Hold & Sign: A Novel Behavioral Biometrics for Smartphone User Authentication

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Process of identifying the author of a given text.

• Handwritten signature is a specific instance.

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Barack Obama

- Process of identifying the author of a given text.
 - Handwritten signature is a specific instance.
- ► Wide social and legal acceptance in daily life.
 - Witness intentions (signing a contract)
 - indicate physical presence (signing in for work)

Barack Obama

In static systems, scanned image is processed and compared with the stored image to authenticate.

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- ► In dynamic/online systems, signature is acquired (by stylus or pen) and processed in real time and extracted global features are compared to authenticate.

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 iSignOn, SignEasy and eSignature (sutisoft¹) use signature image to authenticate.

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Must be able to run on the smartphones

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- Must be able to run on the smartphones
- ► Usability is very important for user acceptance
 - Must be quick
 - easy to use
 - "I Feel Like I'm Taking Selfies All Day!"
 - "It can be a bit embarrassing to hold your phone up in a weird position while in public just to unlock it!"²

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► Must be secure, i.e., should provide acceptable security.

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Bi-Modal System

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 - Sign touch (points pressed while signing)

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Bi-Modal System

- Sign touch (points pressed while signing)
- Hold phone's micro-movements while signing
- ► No additional hardware required.



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- ► Accelerometer 3 variants, i.e., RAW, LPF and HPF
 - accelerations in 3 dimensions
- ► Gravity
 - gravity forces in 3 dimensions
- Magnetometer
 - Strength and direction of magnetic field in 3 dimensions
- Touchscreen
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Validation

Data Collection

- 30 samples in each activity (total 90)
- 3 activities (sitting, standing, walking)



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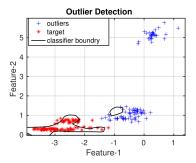
200

 30 participants (22 male), all of them Masters/PhD students, Google Nexus 5

One Class Classification (Anomaly Detection)

► Verifiers

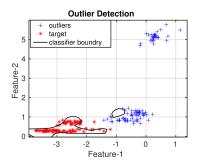
- We used a set of four conceptually different verifiers.
 - 3-NN (K-Nearest Neighbor)
 - 3-layer MLP (Multilayer Perceptron)
 - BayesNet
 - Random Forest



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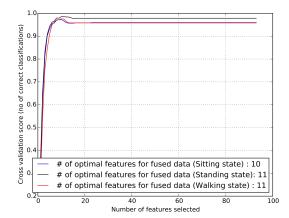
 After initial experiments we selected MLP since it performed consistently across situations(TAR=79%, FAR=0.1%)

Feature Selection

▶ 93 features (80 hold and 13 sign)

Feature Selection

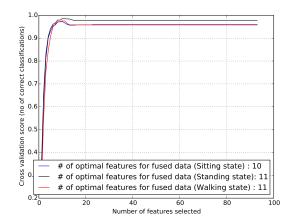
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- Recursive Feature Elimination (RFE) Method



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Feature Selection

- ▶ 93 features (80 hold and 13 sign)
- Recursive Feature Elimination (RFE) Method
 - Scikit-learn, 10-fold cross validation

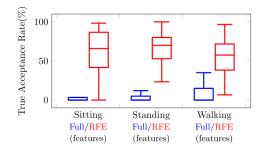


Results

► The results are averaged over 30 users

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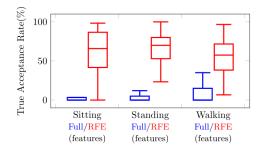


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Results

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 - Inter-Activity



Combined-Activities (RFE)

Classifier	TAR(%)	FRR (%)	FAR (%)	TRR(%)
MLP	94.8	5.2	3.1	96.9

Sample Acquisition Time

Method	Sample Acquisition Time(s)		
Hold & Sign	3.5		
PIN	3.7		
Voice	5.15		
Face	5.55		
Password	7.46		
Face + Voice	7.63		
Gesture	8.10		
Gesture + Voice	9.91		

S. Trewin, et. al, Biometric authentication on a mobile device: a study of user effort, error and task disruption, 28th ACM ACSAC, 2012, pp. 159168.

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► Testing Time

Scheme	Device	Classifiers	# Users	Testing Time (s)
Hold & Sign	Nexus 5	MLP	30	0.215 - 0.250
Lee & Lee ³	Nexus 5	SVM	8	20
Li et al ⁴ .	Motorolla Droid	Sliding patterns	75	0.648
Nickel et al ⁵ .	Motorolla Milestoon	KNN	36	30

 $^{4}\text{Li},$ X. Zhao, and G. Xue, Unobservable re-authentication for smartphones. in NDSS, 2013.

 5 C. Nickel, T. Wirtl, and C. Busch, Authentication of smartphone users based on the way they walk using k-nn algorithm, in 8th IEEE International Conference on Intelligent Information Hiding and Multimedia Signal Processing ($\frac{1}{H}$ H-MSP), 2012

SQA

 $^{^3 \}rm W.-H.$ Lee and R. B. Lee, Multi-sensor authentication to improve smartphone security, in International Conference on Information Systems Security and Privacy, 2015.

Power Consumption

⁶Sitova et al. "HMOG: New Behavioral Biometric Features for Continuous Authentication of Smartphone Users." IEEE Transactions on Information Forensics and Security 11.5 (2016): 877-892.

⁷https://play.google.com/store/apps/details?id=com.quicinc.trepn

Power Consumption

In order to check the overhead resulting from use of the application (in different steps), we terminated all the running applications and all Google services, switched off WiFi, Bluetooth, and cellular radios. The screen was kept running for the entire duration of the experiment with brightness at the lowest level and automatic brightness adjustment disabled⁶. We used Trepn⁷to profile power usage.

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⁸https://developer.qualcomm.com/blog/ mobile-apps-and-power-consumption-basics-part-1

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- our app consumes ≈ 1000 mW (for 35 attempts) all in the stages of decision making (less than the power consumed in one-minute of phone call(≈ 1054 mW⁸) during all the stages.

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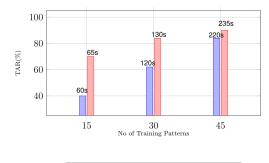
Usability

Trade-off between training & accuracy.

 $^{^9} https://www.usability.gov/how-to-and-tools/methods/system=usability-scale.html <math>\circ \circ \circ \circ \circ$

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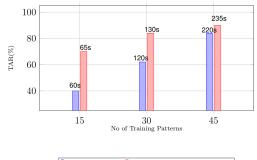


■ Full Features ■ RFE Reduced Features

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□ Full Features □ RFE Reduced Features

► System Usability Scale (SUS⁹) Evaluation - 68.3% (72% or higher is considered good.)

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► This one-minute video shows the working of Hold & Sign.

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• Extend number of testers and run tests on the field.

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- Extend number of testers and run tests on the field.
- Reduce training time without compromising (too much) security

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Unplanned Situations

- Extend number of testers and run tests on the field.
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- Unplanned Situations
- Assess its robustness to multiple attacks.

- Extend number of testers and run tests on the field.
- Reduce training time without compromising (too much) security
- Unplanned Situations
- Assess its robustness to multiple attacks.
- Variability depending on OS (iOS) and on HW (different phones, sensors)

Thank You!

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