LocIn: Inferring Semantic Location from Spatial Maps in Mixed Reality

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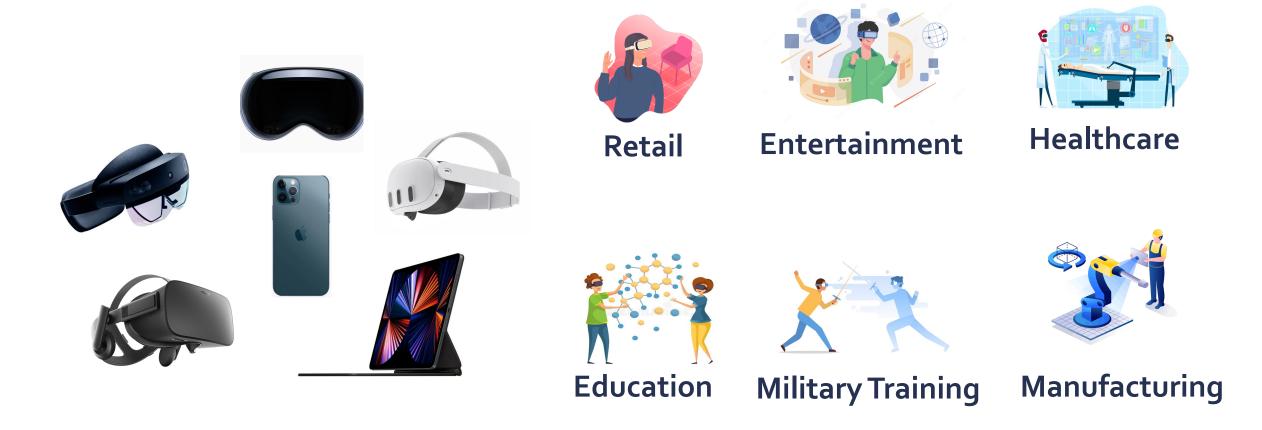






Mixed reality (MR)

MR devices allow users to view and interact with real and virtual content in their physical environment.

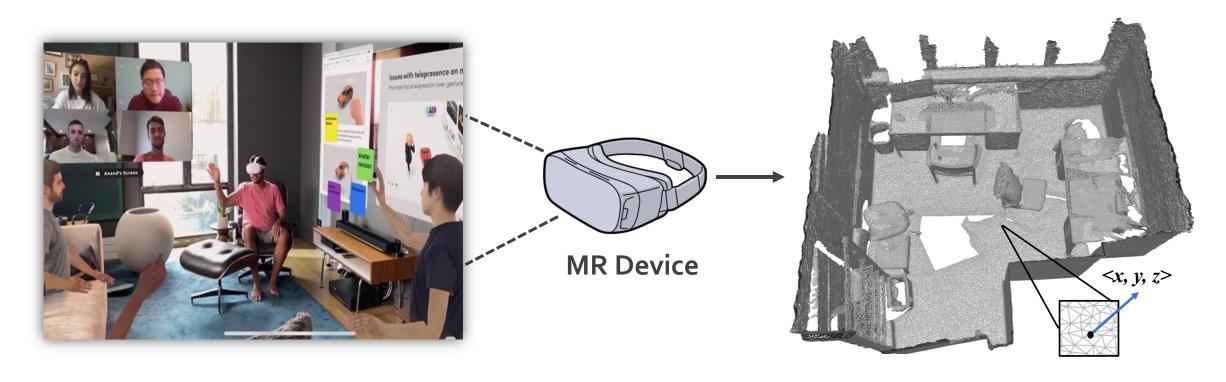


3D spatial maps in mixed reality

3D spatial maps - measure the distance between the device and points in the real environment.

