## Whose Move is it Anyway? Authenticating Smart Wearable Devices Using Unique Head Movement Patterns

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# **Project Highlights**

- We design HeadBanger
- We achieve accurate user authentication
  - Experiment with 95 subjects
  - High TPR and Low FAR
  - Robust against Attack
- We build an running app on Google Glass



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## Personal Information Is in Your Wearables !



## **Existing Approaches: Indirect Authentication**









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## **Existing Approaches: Built-in Authentication**

### **X**Limited Input Area

**X**Long Input Period

### **X**Not Intuitive Pattern







## **Existing Approaches: Biometrics**

### **Physical Biometrics :**

- Additional Hardware
- Not always applicable for head-mounted device









## **Existing Approaches: Biometrics**

#### Behavioral Biometrics: Walking gait arm

Walking gait, arm swing, finger gesture, etc.



### But, for head movements:

- Hard to collect longterm movement patterns
- Do not have high
  Degree of Freedom







## Challenges







## Music-induced Head Movement







## Music-induced Head Movement



- 30 Subjects
- Same movement and same music 30 times

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## **Response Time Is Not Enough**



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## Headbanger Rationales







## Headbanger Overview







## Data Filtering



- Accelerometer Contains High Frequency Noise
- Head Movement is at Low Frequency ( < 5 Hz)





## Headbanger Overview





## Dynamic Time Warping

Time-normalized distance between A and B :  $D(\mathbf{A}, \mathbf{B}) = \left| \frac{\sum_{s=1}^{k} d(p_s) \cdot w_s}{\sum_{s=1}^{k} w_s} \right|$  $d(p_s)$ : distance between  $i_s$  and  $j_s$  $w_s > 0$ : weighting coefficient. *<u>Best alignment path</u>* between **A** and **B**:  $P_1 = \underset{D}{\operatorname{arg\,min}}(D(\mathbf{A}, \mathbf{B})).$ 



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Time Series B

## Headbanger Overview







# Reduce the Computing Overhead

### Choose Representative Samples

- Compute the average distance to other samples
- Rank the samples based on their average distance
- Threshold can be expressed as: Threshold =  $\overline{d_k}$  + n ×  $\sigma_k$



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## Repeatability & Similarity Experiment

## **Objectives:**

- True user can login with high probability
- Different user do different movement
- Low computing cost

### Setup:

30 subjects are involved

- Each of them design its own pattern
- Each of them performs it 40 times

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## **Evaluation Metrics**

**D** True Positive Rate

**D** False Accepted Rate

**\Box** Equal Error Rate EER = TRR(n) = FAR(n)





## Impact of Distance Algorithm



## Impact of Voting Scheme



## Impact of Training Size



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## Impact of Music Duration



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## Overall



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## Let's Attack it!







## Let's Attack It!







## Attack Results

Target	# of Attackers	# of Successful Attackers	Average # of Trials before 1 <sup>st</sup> Successful Attack	FAR (%)
А	12	7	10	15.83
В	13	3	14	2.77
С	12	3	17	2.72
Overall	38	13	13	6.94



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## Prototyping

- **D** Google Glass Development Kit
- Java Speech Tool Kit
- **□** Fast DTW:  $O(n^2) \rightarrow O(n)$
- Task pipelining

Music Cue Duration (s)	Data processing latency (s)
10	1.93
6	1.15
5	0.88





## Conclusion

We design Headbanger

# We Conduct Intensive Experiment Repeatability, Robustness

### ■ We develop a running App on Google Glass





## Thank you! Questions?



