

Eavesdropping on Controller Acoustic Emanation for Keystroke Inference Attack in Virtual Reality

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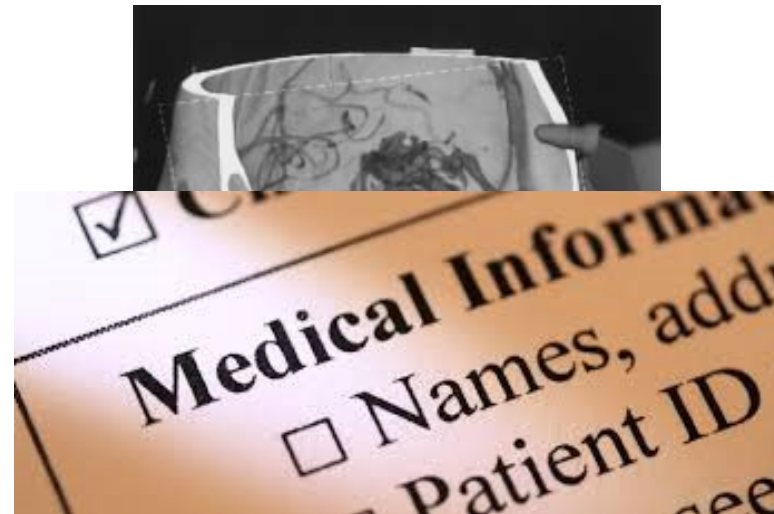


Diverse Applications

Gaming



Healthcare



Training



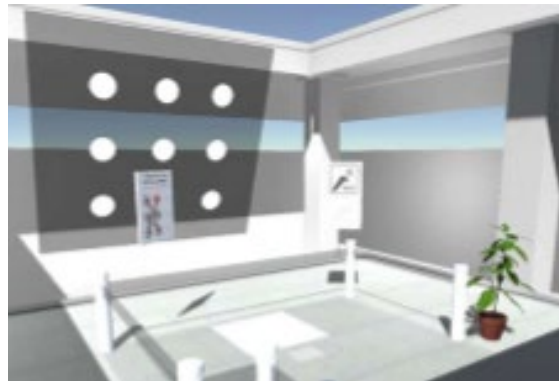
For Secure VR Systems

- Protect the sensitive data.

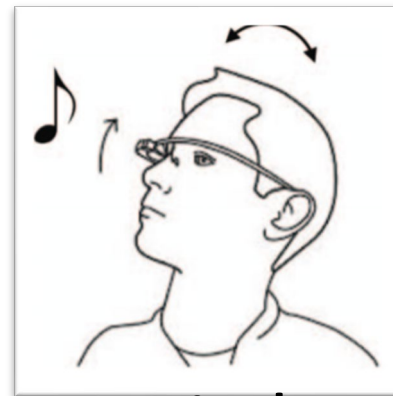


Problems of Existing VR Data Protection Works

- Existing protection:
 - Verify users' identity using authentication schemes.
 - Prevent unauthorized access when attackers physically possess HMD.



