KeyStroke Biometrics with RNN(LSTM)

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*Abstract – Cybercrimes have been on the rise in the recent decade people can easily hack into people’s account after gaining access to the user’s password. Black Hat hackers are using various methods to retrieve the password of an account. More sophisticated authentication should therefore be used, one authentication method that authentications based on behavioral patterns instead of a string of characters that everyone can mimic. The objective of the paper is to leverage on LSTM to authenticate a user based on user’s keystroke behavior.*

Keywords—Deep learning, Artificial Intelligence, Behavioral biometrics, Long Short Term Memory (LSTM) (key words)

# Introduction

cybercrime — which is predicted to inflict damages totalling $6 trillion USD globally in 2021 — would be the world’s third-largest economy after the U.S. and China.by Calif, S [1] Trillions of dollars are being lost to cyber-crimes yearly which strongly suggest the need for a robust security system. In recent years, behavioural authentication caught the eyes of security experts as they are harder for hackers to bypass. In this article, we will discuss the basics of keystroke dynamics and how it can be used to authenticate users.

Keystroke biometric is a technique to verify an individual's identity by analysing his or her typing patterns. This technology has several advantages over the current security features that are implemented such as password access, One Time Passcode (OTP). Firstly, Keystroke biometrics are easy to implement for the user since they do not even need to remember any extra information and interact with the keyboard as they normally would. Secondly, individual users can be constantly authenticated instead of a one-time password which the hackers can retrieve once and access the system for a certain period.

A keystroke biometric dataset consists of a set of input sequences (keystrokes) that are associated with a specific individual and a corresponding set of output sequences (texts). This data can be collected either manually or through an automated system by either capturing the user's actions or installing a keyboard capture driver on the device. Once the dataset is collected, it can be processed using machine learning algorithms to identify suitable sequences for use in keystroke authentication.

# Related Works

The idea of keystroke biometrics have been around for decades and several engineers like Kevin S. Killourhy and Roy A. Maxion [2] have collected and experimented various machine learning models on the data. Certain models or algorithms include Support Vector Machines (SVM) , KMeans Clustering and Dense Neural networks. However, LSTM was not tried to tackle this problem and since it is a time series problem which also requires memory , we will look at how LSTM can help with Keystroke biometrics.

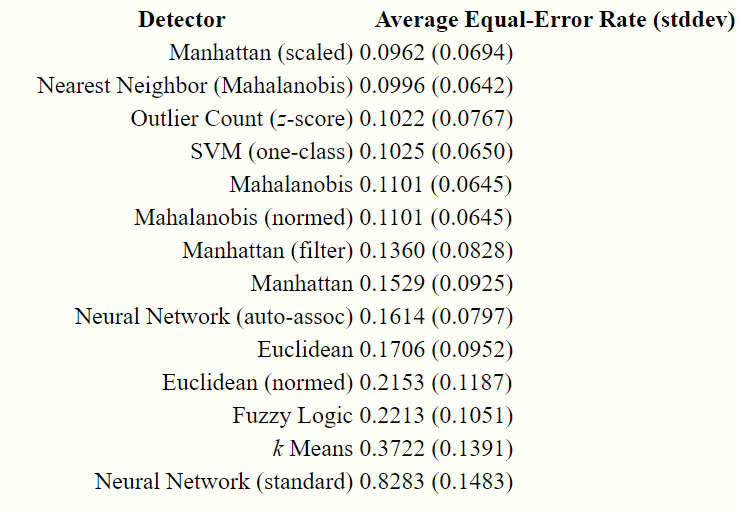


Fig. 1. List of Machine learning Equal Error Rate (EER)

As mentioned by C.,Eric and F.,Joshua EER also known as Crossover Error Rate(CER) is when the False Accept Rate(FAR) and the False Reject Rate (FRR) meet when the sensitivity is lowered .[3]

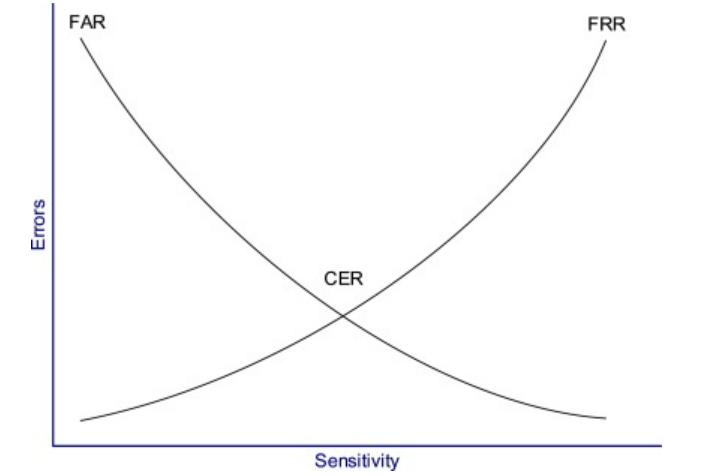


Fig. 2. Graph of CER where FAR meets FRR [3]

FAR is wrongly accepting a person and FRR is wrongly rejecting the legitimate user. This gave a deeper insight into how I can evaluate the model later.



