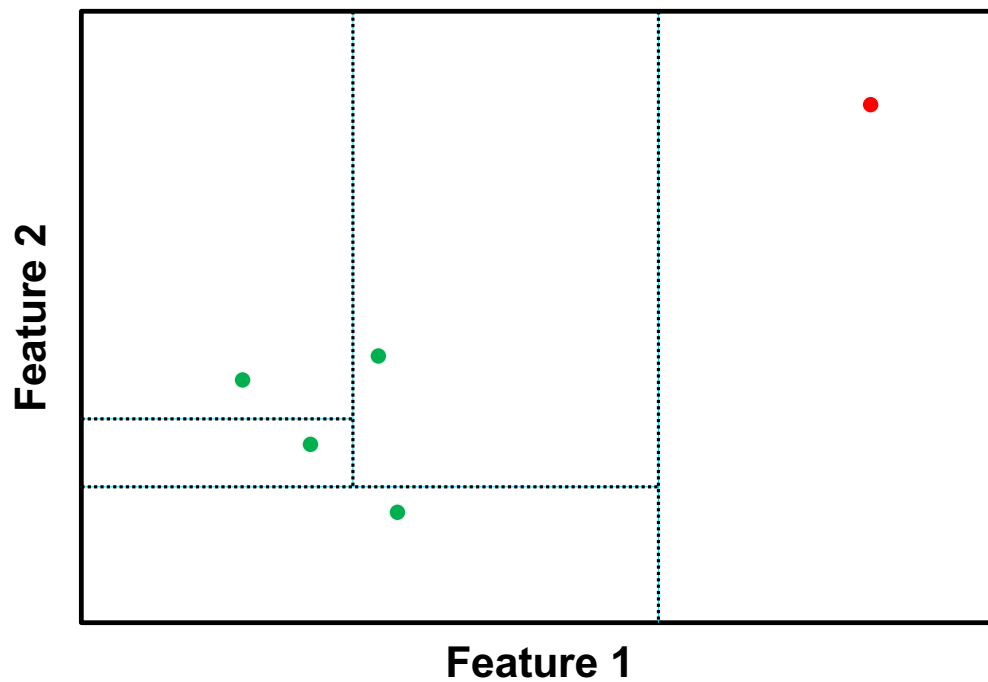
```
manifold = umap.UMAP().fit(X)  
X_reduced = manifold.transform(X)
```

# Outlier Detection in Cybersecurity

- Consider a busy SOC analyst
- Recently installed NDR/EDR anomaly detection
- System promises to identify novel zero-day threats through the power of ML...