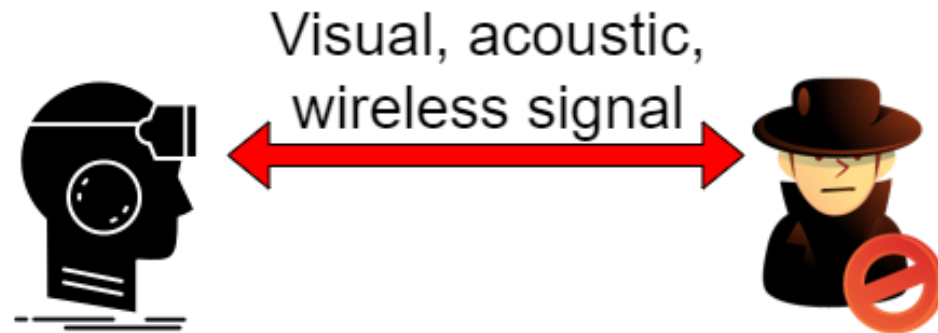
Knowledge-based



Biometric-based

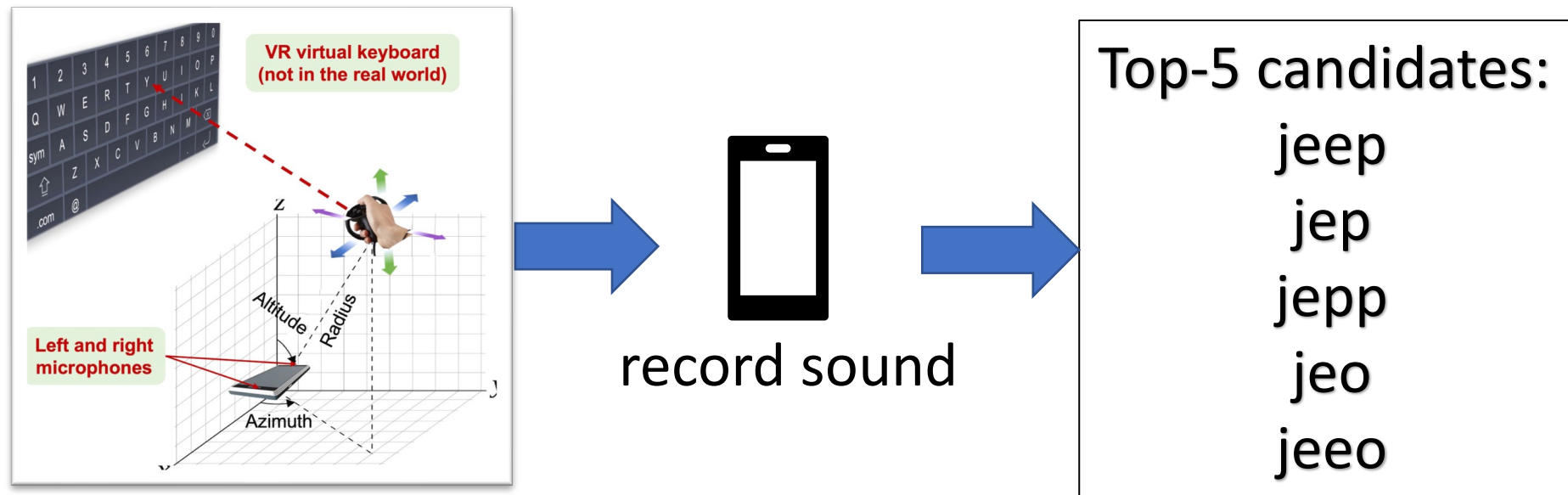
Problems of Existing VR Data Protection Works

- Interaction between users and HMD is exposed to the public.
 - Attackers can record user interaction through side channels.
 - Video, wireless signal...
 - Attackers can infer sensitive input without possession of the HMD.



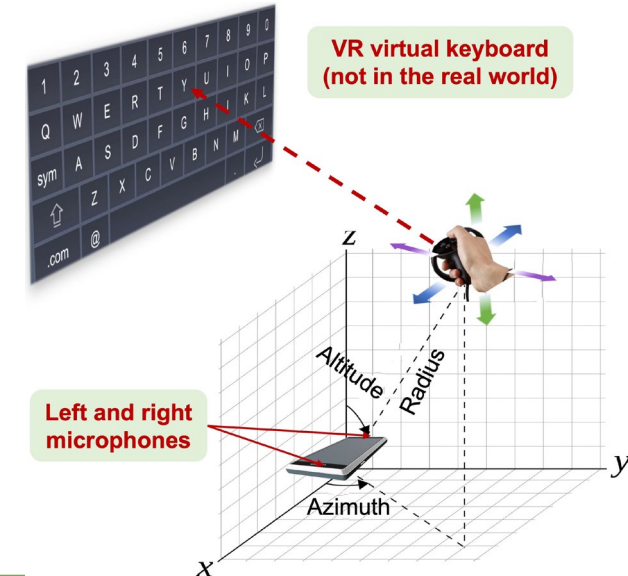
Keystroke Inference and Acoustic Signal

- We expose a VR keystroke inference attack exploiting the acoustic emanations from the controller.
 - Microphone records sounds anywhere around the victim.
 - More flexible placement than existing side channels.



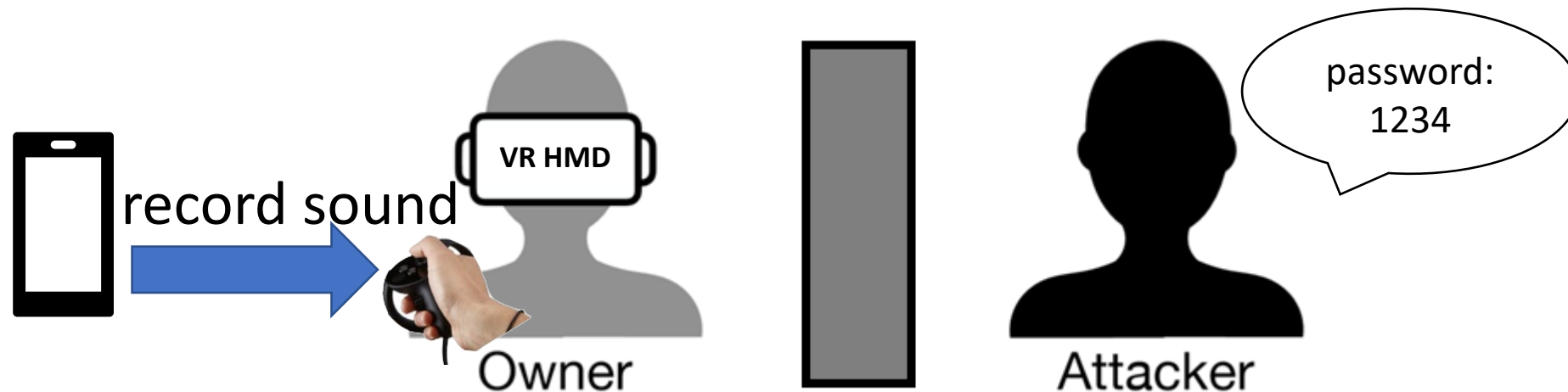
Keystroke Inference and Acoustic Signal

- Controller-based keystroke method in VR:
 - Users move a controller to navigate a virtual cursor through a virtual keyboard.
 - Users click the confirm button to commit keystrokes.
- Attackers place a malicious smartphone nearby.
 - The controller emits keystroke clicking sounds at various locations.
 - The sound arrive the smartphone from different directions.
 - Each keystroke uniquely defines its clicking sound direction.



