

# Siren Song: Acoustic Attacks on Pose Estimation in XR Headsets

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#### Motivation

XR has applications across many areas:

#### But...what if pose estimation is under attack?

• Importance of pose estimation:



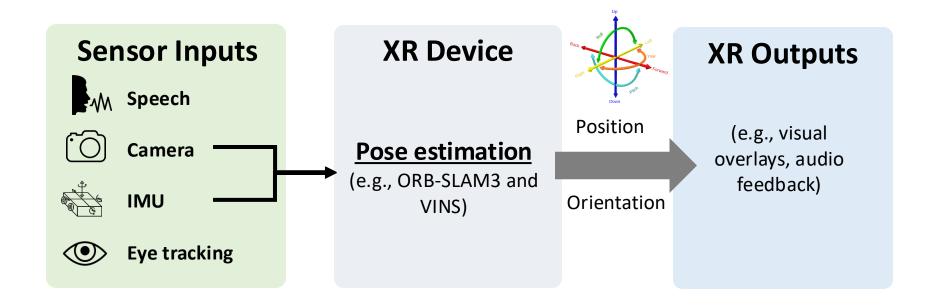
Rendering





#### Background: Pose estimation in XR

• XR devices fuse <u>IMU</u> and camera data to estimate the headset's <u>location</u> and <u>orientation</u>.



### Background: Acoustic Attack

- Plenty of acoustic attacks on drones, vehicles, and phones.
  - Key Challenges:
    - Noninvasive attack setup.
    - XR-specific pipeline.
    - Impact on user experience.

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This paper is in

August

ready known to cause malfunctions by dis gyroscopes. However, an open question ren beyond denial of service attacks to acl control of sensor outputs. Our work inv acoustic injection attacks can damage the popular type of sensor: the capacitive M Spoofing such sensors with intentional enables an out-of-spec pathway for attack digital values to microprocessors and en 24th US blindly trust the unvalidated integrity of contributions include (1) modeling the acoustic interference on MEMS acceleron the circuit-level security flaws that cause measuring acoustic injection attacks on M

> as well as systems that employ on these sensors, and (3) two software-only defenses that mitigate many of the risks to the integrity of MEMS accelerometer outputs.

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> vulnerabilities, but fewer methodologies exist hardware domain

It is already known that acoustic inte cause denial of service (DoS) attacks against

alleviate image blurring caused by camera such a trend opens a new attack surface. Thi a system-level vulnerability resulting from th the emerging image stabilizer hardware susce manipulation and the object detection algoadversarial examples. By emitting deliberately tic signals, an adversary can control the out sensor, which triggers unnecessary motion c results in a blurred image, even if the came blurred images can then induce object miscla ing safety-critical decision making. We mod of such acoustic manipulation and design an that can accomplish three types of attacks, i

ing, and altering objects. Evaluation results demonstrate and effectiveness of our attacks against four academic object detectors (YOLO V3/V4/V5 and Fast R-CNN), and one commercial detector (Apollo). We further introduce the concept of AMpLe attacks, a new class of system-level security vulnerabilities resulting from a combination of adversarial machine learning and physics-based injection of information-carrying signals

#### This paper is included in the Proceedings of the 27th USENIX Security Symposium.

August 15–17, 2018 • Baltimore, MD, USA

978-1-939133-04-5

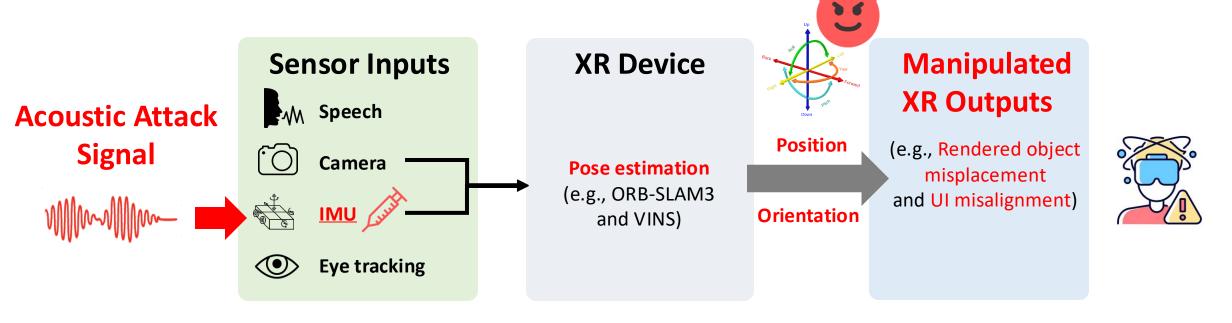
of the emergent image stabilization hardware susceptible to acoustic manipulation and the object detection algorithms subject to adversarial examples, and design Poltergeist attacks (in short, PG attacks) that exploit such vulnerabilities. Unlike existing work that focused on altering what the

