REITS: Reflective Surface for Intelligent Transportation Systems

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Abstract
Autonomous vehicles are predicted to dominate the transportation industry in the foreseeable future. Safety is one of the major challenges to the early deployment of self-driving systems. To ensure safety, self-driving vehicles must sense and detect humans, other vehicles, and road infrastructure accurately, robustly, and timely. However, existing sensing techniques used by self-driving vehicles may not be absolutely reliable. In this paper, we design REITS, a system to improve the reliability of RF-based sensing modules for autonomous vehicles. We conduct theoretical analysis on possible failures of existing RF-based sensing systems. Based on the analysis, REITS adopts a multi-antenna design, which enables constructive blind beamforming to return an enhanced radar signal in the incident direction. REITS can also let the existing radar system sense identification information by switching between constructive beamforming and destructive beamforming state. Preliminary results show that REITS improves the detection distance of a self-driving car radar by a factor of 3.63.

CCS Concepts
• Networks → Mobile networks; Mobile and wireless security; • Hardware → Hardware reliability;

Keywords
Self-driving car, Robustness, Surface

ACM Reference Format:
Zhuqi Li∗, Can Wu∗, Sigurd Wagner, James C. Sturm, Naveen Verma, and Kyle Jamieson. 2021. REITS: Reflective Surface for Intelligent Transportation Systems. In The 22nd International Workshop on Mobile Computing Systems and Applications (HotMobile 2021), February 24–26, 2021, Virtual, United Kingdom. ACM, New York, NY, USA, 7 pages. https://doi.org/10.1145/3446382.3448650

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HotMobile 2021, February 24–26, 2021, Virtual, United Kingdom
© 2021 Association for Computing Machinery.
ACM ISBN 978-1-4503-8323-3/21/02...$15.00
https://doi.org/10.1145/3446382.3448650

1 Introduction
Autonomous vehicles are becoming ever more attractive due to their potential to enable more efficient transportation by reducing traffic congestion, fuel consumption, and air pollution. To realize widespread adoption, autonomous vehicles must meet extremely high safety standards. For example, an autonomous driving system that is comparable to human drivers should achieve a failure rate of less than 1.09 fatalities per 100 million miles [4].

In order to meet the high safety standard, a lot of previous works have improved the autonomous control systems with deep learning based algorithms [5]. However, the reliability of those algorithms heavily depends on the signal input from sensors. A safe autonomous vehicle must reliably sense humans, other vehicles, and proximate road infrastructure accurately and robustly. Therefore, autonomous vehicles are typically equipped with two categories of sensors. One category is light-based sensors, i.e., cameras or Lidar. However, light-based sensors suffer from excessive absorption or scattering of the probing signal during inclement weather. Specifically, light-based sensors may fail when light is blocked or scattered by fog, rain, or snow. The other category is RF-based sensors, i.e., radar. RF-based sensors are more reliable in extreme weather, but they also may fail to detect the object of interest when the surface of the target is not perpendicular to the incident wave. In such a case, most of the probing signal is reflected to a direction different from the incident wave, making the radar system receive a subthreshold return signal. This causes irrelevant reflections from the environment to dominate the return signal and significantly degrade its SNR. Since both types of sensors are not absolutely reliable, the self-driving system, therefore, may fail to sense the surrounding environment, which often leads to deadly consequences [2]. The signal reflection
pattern only depends on the shape and orientation of the object, which means optimization at the radar is not going to solve the problem. The natural solution is to change the radar reflection pattern of the target object’s surface.

The main challenge to achieve this goal is the dilemma between signal strength requirement and the low power constraint. On the one hand, the solution must deliver significant energy to the direction of radar receiver so that radar system can get signal with significant high SNR. On the other hand, the solution must be low power since it has to attach to the surface of any object, which excludes the design choices of using active electronic elements. Another challenge for self-driving car radar is the limitation on its sensing capability. Although self-driving radar can sense the distance, angle, and velocity of the target object, it typically cannot get the identification information (e.g. whether a traffic sign is a stop sign or speed limit sign). The limitation on radar sensing capability further reduces the robustness of self-driving system in extreme weather when vision based sensing does not work.