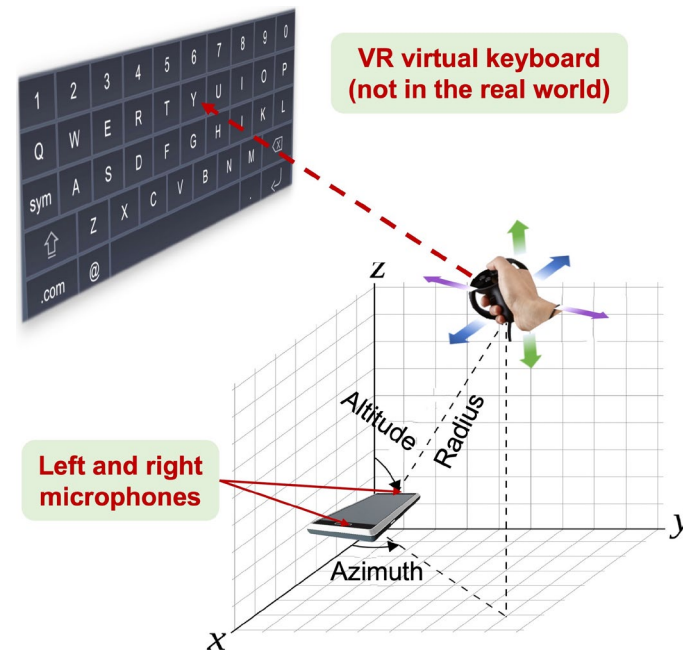
Threat Model

- The user enters sensitive information in VR in a confined setting
 - e.g., a shared table in a library.
- A malicious smartphone is placed nearby.
- The smartphone records the acoustic emanations from the controller.
- The attacker knows the layout of the virtual keyboard.



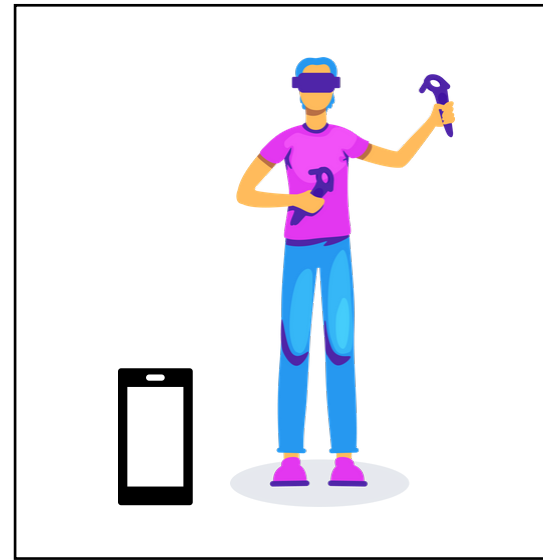
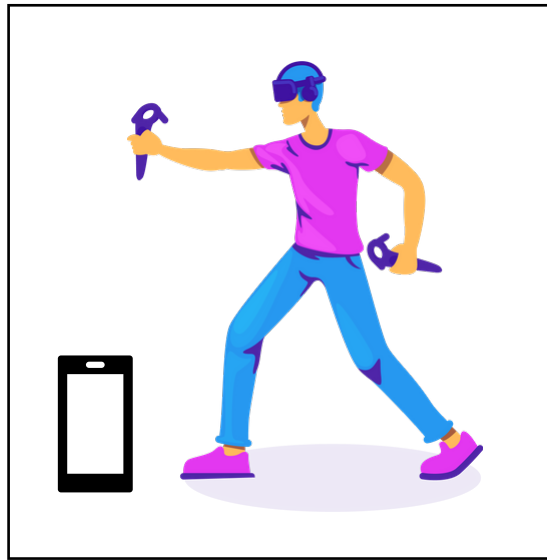
Challenge 1 – Sound Source in 3D Space

- It is difficult to differentiate the clicking sounds.
 - Sound source locations are distributed in 3D space rather than 2D plane*.
 - The omnidirectional smartphone microphone cannot differentiate keystroke sounds.



Challenge 2 – Various User Microphone Placement

- The relative position and orientation between the user and the microphones vary in different attack scenarios.
 - The direction of arrival (DOA) of the same key's keystroke sounds varies.
 - The mapping between DOAs and keys varies.