# Methodology

In this paper, the following Python libraries will be used:

**Pandas** - Built on top of NumPy, it is a useful data science tool that can perform data wrangling, display numerical summaries and other data cleaning tasks [3].

**Seaborn** - A visualization library that is built on Matplotlib. It can generate many types of statistical graphs and works well with Pandas data structures [4].

**sklearn** - An extremely useful Python library for Machine Learning. This library contains various Machine Learning algorithms and statistical modeling tools [5].

**TensorFlow** – A Deep learning framework which is often seen as the most popular deep learning framework, which is very powerful & easy-to-use and has excellent community support.[6]

## Data

Kevin S. Killourhy and Roy A. Maxion has kindly shared the data they have collected in a controlled environment back in 2007 where 51 typist are presented with a password to type and if there were errors, the software will prompt the typist to retype until 50 correctly typed password in each session. There were a total of 8 sessions which is 400 data points per typist.

## Exploratory Data Analysis

Exploring the data is important to draw useful insights that can help improve the model or better feature engineer it. I first used pandas to import the csv file that was downloaded from the paper published by Kevin S. Killourhy and Roy A. Maxion

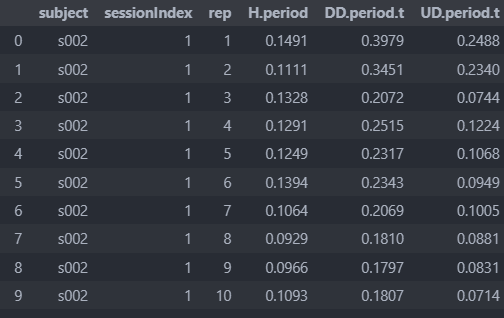


Fig. 3 Head of Pandas Dataframe of Keystroke Dynamics Dataset

There were missing subjects as not all the subjects who registered for the experiment but did turn up on that day. Subject column is in varchar initially. Each typist typed the password 400 times taking a break between 8 sessions.

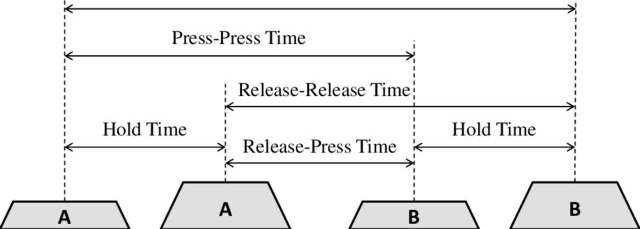


Fig. 4 Representation of Keystroke features/columns [6]

H.{key} represents the duration of which the {key} was held onto.

DD.{key1}.{key2} is the duration between pressing {key1} and then pressing {key2}

UD.{key1}.{key2} is the duration between releasing {key1} and then pressing {key2}.note that UD.{key1}{key2} can be negative if the typist do not release {key1} before hitting {key2}

There is also Release-Release or Press-Press time, but it probably will be that important to process that column since there is it can feature engineered easily by adding Release-Press time with Hold Time

### Data visualisation

Plot distribution

### Simple Clustering Analysis

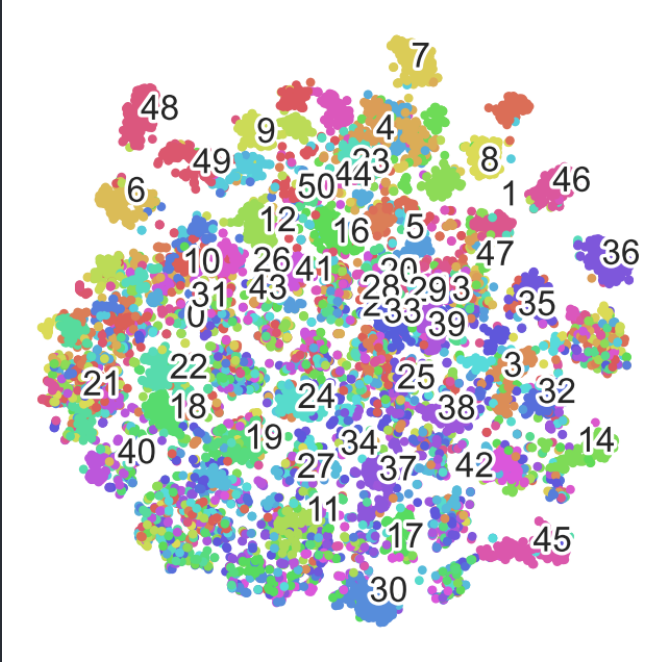


Fig. 5. TSNE clustering

Simple TSNE clustering was carried out on the raw data without much wrangling to see if its possible to segment and cluster all 51 typists. However, although there are some typists that are quite separated from the cluster, there is still quite a number who types pretty similarly to one another

## Feature Engineering / Preproccessing

### Data Wrangling

I created a new column called ‘id’ which is a label encoded column of ‘subject’ this can help the model better understand and for evaluation as well.

