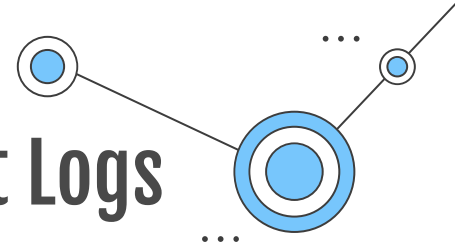


Malicious Behavior Detection Using Audit Logs



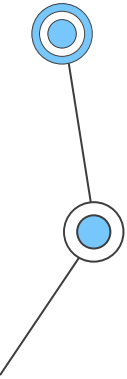
Beta Release

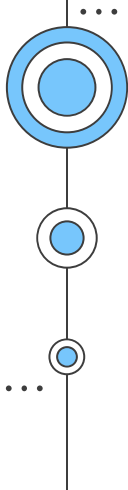
Team ID: 19

Project ID: CS_Egcert3_2023

Presented by :

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Project Mentor
Dr. Ahmed Hamdy



Project Sponsor:
EGCERT company



Ottawa support:
Dr. Miguel Garzon

Safaa Sarhan - 300389876

Problem definition



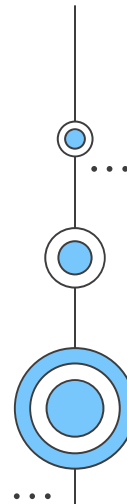
Increasing Cybersecurity Threats



Insufficient Detection Mechanisms



Challenges in Manual Analysis of logs



Users and Benefits

01

detect anomalies in
user activity.

02

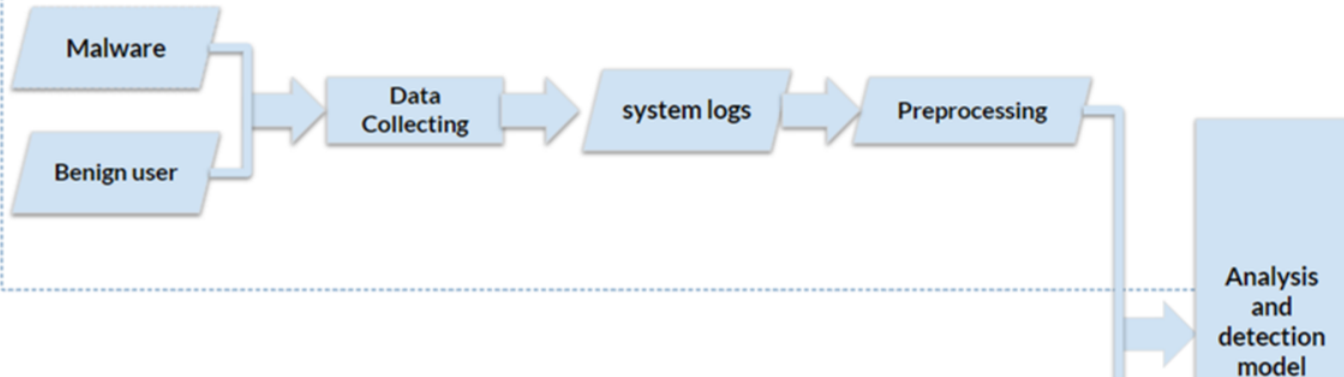
Improved threat
detection and
incident response