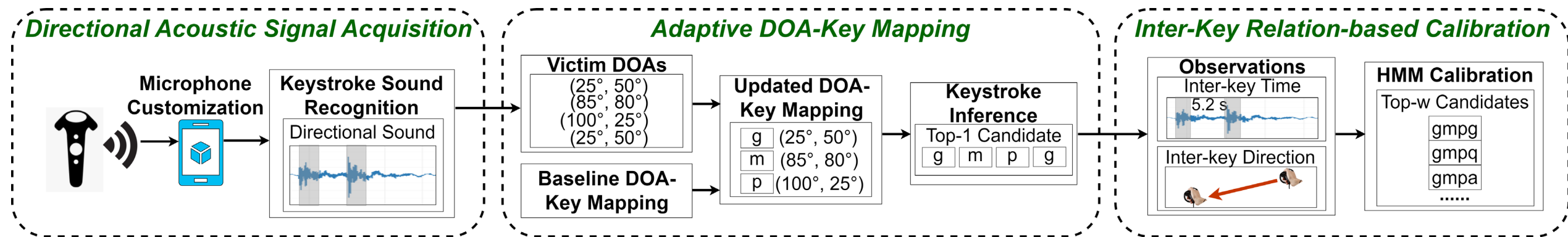
Challenge 3 – Mapping Error

- No strict one-to-one mapping between keystroke sound and keys.
 - Users may rotate the controller to navigate the cursor.
 - Keystroke sound with the same DOA results in different keys.



Heimdall System Architecture

- Module 1: record differentiable keystroke sound using directional microphones.
- Module 2: adapt DOA-Key mapping to the attack case.
- Module 3: correct the mapping errors using a Hidden Markov Model (HMM).



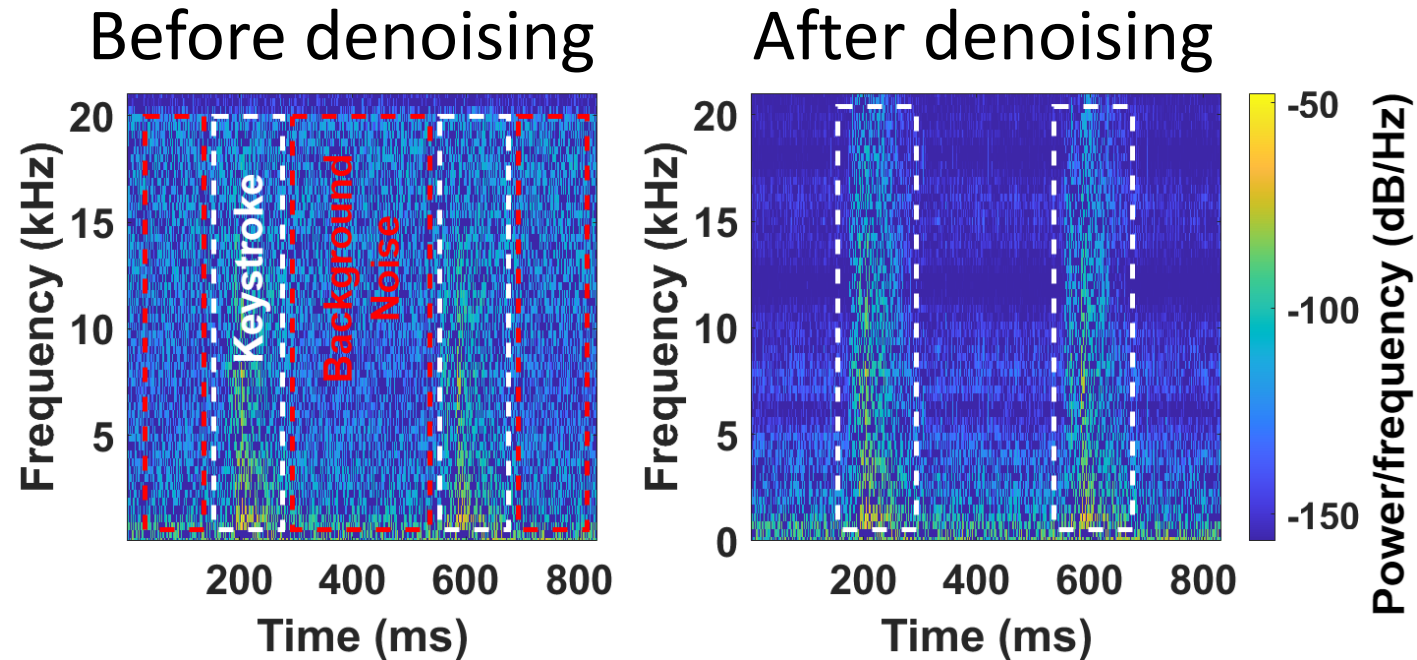
Module 1 – Hardware Design

- Convert the omnidirectional microphones into directional ones.
 - Inspired by a shotgun directional microphone.
 - Two tubes with side slots are attached to the microphones.
 - The microphones change the intensity and phase of recorded sounds based on their DOA.



