A Model based Decimeter-scale Device-free Localization System Using COTS Wi-Fi Devices

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Abstract

Commodity Wi-Fi based device-free localization has attracted a great attention in recent years. Previous related work is either fingerprint-based or model-based. In this demo, we will demonstrate a generic Fresnel Penetration Model (FPM) based real-time device-free localization system called MFDL. Using only three to four commodity Wi-Fi devices, it can localize a metal plate reflector with 6cm median error in the open space and localize a moving person with 45cm median error in an outdoor space of $36m^2$ and 50-75cm median error in indoor environments with a size ranging from $25m^2$ to $72m^2$, outperforming the state-of-theart device-free localization approaches in similar settings.

Author Keywords

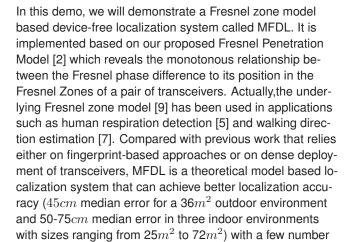
Fresnel zone model; Wi-Fi; Channel State Information (CSI); Localization.

ACM Classification Keywords

H.5.m [Information interfaces and presentation (e.g., HCI)]: Miscellaneous

Introduction

The localization of moving objects and human targets plays an important role in ubiquitous computing. Among the different technologies proposed, Wi-Fi based device-free localization has attracted a lot of attention in recent years. Previous related works can be roughly classified into fingerprint-based and model-based. The fingerprint-based approaches [8] collect CSI measurement [1](or RSSI in earlier works [4]) at each location and build a sitemap of the whole environment. The model-based approaches [3][6] establish a geometric model to quantify the relationship between CSI measuremenst and the location of target. However, most model-based works require dense deployment of devices which is not cost effective in real life.



The core idea behind MFDL can be explained as 3 steps.

of transceivers.

- 1) Knowing the monotonous relationship betweenthe Fresnel phase difference in multicarriers and the number of resided layer in Fresnel Zones as shown in Fig.2, we can map subject's location in one dimension in Fresnel Zones with one pair of transceivers.
- 2) By using two pairs of Wi-Fi transceivers, we can locate a

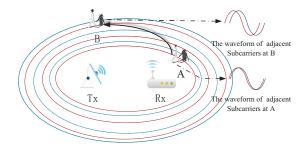


Figure 2: Conceptual illustration of the MFDL

moving subject by finding the intersection area of two Fresnel Zones as shown in Fig.3

Demonstration System

To achieve 2-D localization, we deploy two pairs of transceivers to cover the sensing area of $36m^2$ as shown in Fig.1, with one Wi-Fi transmitter (Tx) and two receivers (Rx).

As shown in Fig.4, the interface of our system consists of three components: the real-time video captured by a camera as ground truth, the calculated localization output shown in a x-y coordinate system, and the cooresponding coordinates of the localization result.

The details of our demonstration process are explained as follows:

1) During the demonstration, a subject is requested to walk in the sensing area. After pre-processing of collected CSI data, MFDL will estimate the Fresnel phase difference across multiple subcarriers for each pair of transceivers, and then compute the precise layer number in the corresponding Fresnel Zones based on the Fresnel phase difference.



Figure 1: Environment setup

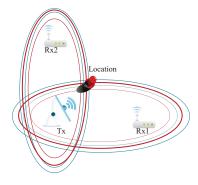


Figure 3: Locate the subject by finding the intersection area between two Fresnel Zones

- 2) Knowing the precise layer number of Fresnel Zones for each pair of transceivers, MFDL can calculate the intersection point of two Fresnel Zones to obtain the subject's location in the X-Y coordinate system.
- 3) Finally, the coordinates of the subject will be displayed with red dots in the user interface as shown in Fig.4.

Demonstration Setup

The demostration setup consists of one transmitter(Tx) and two receivers(Rx). Both receivers are mini-PCs equipped with an Intel Wi-Fi link 5300 NIC to collect Channel State Information(CSI) [1] as shown in Fig.1.

All the Wi-Fi transceivers are mounted on individual tripod with three antennas running on the 5.24GHz channel in an 802.11n Wi-Fi network. The two pairs of transceivers are placed perpendicular to each other and the length between each pair of transceivers (LOS) is set to 6 m to cover a $6m^*6m$ sensing area.

