

## Enhanced In-air Signature Verification via Hand Skeleton Tracking to Defeat Robot-level Replays

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## **Current Authentication Methods**

#### **C** Knowledge-based secret

- PIN/password
- Patterns



#### **Static Authentication Input:**

#### Physiologic

- Fingerprint
- Iris

- Can be lost, stolen, forgotten
- Can be spoofed and replicated





## **Emerging Behavioral Biometric Authentication**

□ Verifying dynamic motion characteristics

- Gait patheme
  Hard to be copied or reproduced
- Keystrol Less dependent on dedicated hardware

In-air 3D signature is one representative of behavioral biometrics







## **In-air 3D signature**

- Representative behavioral biometric authentication
- Inherits the traditional signature's legal effect
- Enhanced security
  - 3D handwriting curves
  - Signing behaviors
- Eliminates the need for a writing surface
- Supported by existing hand-tracking interfaces







## **Current hand-tracking interfaces**



## Have not considered threats of **3D printing and motion-copy robots**







## **Vulnerabilities of Hand Tracking Interfaces**









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## **Defense Against Robot Replays**

## Current robots are still not able to copy hand-joint-level motions

#### **Novel hand joint-level authentication**

- Extend the dimension of in-air signatures from a single point to multiple hand joints
- Leverage the hand's kinematic structure motions to prevent robot replays







## **3D Hand Skeleton Signature System**







## **Hand Skeleton Motion Data Extraction**

□ 3D landmarks of a hand captured by visual sensor

**Examine 3D in-air signatures based on a novel graphical representation** 







## **Multi-joint Data Normalization and Alignment**

#### Camera placement

- Predefined direction alignment
- Hand size normalization

#### □ Inconsistent signature curve

• Trajectory normalization

#### □ Varying signing speed

• Trajectory interpolation







## **Joint-level Motion Features Presentation**

#### □ Hand skeleton signature

- Signature trajectory
- Signing behavior
- Hand geometry
- □ Integrate time information
- Examine from 3 different perspectives



![](_page_11_Picture_9.jpeg)

![](_page_12_Picture_0.jpeg)

## **Three View-based Biometric Feature Presentation**

#### Presenting spatial information

Three-view projection

## Presenting temporal Information

Gradient color from light to dark

#### □ Joint significance weight assignment

![](_page_12_Picture_8.jpeg)

![](_page_12_Picture_9.jpeg)

![](_page_12_Picture_10.jpeg)

![](_page_12_Picture_11.jpeg)

![](_page_12_Picture_12.jpeg)

![](_page_12_Picture_13.jpeg)

![](_page_12_Picture_14.jpeg)

![](_page_12_Picture_15.jpeg)

![](_page_12_Picture_16.jpeg)

Front

![](_page_12_Picture_17.jpeg)

Side

![](_page_12_Picture_18.jpeg)

![](_page_12_Picture_19.jpeg)

![](_page_12_Picture_20.jpeg)

50

100

Erame Erame

200

250

300

1

## **Inter-joint Motion Feature Derivation**

#### **Relative distance relationships between hand joints**

- Distinguish users
- Indicate human or robot replay

80

40

20

#### Inter-joint motion profile: variance over time

![](_page_13_Figure_6.jpeg)

![](_page_13_Figure_7.jpeg)

![](_page_13_Figure_8.jpeg)

140

10

(c) User 1 replayed by a robot & 3D-printed hand.

	Joint	Avg. Score	Joint	Avg. Score
ĺ	0	1.82	11	3.22
	1	2.23	12	4.62
	2	2.55	13	1.02
	3	3.59	14	1.40
	4	4.51	15	2.28
	5	0.75	16	3.09
	6	1.57	17	1.35
	7	4.43	18	1.47
	8	8.38	19	2.01
	9	0.78	20	2.41

1.40

![](_page_13_Picture_11.jpeg)

![](_page_13_Picture_12.jpeg)

![](_page_13_Picture_13.jpeg)

![](_page_14_Picture_0.jpeg)

## **CNN-based Authentication Algorithm**

![](_page_14_Figure_2.jpeg)

![](_page_14_Picture_3.jpeg)