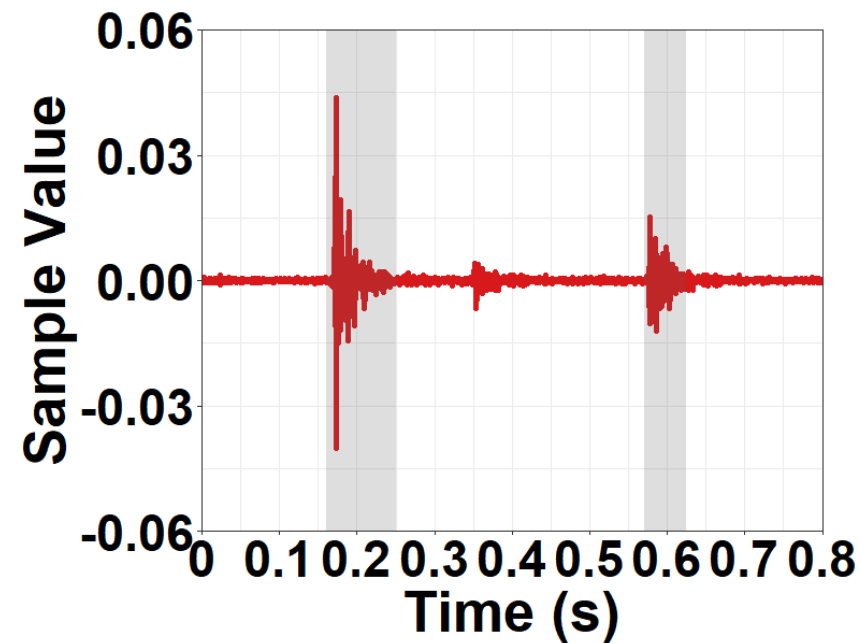
Module 1 – Acoustic Signal Acquisition

- Remove ambient noise using wavelet denoising*.



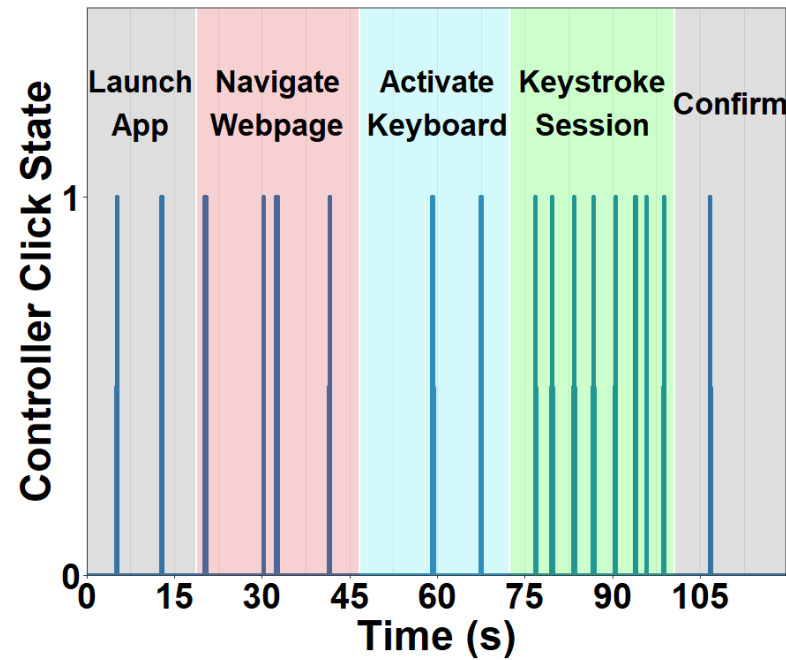
Module 1 – Signal Processing

- Signal peaks exceeding a threshold are extracted as keystroke clicking sounds.
 - Optimal threshold value is 0.028.



Module 1 – Keystroke Session Recognition

- We identify keystroke session based on the controller click frequency*.



Module 2 – Keystroke Mapping

- We derive a baseline DOA-Key mapping table.
 - The mapping stores DOA and their corresponding keys (26 English letters and 10 digits).
- We update the baseline DOA-key mapping to match the DOAs in the attack.

DOA [azimuth altitude]	Key
[-80° 70°]	1
[-60° 70°]	2
...	...
[40° -20°]	N
[60° -20°]	M