To solve those challenges, we propose REITS, a system designed for robust detection of objects by RF-based sensing. REITS augments existing radar systems for self-driving vehicles, and includes two design components: (1) a surface based on a Van Atta type array reflector [15] that beamforms the return signal from the object by constructive interference, and (2) programmable RF switches that alternate between constructive and destructive beamforming. When set to constructive interference, surface beamforms the signal back to its incident direction so that the self-driving radar gets high SNR. When set to destructive interference, signal absorbed by different parts of the surface cancel out to each other, leaving no reflection to the direction of incoming signal. By switching between constructive beamforming state and destructive beamforming state, REITS can encode object identification information so that the self-driving car radar can get more detail information about the target (e.g. whether the object is a vehicle or pedestrian). Wide-spread deployment of REITS in self-driving vehicle applications is anticipated from its reduced complexity and cost, because only passive antennas and switches are needed in the design, in contrast to conventional radar reflectors based on sophisticated active electronics [3, 17].

By attaching REITS to the surface of vehicles, road infrastructures, and clothes, commercial self-driving systems will be able to detect objects of interest in a more reliable manner. To achieve such a design, we make the following key contributions in this paper:

- We study the root cause of RF-based sensor failure for self-driving vehicles and propose an RF circuit design based on a Van Atta type array reflector, which automatically beamforms the signal back to the radar receiver of self-driving vehicles.
- We propose a method to transmit object identification information from the surface to self-driving car radar by switching between constructive beamforming state and destructive beamforming state.
- We verify the design of REITS with SIMULIA CST, a time-domain 3D full-wave electromagnetic solver based on finite integration and conduct a preliminary evaluation of the performance of REITS on 24-GHz self-driving radar band. The results show that a four-antenna REITS improves the signal strength by 11.2 dB and correspondingly extends the radar detection range by 3.63×.

2 Failure Analysis for Self-driving Radar

Self-driving radar uses Frequency-Modulated Continuous-Wave Radar (FMCW) technique to detect objects of interest. As shown in Figure 1, the radar sends a chirp signal that swipes a wide frequency band as the probing signal. After sending the chirp signal, the radar captures the signal reflected by the objects of interest. If the reflected signal strength exceeds a threshold, the self-driving radar can calculate the distance, direction, and speed of the target objects by applying radar detection algorithm. Since the radar signal receiver and transmitter are co-located at the same vehicle, the radar system can only capture the signal that is reflected back in the incident direction. As shown in Figure 1, the signal might also be reflected into a different direction, which cannot be received by the radar receiver. In order to have a reliable sensing capability, the signal captured by the radar receiver must always exceed the detection threshold.

The signal reflection follows Snell’s Law, as shown in Figure 2. If the object is made of a dielectric material, part of the incoming RF signal penetrates and the other part bounces off the surface and heads in a different direction, symmetric by the surface normal to the incoming angle, i.e., $\theta' = \theta$. When the object is made of metal, nearly the entire RF signal ($E' \approx E$) bounces off like a ball hitting a wall. In order to have the energy reflected back to the original direction, the incoming angle of the signal has to be zero ($\theta = 0^\circ$). But in most cases, the surface of the target is not perpendicular to the incoming signal. In such cases, only a small portion of the RF signal returns along the incoming path, resulting in a small signal at the receiver. The signal strength returning to the radar receiver becomes weak, which significantly degrades the detectable distance of radar systems and leads to failure of radar detection.

Based on the analysis, the returning signal strength of the self-driving radar depends on the area of the surface that is perpendicular to the incoming wave. Such impact to the robustness of radar signal detection is typically modeled as Radar Cross Sector (RCS). An object may not always have a large RCS in the view of self-driving car radar. In a real road scenario, the RCS of a target depends on its position, shape, and orientation. Therefore, the RCS of the object changes constantly along with dynamic road environment, leading to uncertainty in radar detection.

