

SpooF-resistant Authentication Through Phone and Wearables



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The objective of this project is to develop a secure and easy-to-use mechanism of continuous user authentication for the current generation of smart phones

Continuous Authentication System based on Multi-Modal Wearable Devices:

Measure movement dynamics of different body parts (e.g., head, eye muscle, hand and wrist) and neuro-physiological characteristics (e.g., brain wave and eye gaze patterns)

SpooFing Attack Modeling and Analysis

Develop mechanisms to defend against active adversaries who may resort to robotic and sensor-spoofing attacks

System Usability Evaluations

Develop evaluation mechanisms focusing on not only authentication efficiency and accuracy but also user experience and user perceptions

Devices and Respective Features

- Movement Patterns using Inertial Sensors (on phone, watch and glasses)
- Brain activity Patterns from EEG Sensors (on BCI headset)
- Eye Movement Patterns (on glass)
- Touch Patterns (on phone)



Global view of our device interconnection setup

Approach

Template as, $T = \{TIP_1, TIP_2, \dots, TIP_n\}$, where P_1 to P_n are the base activities such as walking, standing, etc.

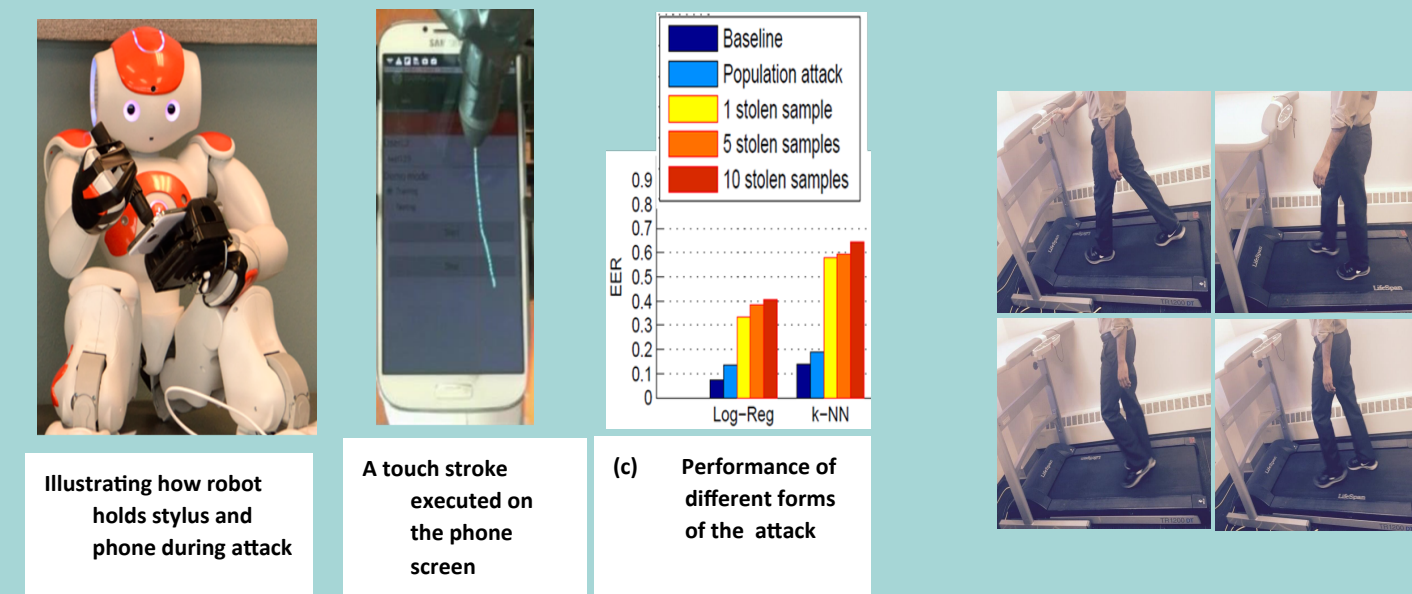
Assuming availability of data from smart watch, BCI devices, etc. update each of the sub-profiles in T to include information from the other sensors.

Fusing Devices and Sensors (including Preliminary Results) Classification Algorithms to be Studied