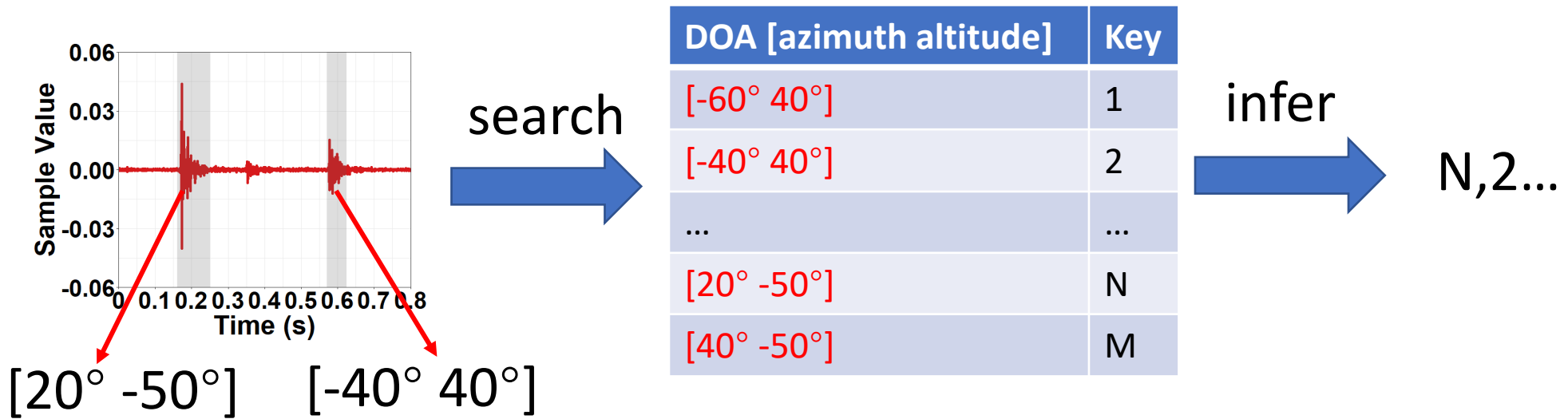
update



DOA [azimuth altitude]	Key
[-60° 40°]	1
[-40° 40°]	2
...	...
[20° -50°]	N
[40° -50°]	M

Module 2 – Keystroke Mapping

- We infer keystrokes in attack by looking up their DOA in the updated mapping.

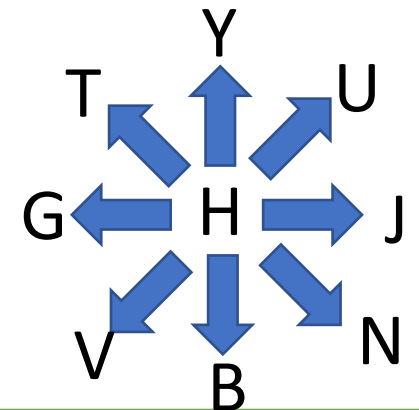
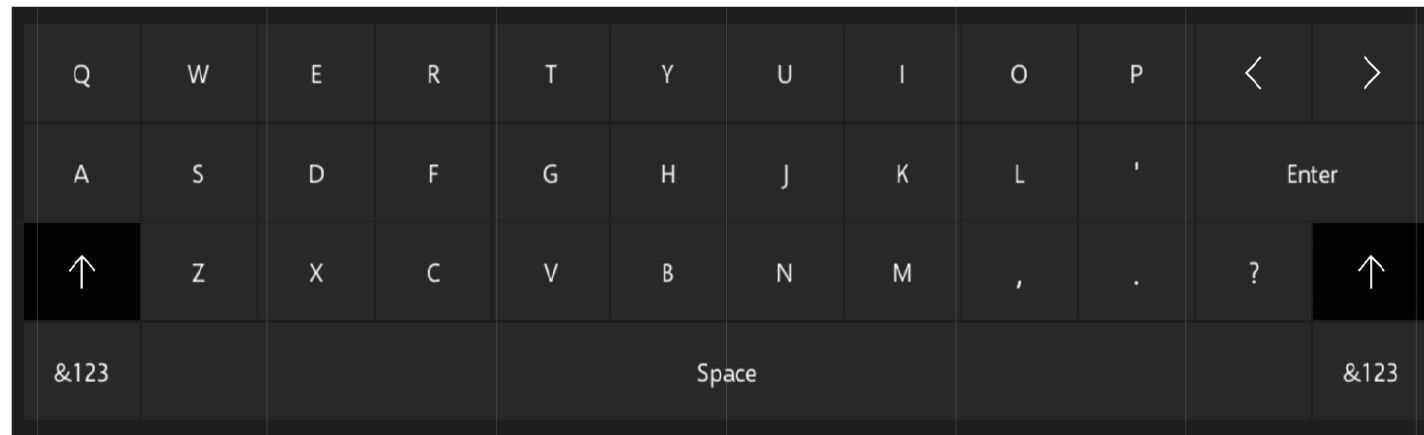


Module 3 – Transition between Keystrokes

- Transition between keystrokes contains information on the keys.
 - Time interval between two successive keystrokes depends on their inter-distance.
 - Controller moving direction committing two keystrokes depends on keys location on the keyboard.

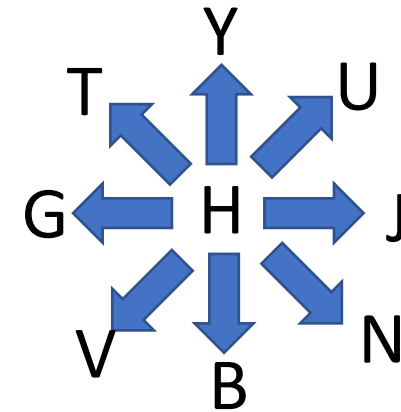
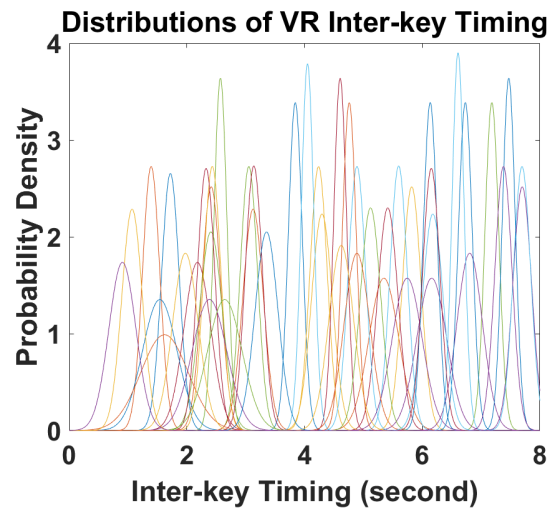
H $\xrightarrow{0.8s}$ J