03

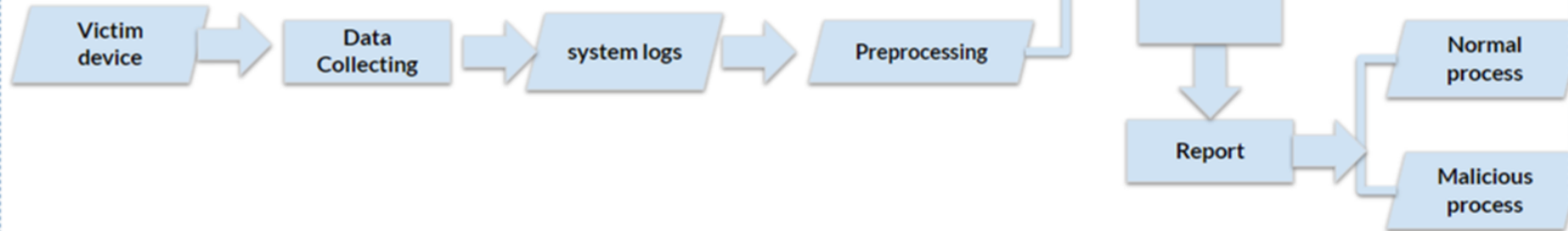
Reduced risk of
successful attacks

The proposed log analysis and detection method

Training



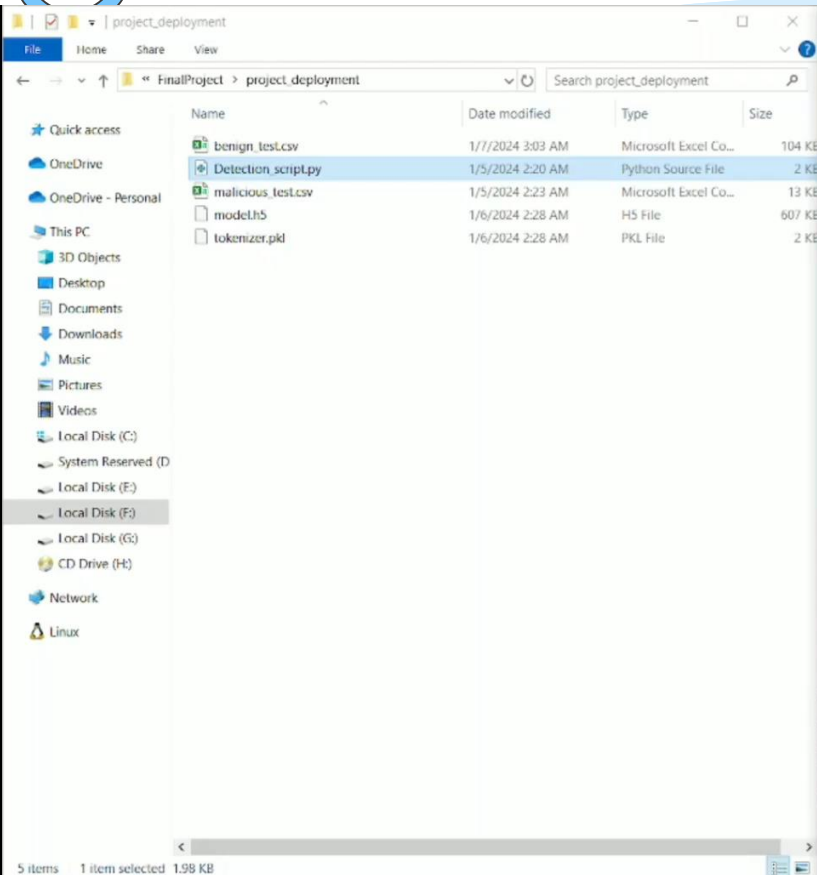
Analysis and detection





Our Demo

Safaa Sarhan - 300389876



01

Data collection

Why sysmon?

enable collection of detailed information : various system events, processes, network connections..etc.