REITS is designed to eliminate the uncertainty of RF-based sensing for self-driving vehicles. It makes the target object have consistently good RCS regardless of its position, shape, and material so that the radar receiver can always capture strong reflection from the object of interest for sensing and detection.
An obvious approach is to collect the energy, manipulate the signal, and send it back. However, a single antenna does not collect sufficient power to amplify the RCS for two reasons: (1) the millimeter-wave signal used by self-driving car radars experiences high propagation loss; (2) antennas operated at millimeter-wave band are small in size, typically \(1.97-6.25 \text{ mm} \) (half wavelength for \(24 - 76 \text{ GHz radio} \)). Such small antennas do not receive enough power for enabling active operations such as signal amplification.

To raise the intercepted power, the target can be equipped with an array of antennas, where the array beamforms the signal back into the incident direction. However, implementing beamforming by traditional methods requires either multiple radio chains [16] or a phased array [8], which typically exceeds the energy budget of a low-power device. REITS takes a different approach by implementing a blind beamforming design to enable an array of antennas beamforming the radar signal back to its incident direction. We first illustrate the idea of blind beamforming with a one-dimensional array (§3.1.1) and then scale up the design to a two-dimensional surface (§3.1.2). We also introduce programmability to the surface, for signal modulation (§3.1.3).

### 3.1.1 Blind beamforming array

REITS adopts a blind beamformer that makes the return signals constructively interfere in the incident direction. Figure 3 is an example of a blind beamformer design using a linear four-antenna array. \(a_1, a_2, b_1, b_2\) are antennas, and \(s_1, s_2\) are transmission lines. The gap between each pair of adjacent antennas is \(d\). Suppose the distance between signal source and \(a_1\) is \(L_{a_1}\). In practice, the incident wave can be approximated as a plane wave since \(L_{a_1} \gg d\). The distances between the signal source and \(a_2, b_2, b_3\) can be expressed as \(L_{a_1} + dsin(\theta), L_{a_1} + 2dsin(\theta), L_{a_1} + 3dsin(\theta)\) respectively. The total length of signal propagation path that goes through antenna \(a_1\) and antenna \(b_1\) is

\[
2L_{a_1} + 3dsin(\theta) + l_{s_1},
\]

where \(l_{s_1}\) is the electrical length of the transmission line \(s_1\). Similarly, the length of the path that goes through antenna \(a_2\) and antenna \(b_2\) is

\[
2L_{a_1} + 3dsin(\theta) + l_{s_2}.
\]

We can control the transmission line length \(l_{s_1}\) and \(l_{s_2}\) in the board design so that they differ by a wavelength. With such a design, the signals from the four antennas constructively interfere in the incident direction.

The characteristic of the blind beamformer can be generalized as
long as the configuration of the antenna array satisfies the following two requirements:

- All pairs of antennas are centrosymmetric to the same central point. This requirement ensures that the total distance between the signal source and any pair of antennas \((a_i, b_j)\), \(L_{a_i} + L_{b_j}\), equals twice the distance between the signal source and the central point. Therefore, this part of path length is independent of the incident angle, making the beamformer work for any incident direction.
- The length difference of any two transmission lines \(s_i\) and \(s_j\) satisfies \(|l_{a_i} - l_{a_j}| = k\lambda\), where \(k\) is an integer and \(\lambda\) is the wavelength. This requirement ensures that any two propagation paths have identical phase delays. Therefore, the signals that travel through all propagation paths constructively interfere with each other at the radar receiver.

### 3.1.2 Scale to a two-dimensional surface

We proceed to scale up the blind beamforming design from a one-dimensional array to a two-dimensional array. To harvest sufficient energy for radar detection, the deployment of antennas on the surface should be dense. Figure 4 shows the design of the two-dimensional surface for REITS. All antennas are placed on concentric circles. Any pair of antennas is connected by a 180-degree arc transmission line placed on one of the concentric circles. The radius difference of any two circles is \(\frac{\lambda}{2}\), which ensures that the lengths of the transmission lines differ by \(k\lambda\). In addition, the antennas are spaced apart by at least \(\lambda/2\) to ensure minimal crosstalk.