Fusion: Naïve Bayes; Weighted majority voting; Code alignments

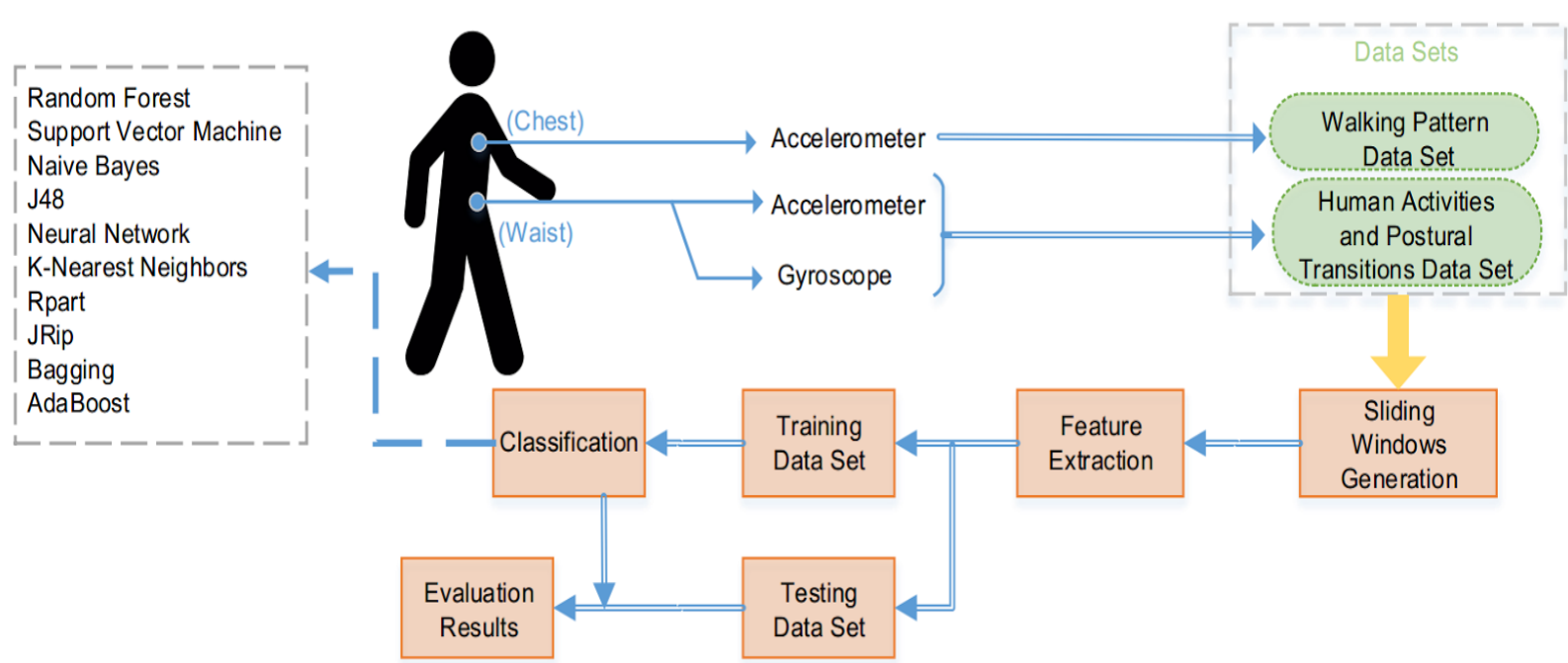
Classifiers: Naïve Bayes, 2) Tree Augmented Naïve Bayes, 3) Logistic Regression, 4) k-Nearest Neighbor, 5) Scaled Manhattan, 6) Support Vector Machines, 7) Scaled Mahalanobis, 8) Neural Network, and 9) Random Forests.