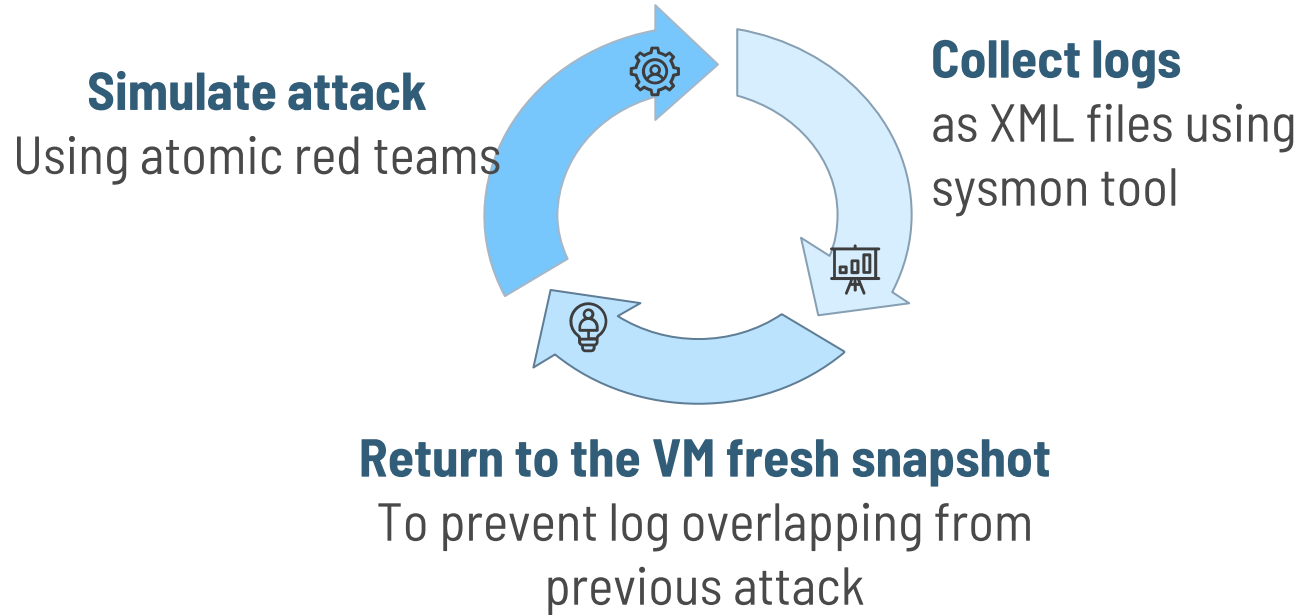
benign data : collected from our daily used devices



malicious data : Simulated almost 300 attacks from MITER ATT&CK matrix on a VM



Malicious logs



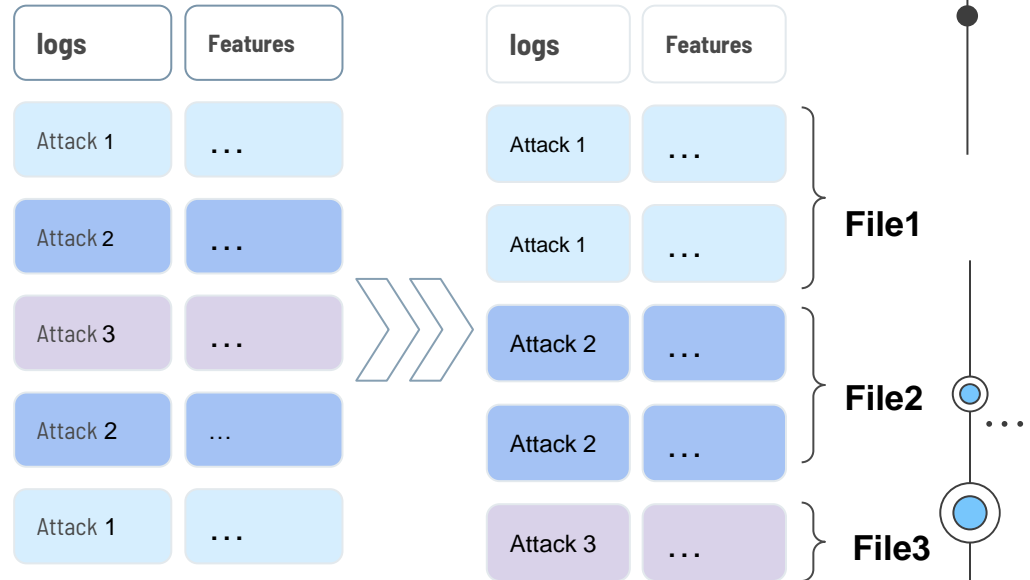
Why this cycle and not simulate all attacks in one VM?