H $\xrightarrow{2.2s}$ P



Module 3 – Keystroke Mapping Correction

- A Hidden Markov Model (HMM) models the keystroke transition.
 - The inter-key time and controller moving direction are observations.
 - The keys are hidden state.



Derive inter-key time from acoustic signals.

Derive inter-key directions based on DOA.

Module 3 – Keystroke Mapping Correction

- We use the HMM to correct mapping errors.
 - The model generates a list of key sequences given on the inter-key time and direction.
 - We select the sequences with high probability and similarity with the original mapping result.

Mapping result: H2GM



HMM generate

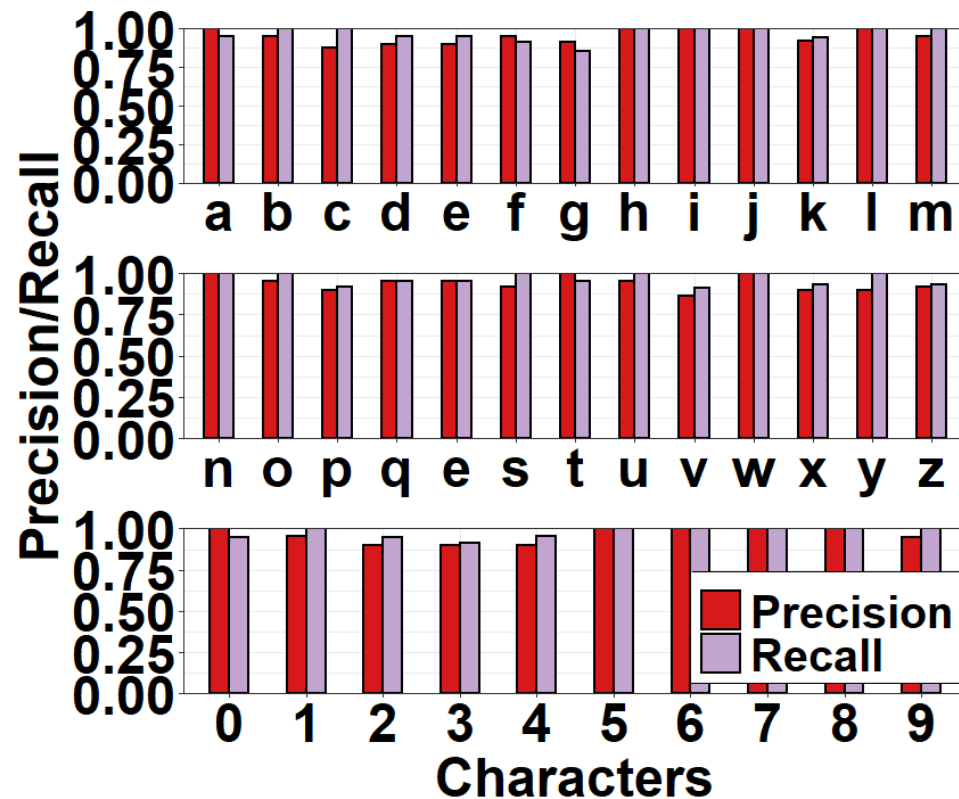
Top-5 candidate list
HWGM
H2HM
H3GM
H2GN
HWGN

Experiment Setup

- 30 participants.
 - They first take a practice session: input several keys as instructed.
 - Each of them inputs 45 passwords for testing.
- Length of passwords ranges from 4 to 8 characters.
 - Passwords are the most common ones*.