### 3.1.3 Enable programmability

REITS can be made programmable for controlling the reflected signal strength. As shown in Figure 5, for this purpose, REITS is provided with RF switches on the transmission lines that are labeled with even indexes. Switching-in a second propagation path increases the length of the transmission line by half a wavelength. Therefore, the phase of the signal increases by \(\pi\). This mechanism enables destructive interference between different antenna pairs. By changing the number of paths that are switched into the destructive phase, the surface controls the total signal strength that is reflected into the incident direction from the target.

### 3.2 Transmitting object identification information

Besides from the location and velocity, self-driving system also requires object identification information to operate safely. For example, it needs to know whether a traffic sign is a stop sign or what is the type of traffic sign as well as the content of the traffic sign to avoid accidents. Self-driving vehicles might send out chirp signal at the same time, which might cause interference in the receiving signal among different self-driving car radars. REITS is inherently resistant to the interference caused by concurrent detection. Since REITS beamforms the signal back to its incident direction, self-driving car radars at different directions can safely receive the reflection of its own signal.

### 3.3 Practical Usage

We discuss the system level design and the use case for REITS in a practical deployment.

#### 3.3.1 Operation flow

In the practical deployment, REITS can use an ultra-low power FPGA device [1] to control the RF switch. Once the surface is attached to a specific object, the FPGA needs to be re-programmed at the same time so that it can transmit the corresponding object identification information. During the operation, the FPGA outputs the programmed control signal to make the REITS switch between the constructive and destructive states. On the self-driving car side, the radar transmits the chirp signal and measures the signal reflected back to the radar receiver. The self-driving radar reconstructs the detection map for every chirp it receives. By comparing the detection map across chirps, the radar system can decode the modulated on-off-keying reflection and get the object identification information.

#### 3.3.2 Concurrent detection

In the real deployment, multiple vehicles might send out chirp signal at the same time, which might cause interference in the receiving signal among different self-driving car radars. REITS is inherently resistant to the interference caused by concurrent detection. Since REITS beamforms the signal back to its incident direction, self-driving car radars at different directions can safely receive the reflection of its own signal.

#### 3.3.3 Use cases

REITS can be to enhance the radar detection robustness for self-driving systems. We can envision three practical use cases for REITS in the real world.

- **Augment existing radar detection.** REITS can enlarge the RCS of an object so that the self-driving car can detect the object reliably. Just like fluorescent clothing can increase the visibility of human being, in the real world, REITS can be worn by pedestrians or cyclists to reduce the change of being hit by self-driving vehicles. It can also be attached to the surface of the vehicles so that they can be detected easily.
- **Detect traffic lines.** In addition, REITS can also be attached to the traffic lines so that the self-driving car can detect the traffic line so that they can keep the right lane even during heavy rains when traffic lines become hard to see.
- **Detect traffic sign.** REITS can also help self-driving radar read the detail content of traffic sign. Specifically, REITS can transmit the type of traffic sign as well as the content of the traffic sign to the self-driving car radar (e.g. whether it is a stop sign or what is the speed limit).

### 4 Preliminary results

We evaluate the design of REITS with SIMULIA CST, a 3D electromagnetic simulator. In the simulation setup, 24 GHz plane wave excitation and open boundary condition are used to emulate a practical scenario, where the REITS device reflects the signal from a radar in far field. The REITS consists of a 0.338 mm-thick Rogers R4350B substrate, a \(\lambda/2\)-spaced linear array of metal patches (4.1 mm \(\times\) 3.12 mm) as antennas, and a ground plane. The patch antennas are designed with 80% radiation efficiency and 50Ω input impedance (realized by a 1.15 mm-long inset feed). Each centrosymmetric antenna pair is connected by a 0.67 mm-wide transmission
line with a characteristic impedance of 50Ω, for impedance match with the antennas. Controllable phase delay is achieved by assigning proper lengths to the transmission lines, for switching between the constructive and destructive settings.