Why..?

analyse pattern behaviour

Some attacks generate logs
after a while

it can interface with the logs
of the next attack

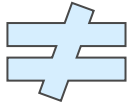


02

Data preprocessing

How to process the data for anomaly detection?

specific sequence
of benign logs

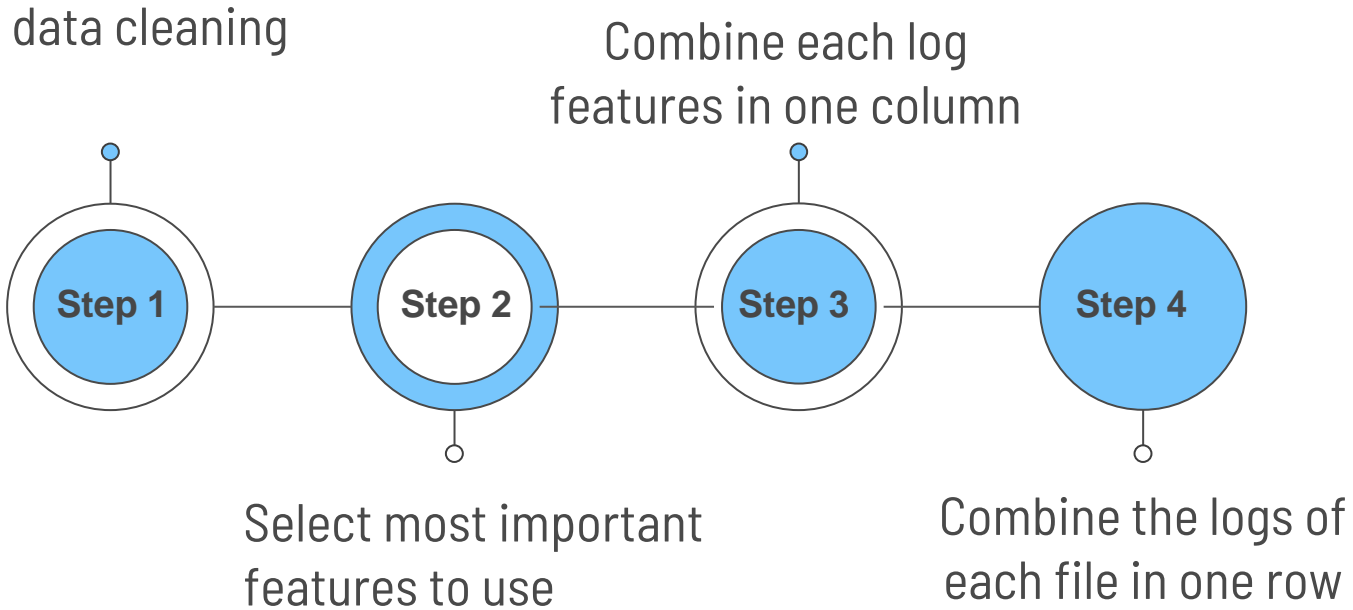


1 normal log

the analysis will be done by
collection not one log
“Pattern analysis”

logs	Features	
Benign 1	...	} Malicious
Benign 2	...	
Benign 3	...	
Benign 4	...	