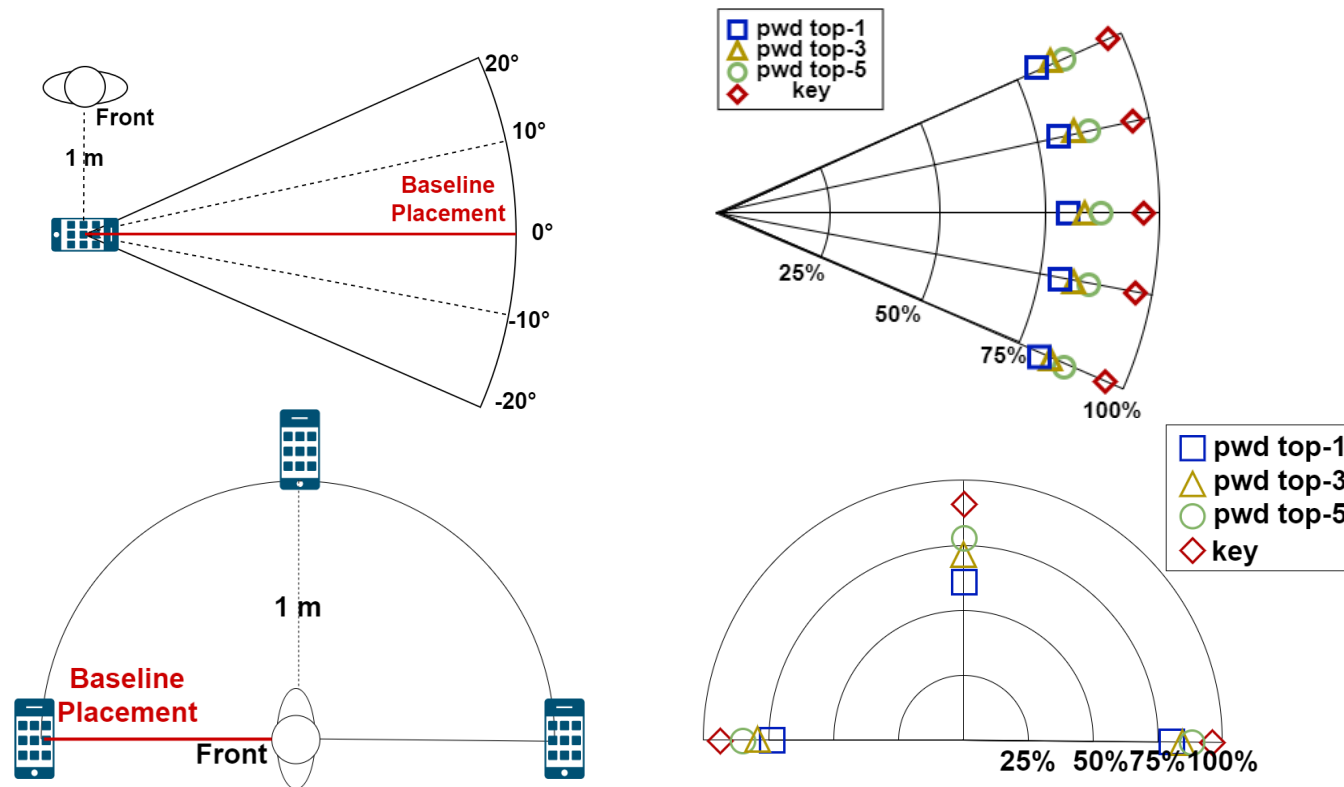
Experiment – Benchmark for All Characters

- The average precision and recall across characters reach 95.14% and 96.29%.



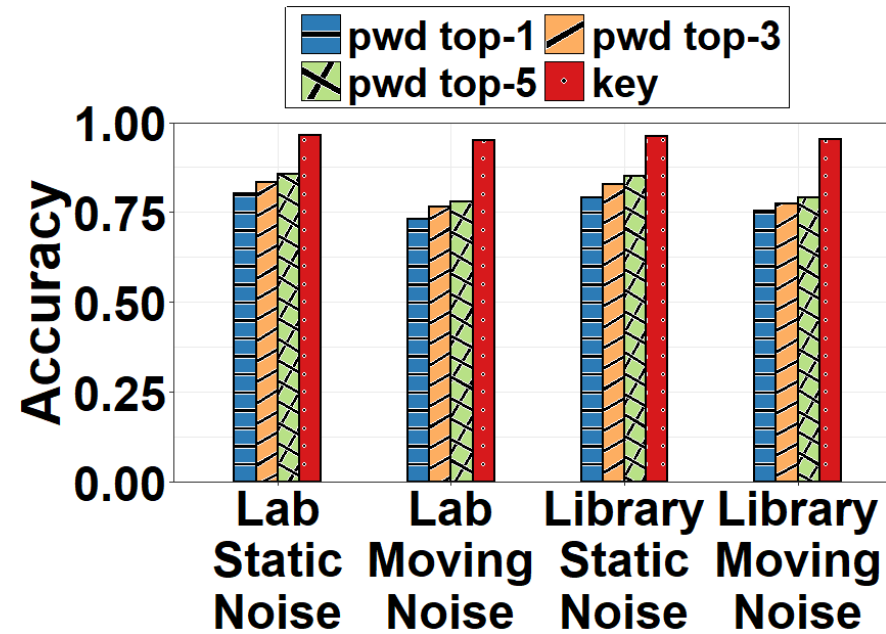
Experiment – Smartphone Placement

- The top-w and key accuracy are consistent across different user-smartphone placements.



Experiment – Different Environment

- The attack can be generalized to different scenarios.
 - Static noise: desktop fans and air conditioners.
 - Moving noise: people talking and walking.



Conclusion

- We expose a security threat in VR systems that allow attackers to infer input by analyzing the acoustic emanations from the controller.
- We propose Heimdall, a placement-flexible acoustic keystroke inference attack in VR.
- We extensively evaluate the Heimdall system in terms of keystroke inference accuracy and its robustness.

Thank you