

# BVRIT Hyderabad College Of Engineering For Women



## KEYLOGGER DETECTION

### Team-10

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April 1, 2023

# Problem Statement

## Detection of Keylogger using Machine Learning

- perform data exploration, preprocessing and visualization
- implement classification model using sklearn library
- evaluate the model using appropriate performance metrics
- develop the Keylogger Prediction system

# Python Packages and Libraries used

- Pandas
- Matplotlib
- seaborn
- sklearn
- Numpy
- tkinter

# Algorithms

- Decision Tree
- K Nearest Neighbor
- Random Forest

# Decision Tree

Decision tree is a fundamental component of the random forest algorithm. Decision tree works by recursively splitting the data into subsets based on the most significant feature, which results in a tree-like structure. Each internal node of the tree represents a test on a feature, and each branch represents the outcome of the test. The leaves of the tree represent the class labels. In the context of keylogger detection, decision tree can be used to classify different types of attacks based on their features, such as the behavior of the user and the system.

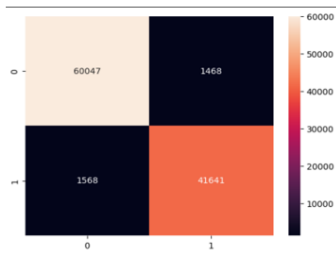
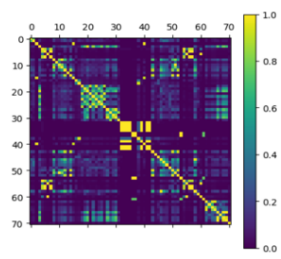
# K Nearest Neighbor

KNN (K-Nearest Neighbors) is a machine learning algorithm that can be used for keylogger detection. KNN works by finding the k-nearest neighbors to a given data point and classifying it based on the majority class of those neighbors. By analyzing the behavior of the keylogger, KNN can identify different keystrokes and patterns generated by the keylogger. KNN may not be the most suitable algorithm for keylogger detection in some cases.

# Random Forest

In keylogger detection, the random forest algorithm works by combining the output of multiple decision trees to classify different types of attacks and improve the accuracy of the classifier. Each individual tree in the random forest spits out a class prediction, and the class with the most votes becomes the model's prediction. By using this algorithm, keylogger detection systems can accurately classify different types of attacks and improve their overall accuracy.

# Confusion Matrix and Correlation Matrix





# Comparison Table

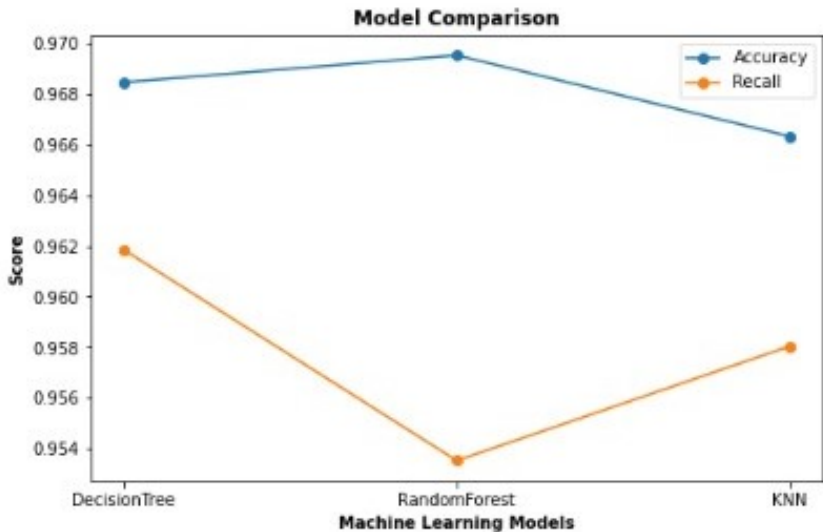
## Comparison of different models based on accuracy and recall

```
✓ 0s # Define the models and their accuracy and recall scores
models = ['DecisionTree', 'RandomForest', 'KNN']
accuracy_scores = [dtcaccuracy, rfccaccuracy, accuracy]
recall_scores = [dtcrecall, rfcrecall, recall]
```

```
✓ [128] comp_tb = pd.DataFrame({'Model': models, 'Accuracy': accuracy_scores, 'Recall': recall_scores})
print(comp_tb)
```

	Model	Accuracy	Recall
0	DecisionTree	0.968450	0.961841
1	RandomForest	0.969512	0.953532
2	KNN	0.966319	0.958042

# Comparison Graph



# Output using Gradio

enter 15 inputs

140931.0 208.0 37.111111 0.0 3.179603e+06 12140931.0 4.200924e+06  
1697.0 46705.624402 1.0 0.0 0.0 0.0 0.0 0.0

Clear

Submit

output

keylogger is not detected

Flag

# Output using GUI

Flow Duration:	<input type="text" value="1"/>
Fwd Packet Length Max:	<input type="text" value="0"/>
Fwd Packet Length Mean:	<input type="text" value="0"/>
Bwd Packet Length Min:	<input type="text" value="1"/>
Flow IAT Std:	<input type="text" value="0"/>
Fwd IAT Total:	<input type="text" value="2"/>
Fwd IAT Std:	<input type="text" value="5"/>
Bwd IAT Total:	<input type="text" value="8"/>
Bwd IAT Std:	<input type="text" value="1"/>
PSH Flag Count:	<input type="text" value="0"/>
ACK Flag Count:	<input type="text" value="1"/>
URG Flag Count:	<input type="text" value="0"/>
Down/Up Ratio:	<input type="text" value="0"/>
Active Mean:	<input type="text" value="0"/>
Idle Mean:	<input type="text" value="0"/>

Predict

keylogger is detected



# Execute code

- COLAB LINK FOR THE CODE

**THANK YOU!**