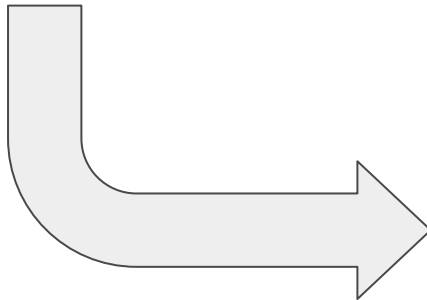
Our Process



Selected features

```
df.columns
```

```
Index(['Channel', 'CommandLine', 'Company', 'Computer', 'CreationUtcTime',  
      'CurrentDirectory', 'Description', 'DestinationHostname',  
      'DestinationIp', 'DestinationIsIpv6', 'DestinationPort',  
      'DestinationPortName', 'Details', 'EventID', 'EventRecordID',  
      'EventType', 'FileVersion', 'Hashes', 'Image', 'Initiated',  
      'IntegrityLevel', 'Keywords', 'Level', 'LogonGuid', 'LogonId',  
      'Message', 'NewThreadId', 'Opcode', 'OriginalFileName',  
      'ParentCommandLine', 'ParentImage', 'ParentProcessGuid',  
      'ParentProcessId', 'ParentUser', 'ProcessGuid', 'ProcessID',  
      'ProcessId', 'Product', 'Protocol', 'QueryName', 'QueryResults',  
      'QueryStatus', 'RuleName', 'SourceHostname', 'SourceImage', 'SourceIp',  
      'SourceIsIpv6', 'SourcePort', 'SourcePortName', 'SourceProcessGuid',  
      'SourceProcessId', 'SourceUser', 'StartAddress', 'StartFunction',  
      'StartModule', 'SystemTime', 'TargetFilename', 'TargetImage',  
      'TargetObject', 'TargetProcessGuid', 'TargetProcessId', 'TargetUser',  
      'Task', 'TerminalSessionId', 'ThreadId', 'User', 'UserID', 'UtcTime',  
      'Version', 'raw', 'Label'],  
      dtype='object')
```



```
selected_features = [
```

```
"CommandLine", "Company", "CurrentDirectory", "DestinationIsIpv6", "DestinationPort",  
"DestinationPortName", "Details", "EventID", "EventType", "FileVersion", "Initiated",  
"IntegrityLevel", "Level", "Message", "ParentCommandLine", "ParentUser", "Product",  
"Protocol", "QueryName", "QueryResults", "QueryStatus", "RuleName", "SourceImage",  
"SourceIsIpv6", "SourcePort", "SourcePortName", "StartFunction", "StartModule",  
"TargetFilename", "TargetImage", "TargetUser", "Task"
```

```
]
```

03

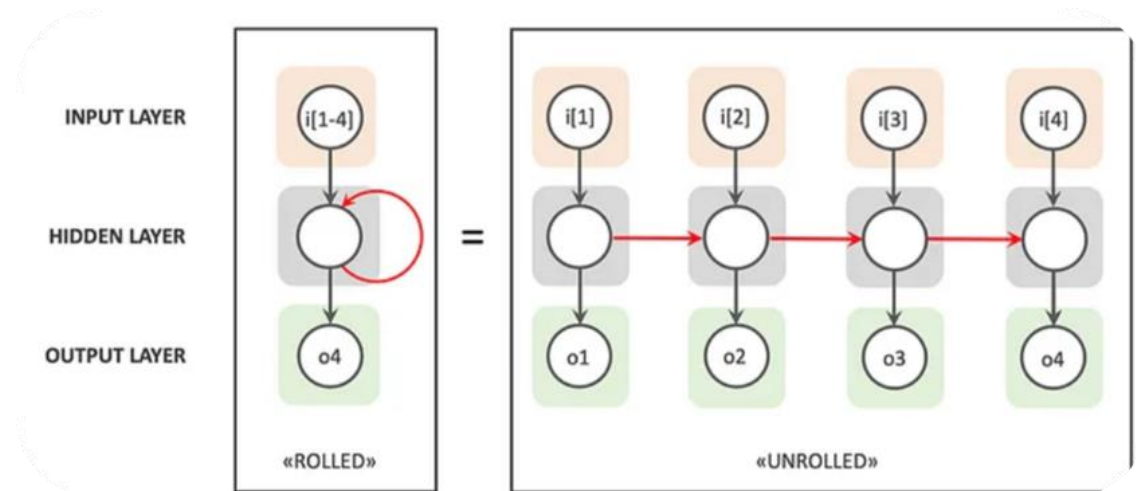
Analysis and detection model

RNN Model

Recurrent Neural Networks model (RNN)

designed to process sequential data

takes some specific input and returns output (Many to one)



...

libraries: Pandas, NumPy and TensorFlow.