SpooF Attack Modeling and Analysis

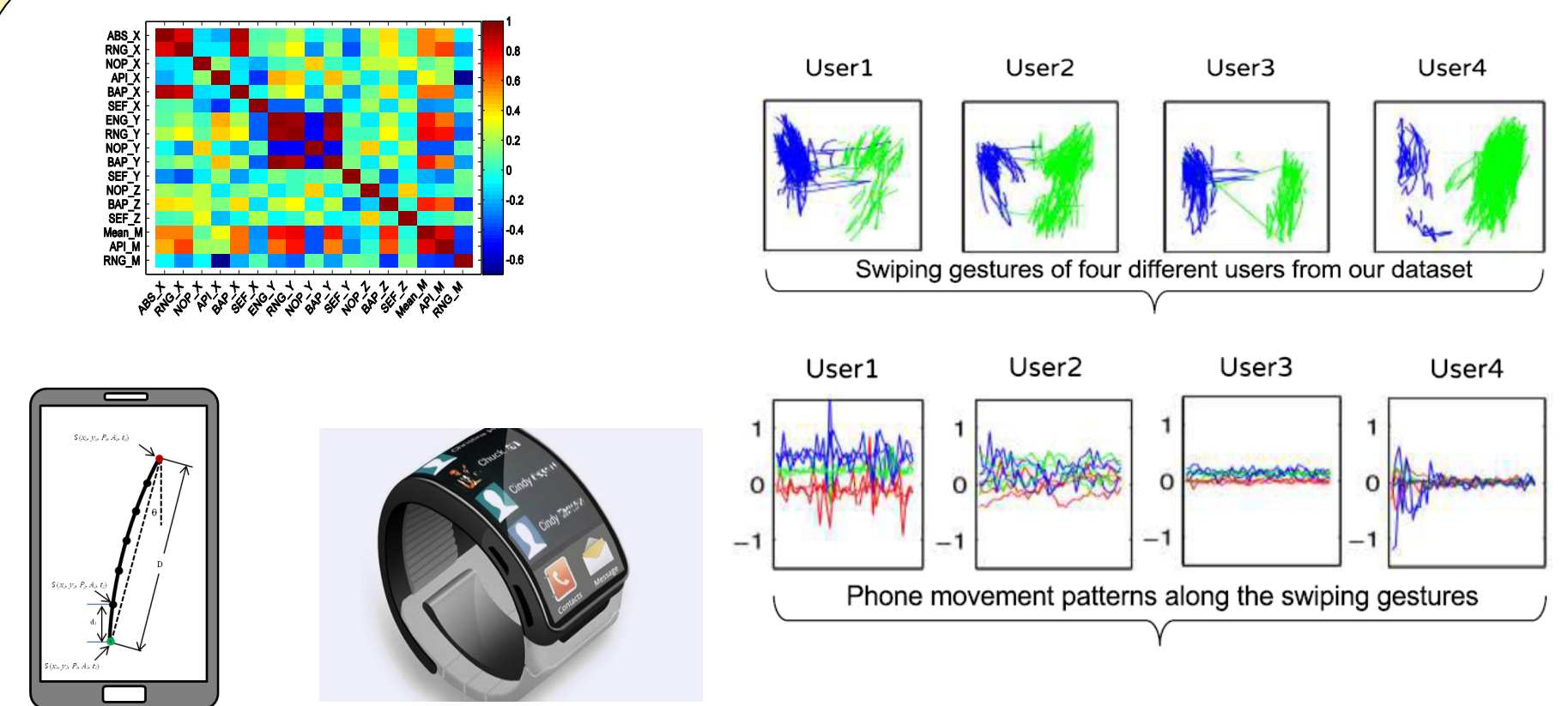


SpooFing the swipes

SpooFing the walk



Comparison of classifier performance through behavioral patterns obtained through phone



Fusion of different activity patterns

Publications

- R. Kumar, V. V. Phoha, and R. Raina, "Continuous Authentication of Smartphone Users by Fusing Typing, Swiping and Phone Movement Patterns," in *IEEE BTAS*, Buffalo, New York, 2016
- A. Serwadda, V. V. Phoha, Z. Wang, R. Kumar, and D. Shukla, "Toward Robotic Robbery on the Touch Screen," *ACM Transactions on Information System Security*, vol. 18, pp. 1-25, 2016
- C. Tang and V. V. Phoha, "An Empirical Evaluation of Activities and Classifiers for User Identification on Smartphones," in *IEEE BTAS*, Buffalo, New York, 2016.
- Otto Huhta, Prakash Shrestha, Swapnil Udar, Mika Juuti, Nitesh Saxena and N. Asokan, "Pitfalls in Designing Zero-Effort Deauthentication: Opportunistic Human Observation Attacks. In the Network and Distributed System Security Symposium (NDSS), February 2016
- Babins Shrestha, Manar Mohamed, Sandeep Tamrakar and Nitesh Saxena, Gametrics: Towards Attack-Resilient Behavioral Authentication with Simple Cognitive Games, In Annual Computer Security Applications Conference (ACSAC), December 2016
- Babins Shrestha, Manar Mohamed, Sandeep Tamrakar and Nitesh Saxena, Theft Resilient Mobile Payments: Transparently Authenticating NFC Users with Tapping Gesture Biometrics, In Annual Computer Security Applications Conference (ACSAC), December 2016

- **Activities:**
 - ❖ Dynamic
 - Walking
 - Walking upstairs
 - Walking downstairs
 - ❖ Static
 - Sitting
 - Standing
 - Lying
- **Transitions:**
 - ❖ Sit-to-stand
 - ❖ Stand-to-sit
 - ❖ Sit-to-lie
 - ❖ Lie-to-sit
 - ❖ Lie-to-stand
 - ❖ Stand-to-lie

Type	Activity	RF	SVM	NB	J48	NN	kNN	Rpart	JRip	Bag	AB	Average
Dynamic	Walk	97.6	89.7	92.4	89.7	97.8	100.0	78.5	80.5	87.7	97.8	91.2
	Upstairs	93.6	79.6	82.9	82.7	90.5	100.0	71.4	71.2	80.9	91.4	84.4
	Downstairs	94.6	71.4	78.0	80.2	86.3	100.0	68.2	67.2	79.2	95.6	82.1
Static	Sit	68.3	22.2	12.9	69.2	62.9	99.8	33.7	48.1	44.3	46.4	50.8
	Stand	72.2	37.5	25.6	74.1	76.8	100.0	52.6	60.7	57.9	59.8	61.7
	Lie	86.6	46.9	34.7	84.1	80.6	99.8	58.8	72.4	66.5	84.5	71.5
Aggregated	Aggregate	86.3	30.6	10.1	80.0	40.2	100.0	14.8	62.8	16.1	18.5	45.9
	Aggregate*	83.4	30.0	10.0	77.1	41.7	100.0	14.8	58.0	16.7	18.7	45.0
	Aggregate**	80.9	29.2	10.2	74.8	35.3	99.2	14.6	57.0	16.4	15.7	43.3

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