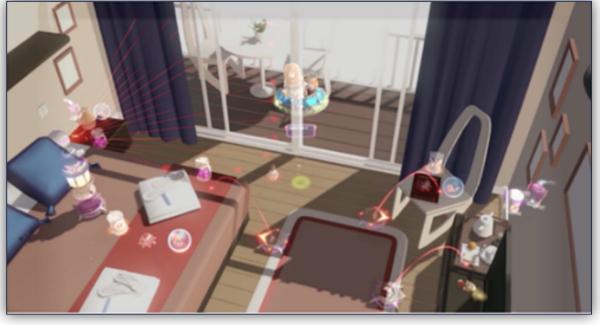
User's environment

Spatial map of user's environment



3D spatial maps in mixed reality







Object Placement

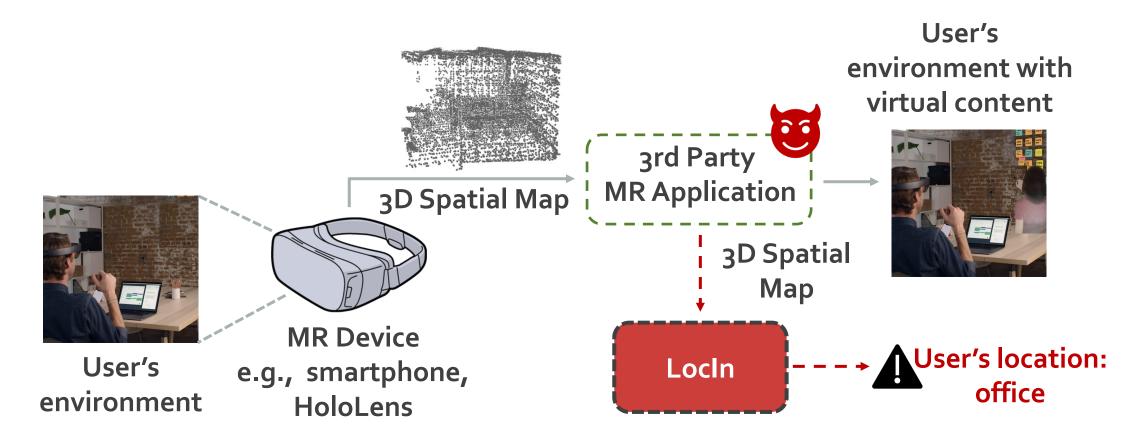
Occlusion Detection

Navigation

Source: https://learn.microsoft.com/en-us/windows/mixed-reality/design/spatial-mapping

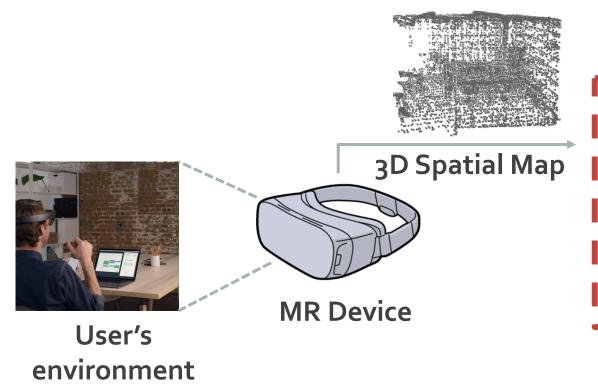
LocIn: location inference from spatial maps in MR

Key insight: A malicious app can exploit the 3D spatial map of the user's environment to infer user's indoor locations i.e., semantic location.



Key Insight

Threat model



3rd Party MR Application

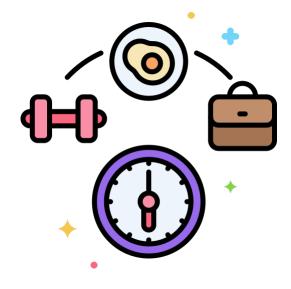
- Aware of device type e.g., HoloLens/iPad
- Only requires camera access.
- Only uses point cloud → no color or normal vector information is needed.