Building the RNN Model:

Contains of :

1. embedding layer
 2. bidirectional LSTM layer
 3. 2 dense layers with a 'sigmoid' activation function
- ...

```
# Build an LSTM model
model = Sequential()
model.add(Embedding(input_dim=len(tokenizer.word_index) + 1, output_dim=50, input_length=maxlen))
model.add(Bidirectional(LSTM(units=50, activation='relu')))
model.add(Dense(units=32, activation='sigmoid'))
model.add(Dense(units=1, activation='sigmoid'))

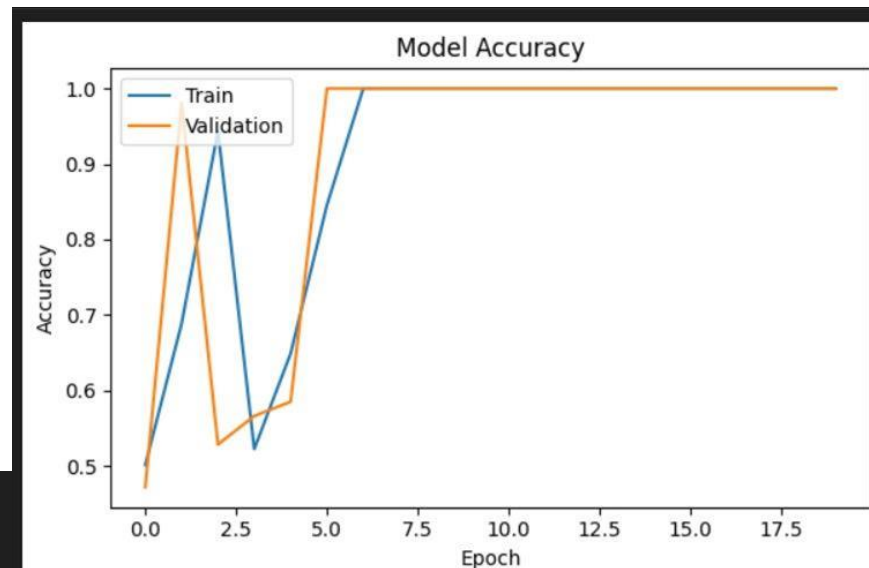
# Compile the model
optimizer = Adam(learning_rate=0.001)
model.compile(optimizer=optimizer, loss='binary_crossentropy', metrics=['accuracy'])
```




Model accuracy

In each iteration:

validation and accuracy
are monitored to check
for overfitting and
model's performance.



[0.98921895]

[0.98921895]

...

[0.98921895]]

4/4 [=====] - 0s 32ms/step - loss: 0.1318 - accuracy: 0.9667

Test Loss: 0.1318

Test Accuracy: 96.67%

04

Deployment

Deployment Plan

01

Extract the XML log

The end-user able to use windows sysmon logs , or a one query log at a time.



02

Anomaly Detection

The end-user can query for anomaly in any of those two files.



Since **XML** logs requires a lot of effort to be processed, in our deployment we configure a one united script to deals with more than XML format .



Final shape of data

[illegible]

Deployment Plan

1

We saved the model in a pickle file

```
# Save the trained model
model.save("F:/FinalProject/new-model/model.h5")

# Save the tokenizer
with open("F:/FinalProject/new-model/tokenizer.pkl", 'wb') as f:
    pd.to_pickle(tokenizer, f)
```

2

Feeding the model with the final shape of data to be predicted .

```
# Make predictions using the loaded model
prediction = model.predict(padded_sequence)
```

Deployment Plan

3

test a new log file -> "malicious" or "benign"

```
Enter the path to the file for detection: F:\FinalProject\new-model\malicious_test.csv
1/1 [=====] - 1s 578ms/step
Result Malicious
PS F:\FinalProject\new-model>
```

```
Enter the path to the file for detection: F:\FinalProject\new-model\benign_test.csv
1/1 [=====] - 1s 593ms/step
Result Benign
```

4

CMD interface enable end-users / security Analysts to test logs from their environments with minimal installation effort , and from only one script.



Future Work



1

Generate reports and dashboards to provide insights into the overall security posture.

2

Improve the model with more threats types





Thanks!

Do you have any questions?