### Feature Scaling

Imported MinMaxScaler from sklearn.preprocessing

# Recurrent Neural Network(LSTM)

## Neural Network

Neural networks are a subset of machine learning, a combination of modern computer science and cognitive psychology. Data processing and analysis are based on the same processes that neurons in the brain process information. (J, Jeremy,2017) Your neurons require a certain amount of activation energy to fire and send information out of the neuron to axons, synapses, other neurons, etc. Data is fed into fully connected neural networks and matrix multiplication is carried out between the weights and the input. The weights are then improved through back propagation which reduces the validation loss of the function.

## Recurrent Neural Network

Recurrent neural work(RNN) is gaining more popularity due to its ability to retain sequential information. RNNs are called recurrent because they recurrently perform the same task for each element found in the sequence of the data with the output dependent on previous computations.

We can think of RNNs as model with memory (Denny ,2015)

## How LSTM works

LSTM stands for Long Short-Term Memory network used in the field of deep learning

# Results and discussions

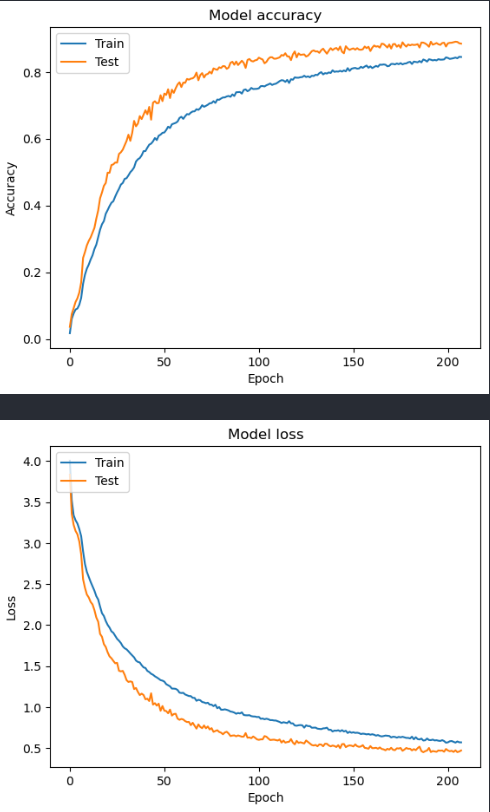


Fig. 6. Learning curve of LSTM model without scaling

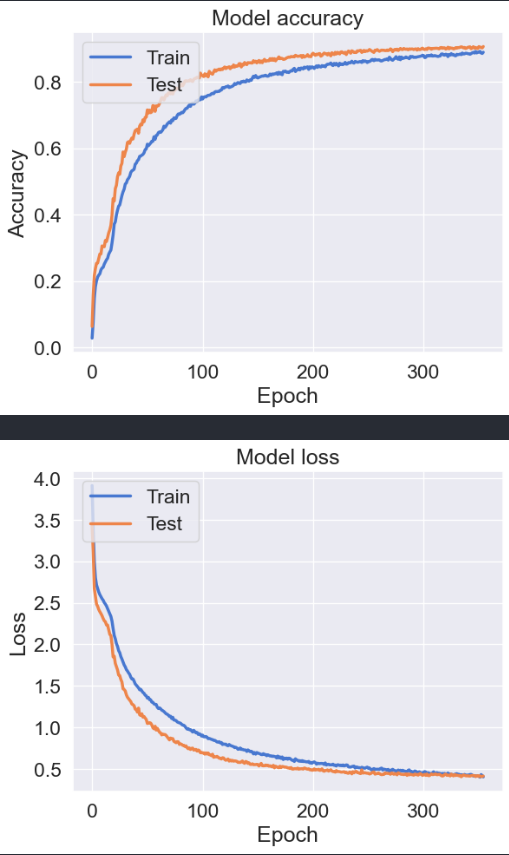


Fig. 7. 90.59% validation accuracy with scaling

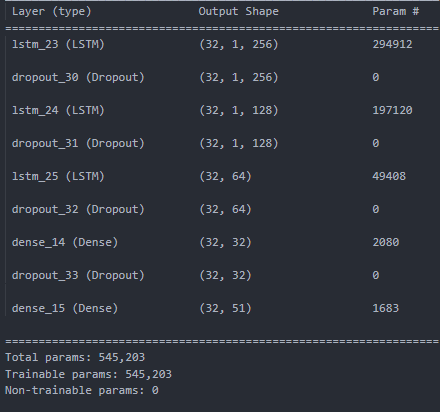


Fig. 8. model summary

Plot of accuracy

## Attackers

## Defenders

What is a good sensitivity to set? Like threshold

# Conclusion

The model was able to predict and identify the correct typist with a test accuracy of 90%.

##### Acknowledgment

I want to specially give credits to an awesome group of young machine learning engineers who inspired me to investigate new ways of authenticating and securing the digital space of our future.

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