![](_page_15_Picture_0.jpeg)

## **Experimental Setup**

#### **Commercial hand-tracking interfaces**

- Google MediaPipe
- Leap Motion

#### Off-the-shelf Devices

- Regular RGB camera (ELECOM Webcam)
- Depth camera (Leap Motion Controller)

#### Data collection

- 25 participants
- Name initials and ``ABC"

#### **Q** Robot replay attack Implementation

- Hidden camera for eavesdropping
- A low-cost robotic arm for replay PincherX 150
- 3D-printed hand of the user

![](_page_15_Picture_15.jpeg)

![](_page_15_Picture_16.jpeg)

![](_page_15_Picture_17.jpeg)

## **User Verification Performance**

![](_page_16_Figure_1.jpeg)

#### **Different Motion Capture Devices**

Multi-joint vs. Single-point

Joint vs. Inter-joint

![](_page_16_Picture_5.jpeg)

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![](_page_17_Picture_0.jpeg)

## **Enrollment Efforts**

Increasing the training data size improves the system's performance but requires higher enrollment efforts.

![](_page_17_Figure_3.jpeg)

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![](_page_18_Picture_0.jpeg)

## **Security Under Impersonation Attacks**

#### □ Traditional single-point signature

- Relatively easy to imitate by an adversary
- Suffers highly from the visual tracking errors incurred by occlusion or self-occlusion

0.3 0.275 82.0 gt Bate Ealse Acceptance E 2.0 E 0.051 0.02 0 Joints + Int Single Point Joints Signature

Multiple joints compensate for the partially occluded hand and examines hand skeletons' inherent behaviors

![](_page_18_Picture_7.jpeg)

![](_page_19_Picture_0.jpeg)

## **Performance Under Robot Replay Attacks**

![](_page_19_Figure_2.jpeg)

![](_page_19_Picture_3.jpeg)

![](_page_20_Picture_0.jpeg)

## Conclusion

- Introduce the 3D hand skeleton signature verification system to address emerging motion-copy robot threats
- Propose a novel three-view presentation method to describe hand skeleton motions
- Develop a CNN-based algorithm to verify in-air signatures at both the hand joint level and inter-joint level
- Implement a physical motion-copy robotic arm and demonstrate a new attack that exploits robots and 3D printing
- Experiments show 3D hand skeleton signature system achieves high performance and defeats robot replay attacks

![](_page_20_Picture_7.jpeg)

![](_page_21_Picture_0.jpeg)

Mobile and Internet SecuriTy (MIST) Lab

![](_page_21_Picture_2.jpeg)

![](_page_22_Picture_0.jpeg)

# Back up

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![](_page_23_Picture_0.jpeg)

# **Signing Behavior**

□ Varying speed at turning

□ Minute non-straight line

![](_page_23_Figure_4.jpeg)

![](_page_23_Picture_5.jpeg)

#### Simulated Robot Relay

 Virtual hand model that precisely followed the user's hand motion data

#### Physical Robot Replay

- Access to both the user's 3D hand skeleton model and signature trajectory samples

3D scanner and 3D printer

# **Attack Setup**

## Impersonation Attack

- Obtain the user's name and signing behavior data
- Observe and mimic

![](_page_24_Picture_14.jpeg)

![](_page_24_Picture_15.jpeg)

![](_page_24_Picture_16.jpeg)

![](_page_25_Picture_0.jpeg)

## **Verification Performance: Standardized Content**

Write out the letters "ABC"

![](_page_25_Figure_3.jpeg)

![](_page_25_Picture_4.jpeg)

![](_page_26_Picture_0.jpeg)

# **Performance Under Simulated Attacks**

![](_page_26_Figure_2.jpeg)

![](_page_26_Picture_3.jpeg)

![](_page_27_Picture_0.jpeg)

# **Question List**

## **3D** input?

- CNN is most efficient with 2D images
- Three different perspective, like in 3D modeling
- Enables us to examine each view more closely

#### **Robot capability?**

- Advancing attack vs. defense
- Commercial devices consider cost

#### Light condition?

![](_page_27_Picture_10.jpeg)