main sensors (e.g., CMOS sensors) perceive by changing the visual appearance of an object [12], [38], [59], [27] or by

**Background** Threat model User study **Attacks** 

### Background: Acoustic Attack

 Acoustic signals trigger MEMS resonance, corrupting IMU motion data.



Background Threat model Attacks Mitigation 5

#### Threat model

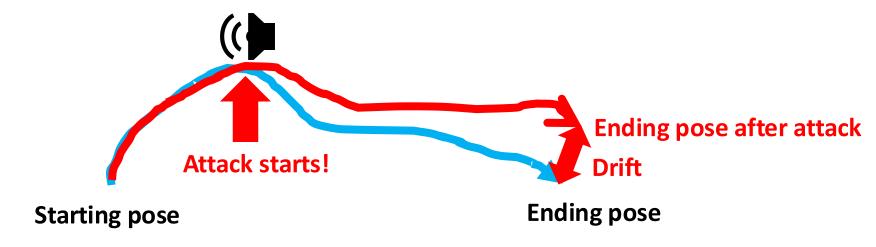


- Attacker's capability:
  - Play sound near victim no physical or sensor access.
- Attacker's resources:
  - Profile same-model sensors to find <u>resonant frequencies</u>.

• Attacker's goal: benefit attacker (cheat) or harm victim (disrupt / induce errors).

### Attacks on open-source systems

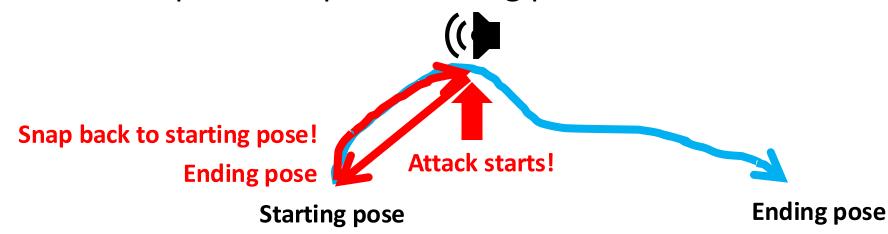
- Tested systems: ILLIXR [1] and ORB-SLAM3 [2].
- Slow drifting attack when IMU bias is <u>small</u>.
  - Small and slow pose drift.



[1] Huzaifa, Muhammad, et al. "ILLIXR: Enabling end-to-end extended reality research." 2021 IEEE International Symposium on Workload Characterization (IISWC). IEEE, 2021. [2] Campos, Carlos, et al. "Orb-slam3: An accurate open-source library for visual, visual-inertial, and multimap slam." IEEE transactions on robotics 37.6 (2021): 1874-1890.

### Attacks on open-source systems

- Slow drifting attack when IMU bias is <u>small</u>.
  - Steady and small pose bias.
- Snapback attack when IMU bias is <u>large</u>.
  - Sudden pose "snap" to starting point.

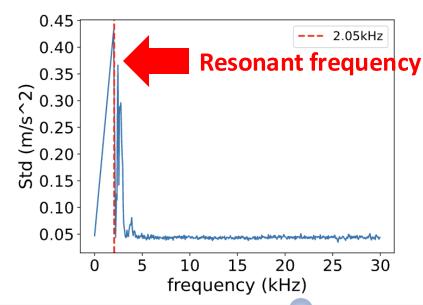


[1] Huzaifa, Muhammad, et al. "ILLIXR: Enabling end-to-end extended reality research." 2021 IEEE International Symposium on Workload Characterization (IISWC). IEEE, 2021. [2] Campos, Carlos, et al. "Orb-slam3: An accurate open-source library for visual, visual-inertial, and multimap slam." IEEE transactions on robotics 37.6 (2021): 1874-1890.