3

Security and privacy implications



Robbery and Physical Attacks

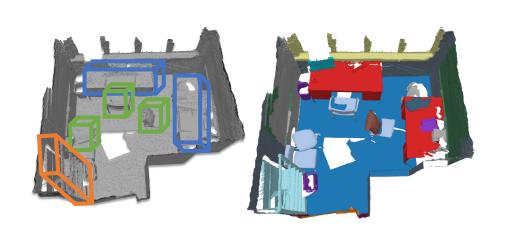


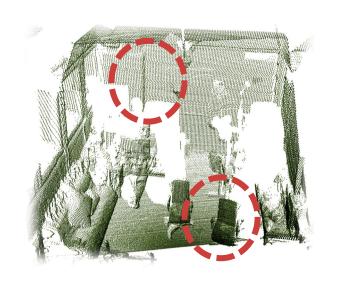
User Routine and Personalized Ads



Socio-economic and Family Status

Extracting location cues from spatial maps





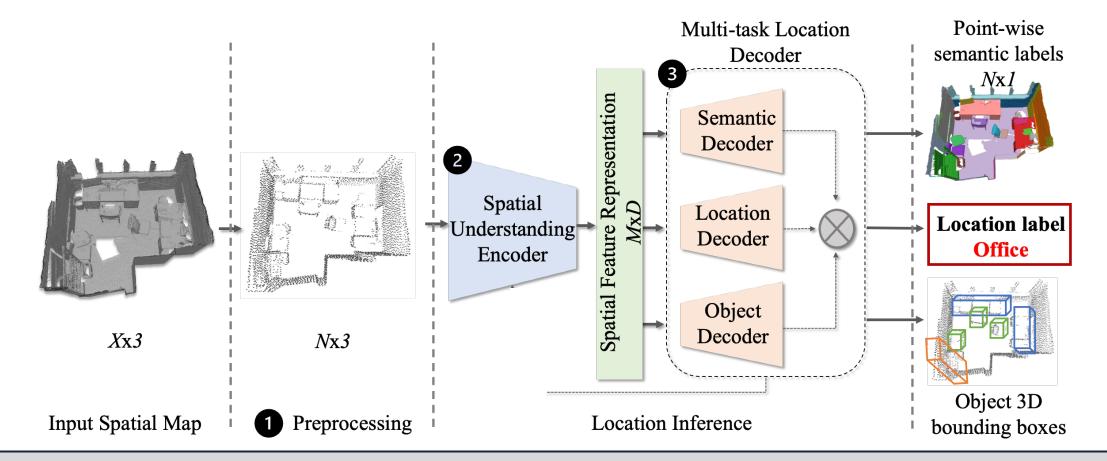


1 Geometric and semantic properties

- 2 Non-uniform point density
- 3 Varying app usage

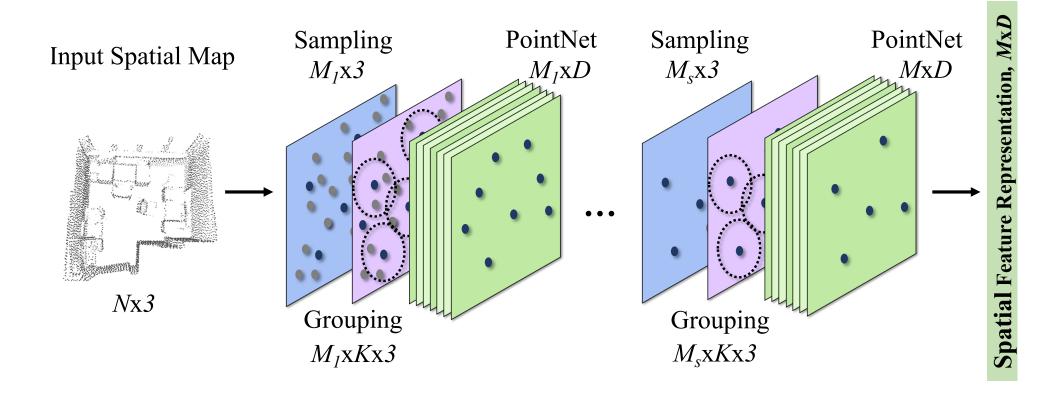
LocIn's overview

 We propose a novel location inference attack with end-to-end encoder-decoder architecture with a multi-task loss function.



LocIn's spatial encoder

• We leverage a hierarchical encoder to capture the geometric and semantic properties of the user's environment.

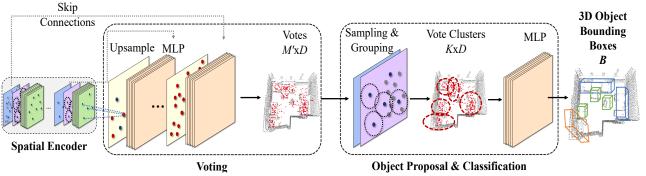


LocIn

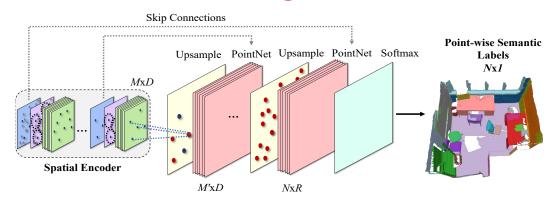
LocIn's multi-task decoder

 We propose a novel composite learning representation for location decoding by leveraging the multi-task learning paradigm.