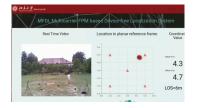


Figure 4: System Interface

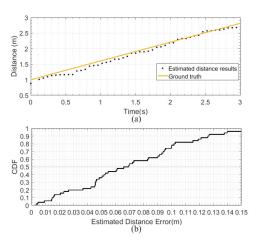


Figure 5: Localization errors with a perfect reflector in an open outspace

When a subject walks in the sensing area, a back-end server will collect the CSI datastream from the two Wi-Fi receivers and generate the localization result in real time.

Evaluation with a Perfect Reflector

We use a steel plate which is a perfect reflector for radio signals as the target object. The two transceivers are placed 4m apart. We move the steel plate along the perpendicular bisector of the transceivers from 1m to 4m in an outdoor open space. Fig.5 shows the experimental results. Fig.5.(a) compares the estimated distances based on our proposed Fresnel Penetration Model(FPM) with the ground truth, and the CDF of the estimated distance errors is depicted in Fig.5(b). As we can see the median localization error is as low as 6cm, demonstrating the great potential of FPM for fine-grained moving object localization.

Conclusion

In this demo, we have designed and implemented a real-time device-free localization system based on our proposed FPM model. Using only few Wi-Fi transceivers with simple offline calibration, our system can achieve decimeter-scale localization accuracy for a moving person in both indoor and outdoor environments, outperforming the state-of-the-art device-free localization approaches.

The demo video for the device free localization system MFDL can be found at http://wanghao13.top/wordpress/index.php/2017/07/21/mfdl/

Acknowledgements

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REFERENCES

- Daniel Halperin, Wenjun Hu, Anmol Sheth, and David Wetherall. 2011. Tool release: Gathering 802.11 n traces with channel state information. ACM SIGCOMM Computer Communication Review 41, 1 (2011), 53–53.
- Wang Hao, Zhang Daqing, Niu Kai, Lv Qin, Liu Yuanhuai, Wu Dan, Gao Ruiyang, and Xie Bing. 2017. MFDL: A Multicarrier Fresnel Penetration Model based Device-Free Localization System leveraging Commodity Wi-Fi Cards. arXiv preprint arXiv:1707.07514 (2017).
- 3. Xiang Li, Shengjie Li, Daqing Zhang, Jie Xiong, Yasha Wang, and Hong Mei. 2016. Dynamic-music: accurate device-free indoor localization. In *Proceedings of the*

- 2016 ACM International Joint Conference on Pervasive and Ubiquitous Computing. ACM, 196–207.
- Moustafa Seifeldin, Ahmed Saeed, Ahmed E Kosba, Amr El-Keyi, and Moustafa Youssef. 2013. Nuzzer: A Large-Scale Device-Free Passive Localization System for Wireless Environments. *IEEE Transactions on Mobile Computing* 12, 7 (2013), 1321–1334.
- Hao Wang, Daqing Zhang, Junyi Ma, Yasha Wang, Yuxiang Wang, Dan Wu, Tao Gu, and Bing Xie. 2016b. Human respiration detection with commodity wifi devices: do user location and body orientation matter?. In ACM International Joint Conference on Pervasive and Ubiquitous Computing. 25–36.
- Ju Wang, Hongbo Jiang, Jie Xiong, Kyle Jamieson, Xiaojiang Chen, Dingyi Fang, and Binbin Xie. 2016a. LiFS: low human-effort, device-free localization with fine-grained subcarrier information. In *International Conference on Mobile Computing and NETWORKING*. 243–256.
- Dan Wu, Daqing Zhang, Chenren Xu, Yasha Wang, and Hao Wang. 2016. WiDir: walking direction estimation using wireless signals. In *Proceedings of the* 2016 ACM International Joint Conference on Pervasive and Ubiquitous Computing. ACM, 351–362.
- Jiang Xiao, Kaishun Wu, Youwen Yi, Lu Wang, and Lionel M. Ni. 2013. Pilot: Passive Device-Free Indoor Localization Using Channel State Information. (2013), 236–245.
- Daqing Zhang, Hao Wang, and Dan Wu. 2017. Toward Centimeter-Scale Human Activity Sensing with Wi-Fi Signals. *Computer* 50, 1 (2017), 48–57.