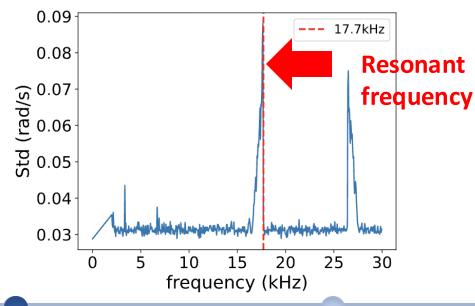
#### Attacks on Microsoft HoloLens

- Resonant frequencies of the HoloLens IMU.
- Snapback attack (Success rate: > 90%).

#### The accelerometer is vulnerable at 2.05 kHz.



#### **Gyroscope** is vulnerable at 17.7 kHz.



#### Attack 1: Manipulating user input (harm to user)

• Attack motivation: Many XR games map head motion to game controls.

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- Attack motivation: Many XR games map head motion to game controls.
- Method (attack design): We place a HoloLens on a remotecontrolled car and inject acoustic signals targeting the IMU.

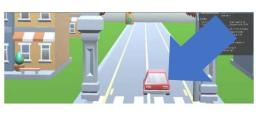


#### Attack 1: Manipulating user input (harm to user)

#### Outcome (attack result):

- The car initially moves but then shifts slightly.
- Due to snapback, its estimated position resets to the start the virtual car is stuck and the game is unusable for the victim.

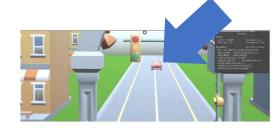
Benign



(a) Benign, time=0



(b) Benign, time=1



(c) Benign, time=2



• Motivation: World-anchored UI (e.g., virtual keyboard) drifting breaks input and can destroy user actions.

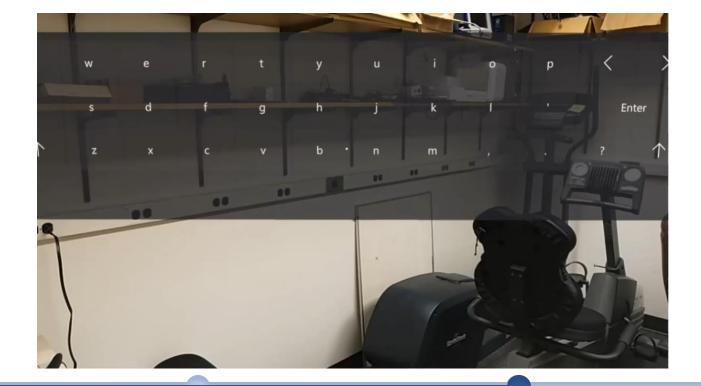
Benign:
User can correctly
type words on
virtual keyboard.



The board is anchored in the fixed location.

• Outcome: The virtual keyboard drifts or disappears

Attack: user mis-clicks or cannot type.



Virtual keyboard drifts up and down, making user mis-click.

- Small user study: goals & setup.
  - RQ1 (Usability degradation): How does the acoustic attack affect usability and task performance?
  - **RQ2** (Impact of the acoustic signal): Under what conditions is the acoustic signal more comfortable?
- Participants & environment: 5 volunteers, HoloLens 2, click-jacking app, 30-minute sessions.
  - RQ1: A timed typing task (three 3-letter words).
  - RQ2: Earmuffs or listening to music.

• **Performance drop:** task completion fell from 100% (benign) to ~46.7% (attack).

• Slower & harder: average completion time rises (increasing from 6s to 16s); all participants said typing was harder under attack.

• **Practical note:** listening to music made the attack more tolerable and still effective — <u>stealthier</u> attack vector.

#### Conclusion

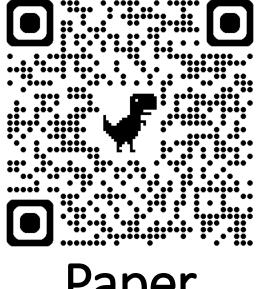
• First to study acoustic attacks on XR pose estimation.

Identified two novel effects on XR pose estimation pipelines.

Designed and demonstrated four proof-of-concept attacks on XR headsets.

Conducted user study showing real usability degradation.

• Discussed defense strategies: hardware filtering, software re-localization.



### Paper

## Thank you! Any questions?

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