Semantic Segmentation



$$f_s = \underset{f \in \mathcal{F}_s}{\operatorname{arg\,min}} \frac{1}{n} \sum_{i=1}^{n} \left[\alpha L_{\text{loc}} + \beta L_{\text{obj}} + \gamma L_{\text{sem}} \right]$$

Location classification loss

Object Semantic detection segmentation loss loss

LocIn's evaluation

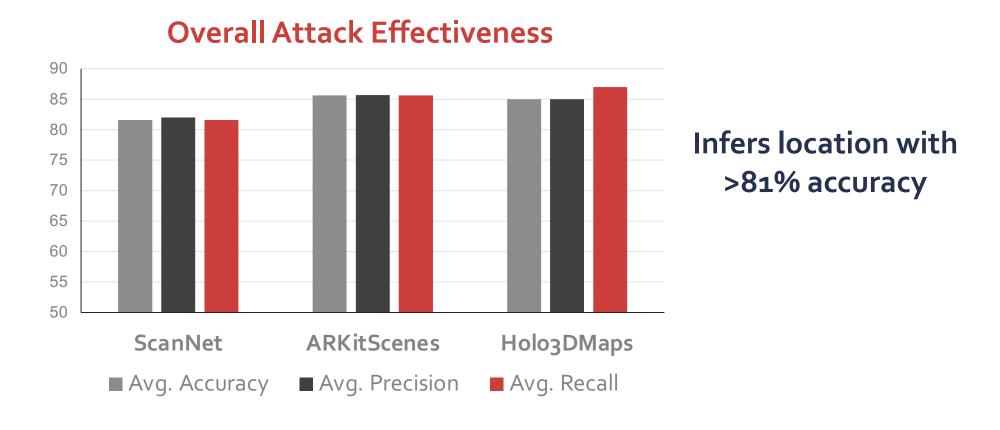
• Three datasets collected from different spatial sensors.

Dataset	MR Device	# of Location Classes	# of Object Classes	# of Spatial Maps
ScanNet	iPad Air2 with depth sensor	13	18	1513
ARKitScenes	iPad Pro with LiDAR scanner	9	17	5030
Holo3DMaps	HoloLens 2 with depth sensor	5	8	20

Evaluation 12

LocIn's effectiveness

Evaluated on spatial map datasets collected from three popular MR devices.

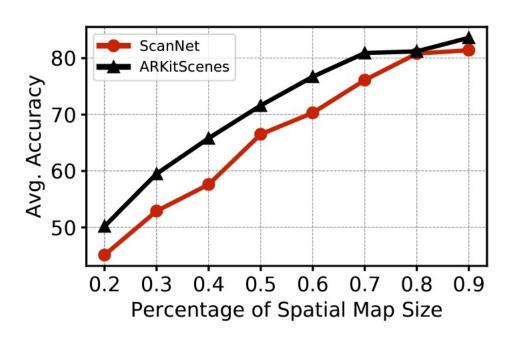


Evaluation

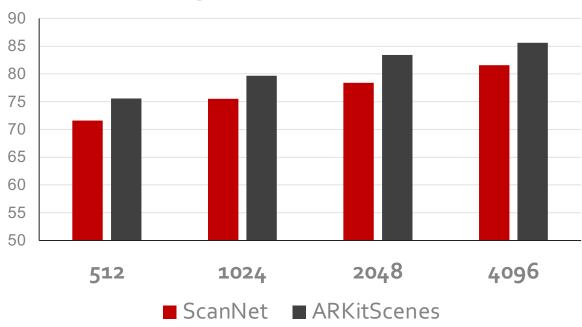
LocIn's robustness

We demonstrate LocIn's robustness against varying spatial map size and sparsity.

Varying spatial map size



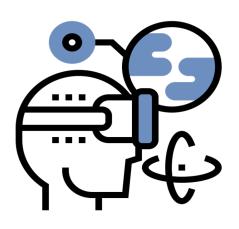
Varying spatial map sparsity



Evaluation 14

Conclusion

LocIn is a new location inference attack on mixed reality (MR) devices via 3D spatial data.



- Study attack effectiveness in complex environments.
- Investigate countermeasures with privacyutility tradeoff.
- Explore users' perception of spatial maps' security & privacy.

Conclusion

Thank you! Questions?

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