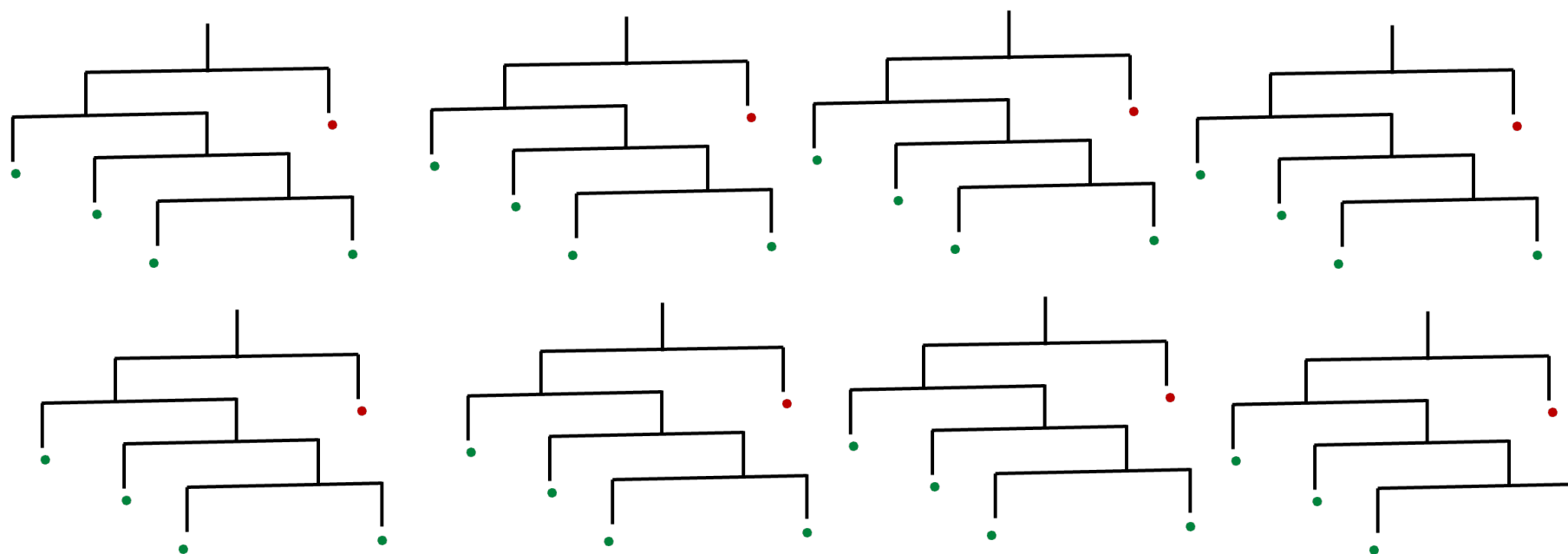


# Isolation Forest





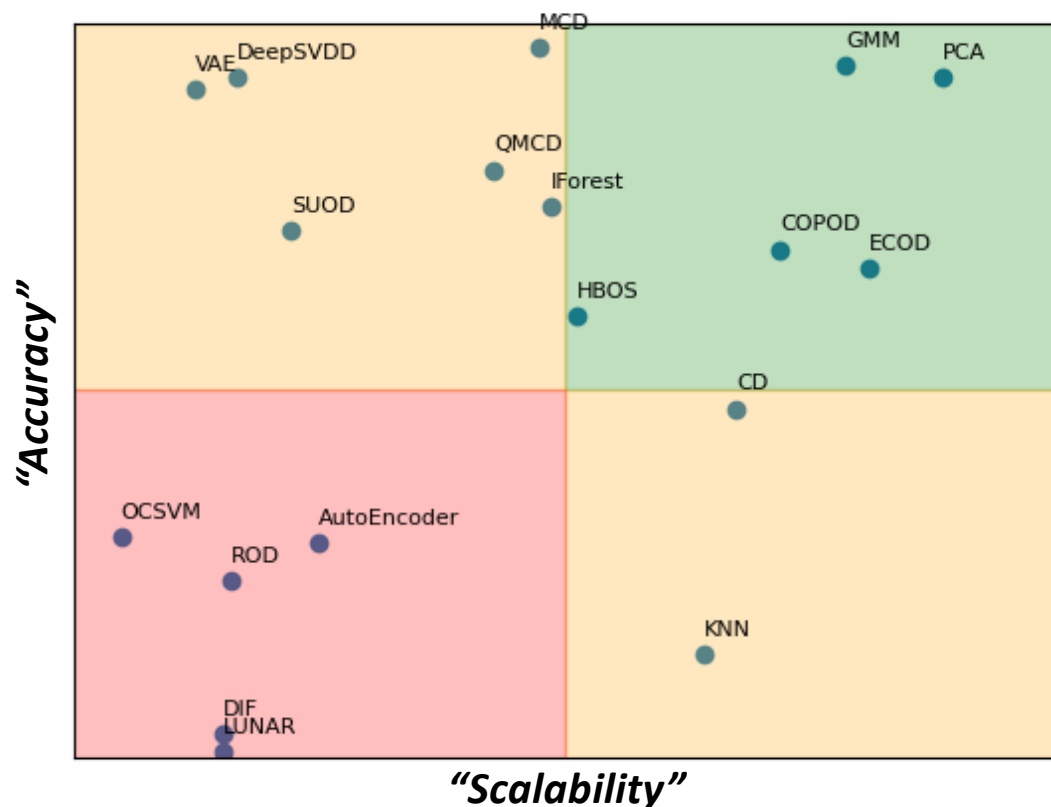
## Isolation Forest (cont.)



# Fit Anomaly Detection Model

```
model = sklearn.ensemble.IsolationForest().fit(X)
```

# Algorithm Performance



- Accuracy-scalability tradeoff
- We tested 18 algorithms on a cybersecurity dataset
- However, these results do not generalize
- **Algorithm performance is application-specific**

# THANK YOU!