gsel26/11/019-C1)

Name of the student A. Scimiuas. leg No 192211019 computer science engineering. Department or. v. Jaganathan. Cuide Name experimental Analysis of sum & RF Title for used Adentification using keystroke demarnics. 1. Introduction: - Approved paragraph - 1: (ADV. V. Inchwa AMA) \* lescalet about: Comparision of suppost vector machine and randomforest for user identification using keywoke dynamics. \* Importance! Evaluating classification performance accuracy and feature importance for keyetrobe based authentication. \* Applications: Cybersecurity, user authentication, frand prevention. paragraph - 2: \* No. of related actides in last 5 years: · Google Schold - 52 asticles. · ScienceDirect - 37 ostides. \* most cited-findings!

a killoushy & maxion, IEEE, 2009 - Sum effective for biometric authentication. [Hal-seience - 29]

- b. Teh et al. patteen recognition, 2013 Landomforestprovides better feature delection. [springer - 50]
- high-dimensional keystrokedata [chevier-42]
- Found RF outperforming sum in user authentication tacks. [springer-50]

paragraph-3!

- 4 Unanswered solution: No direct statistical compacision A SUM and RF on keystroke authebication.
- \* Dim: Compare classification accuracy and Statistical significance of SVM vs RF for Keystroke depromiss.

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# 2 Materials and Methods:

paragraph-1:

& Study setting: SSE, SIMATS

\* Ethical approval: 10/A (No human samples his ently mobiled).

\$ No. of groups: 2(gum, RF)

- \* Sample size | group: 10 each
- \* Zotal sample site: 20
- \* G-power: 80% (killoushy & makion, 1666, 2009)

#### paragraph-1:

- \* Dataset source: collected from come keystroke dynamics
- Feature catraction: Hold time, flight time, dwell time computed from keystroke events.

### pagagraph-3:

- \* preprocessing steps!
  - · outlier detetion and removal using 2-scole method
  - · Feature normalization vering trin-Mar Scaling
  - · Splitting data set into 80% training and 20% testing sets.

## paragraph 4:

#### & Testing Setup;

- · feature Extraction: python & Schit-leaen.
- · Model training: sum us Romdom Forest.
- · Evaluation metrici : Accuracy, precision, lecall / FI-Stole, Loc-AUC cueve.

#### paragraph-5:

· Data Collection method: Accuracy values recorded from multiple test runs and averaged for reliability. paragraph - 6:11

- \* Statistical softwale: spss
- \* Independent variables: modeltupe (sum vs RA)
- \* rependent-variables: classification accuracy.
- \* Analysis: Andependent sample 6-test, Descriptive statistics, Graphical representation (confusion matrix, Bachaet).

### 3. lesuits and Discussion!

### paragraph-11:

& summary of results: et has higher mean accuracy than sum.

#### paragraph-21

- \* secriptive table: Sum (88.14%) RF (93.38%).
- \* killouetry & Marion, 1666, 2009 similar results observed in keystroke authentication.
- \* Teh et al. pattern recognition, 2013- Found RF reducing
- John positives in biometrics.

  A conseneus:
  Random Forest performs significantly better.

## pocagraph-3:

- · Statistically significance (Independent-sample t-ten
  - p20.05, Indicating significant difference
  - · pp significantly outperforms sum in keystroke classification.

### paragraph-4:

- \* Graphical Representation of Results:
  - · confusion mertrex for both models to visualize misclassifications.
  - · Barchaet comparing accuracy score of sum of RF.
  - · ROC-AUC Cueve to show performance of both classifiers.

## paragraph-t:

#### \* himitations!

- · small dataset, which may limit generale zation.
- · landomforest might overfit, requiring turing.
- Keystroke timing voeiables due to user behaviour differences

# paragraph-6:

### \* fature scopes

- · Expand dataset to test voluntness.
- · Experiment with Leep learning models like CNN-LSTM.
- · Amprove real-time keystroke authentication veing ensemble models.

# 4. conducion:

Significantly outperforms SUM en user Adentification veing keystroke dynamics (pco.05).

#### Title1:

#### T-Test

#### **Group Statistics**

|          | GROUP | N  | Mean    | Std. Deviation | Std. Error Mean |
|----------|-------|----|---------|----------------|-----------------|
| ACCURACY | SVM   | 10 | 88.7430 | .65296         | .20648          |
|          | RF    | 10 | 93.3870 | .47237         | .14938          |

#### **Independent Samples Test**

Levene's Test for Equality of Variances

Hest for Equality of Means

|          |                             |       |      |         |        | Significance |             | Mean       | Std. Error | 95% Confidence Interval of the .<br>Difference |          |
|----------|-----------------------------|-------|------|---------|--------|--------------|-------------|------------|------------|--|----------|
|          |                             | F     | Sig  | t       | df     | One-Sided p  | Two-Sided p | Difference | Difference | Lower  | Upper    |
| ACCURACY | Equal variances assumed     | 1.342 | .262 | -18.222 | 18     | < 001        | <.001       | -4.54400   | .25485     | -5.17942                                       | -4.10858 |
|          | Equal variances not assumed |       |      | -18.222 | 16.395 | <.001        | <.001       | -4.64400   | .25485     | -5.18320                                       | -4.10480 |