We characterize the performance gain of REITS against a blank metal baseline on 24-GHz self-driving radar frequency band. In addition, we examine the controllability of REITS by comparing the signal strengths under constructive and destructive settings of the blind beamformer.

Figure 7 shows simulated electric-field patterns for REITS. The three figures correspond to the cases where the incoming EM wave ($\theta = 30^\circ$) is reflected by (1) a blank metallic rectangle; (2) a linear four-element beamformer in the constructive setting; (3) a linear four-element beamformer in the destructive setting. These devices share the same dimension (2.5 cm × 0.9 cm). As illustrated in Figure 7(a), when hitting a metal object, the EM wave is reflected into the angle identical to that of the incoming wave ($\theta' = \theta$) but symmetric by the surface normal. When REITS is set to the constructive state, we can see a clear reflection peak at the incident direction, as shown in Figure 7(b). When REITS is set to the destructive state, we can see a null at the incident direction, as shown in Figure 7(c). Overall, when striking REITS approximately 82% of the EM wave power is absorbed by the four antennas and then re-radiated, forming either a constructive or a destructive beam at the incoming angle, depending on the switch settings.

Figure 8 shows the amplitude of the electric field in the incoming direction ($\theta = 30^\circ$) vs. distance, when the incident amplitude is 1V/m.

We further investigate the scalability of our design, as shown in Figure 9. The grey dashed line is the upper bound, where the signal strength increases linearly with the number of antennas. We can see that REITS can achieve close to linear scale-up as the number of antenna increases. In contrast, the signal reflection in the incident direction by a metal surface remains low.

5 Related Work

REITS is related to two lines of research. One is the research effort to improve the accuracy and robustness of self-driving sensing. Another is the endeavor for long-range passive communication.

5.1 Sensing for self-driving vehicles

Self-driving vehicles depend on accurate and reliable sensing capability to detect obstacles on the road. Existing works have used vision-based methods [6, 9] to extract the information from images captured by the camera. Vision-based methods can provide rich context information for self-driving vehicles. But it might fail due to adverse weather [19] or cyber-physical attack [11]. To deal with the failure of vision-based sensing, self-driving vehicles are also equipped with radars as a backup mechanism since it is more reliable in inclement weather. Many works also investigate the methods to improve the detection accuracy and robustness of self-driving
vehicle radars [13, 14] with signal processing techniques.

Different from existing work, which focuses on optimization from the view of self-driving vehicles, REITS uses a different approach to improve the safety of self-driving vehicles from the infrastructure perspective. In addition, REITS focuses on improving the signal itself. It is complementary to existing work, which applies various signal processing techniques to achieve accurate sensing.

5.2 Passive communication design

Research efforts also focus on the design and implementation of reliable passive tags for communication. One line of research is to extend the communication range of passive RFID tags. PushID [20] uses a distributed beamforming approach to extend the range of RFID communication. mmTag [10] uses a blind beamforming approach to increase the throughput of millimeter tag readers. The design of REITS is also a type of passive hardware. Different from those, REITS focuses on extending the range of sensing.

Another line of research pushes the range of passive tags to long-distance [12, 18] by integrating backscatter techniques with LoRa. These techniques enable backscatter communication even if the signal strength is below the noise floor. But the tag should be deployed close to the excitation signal so that the backscattered signal can reach receivers at distance. REITS targets on self-driving radar applications, where the distance from the tag to the signal source and the receiver are the same. At the same time, the latency of long-range backscatter is too high for the self-driving application, since long-range backscatter techniques have to aggregate dozens or hundreds of readings to boost the signal to noise ratio.

6 Discussion

In this paper, we have investigated the design of a programmable surface that enables robust sensing by self-driving vehicle radar. Our preliminary results show that the surface can extend the detection range of self-driving vehicles significantly. Nevertheless, in order to build a practical system based on the design the following topics require further study:

Frequency trade-off Self-driving radar is licensed on two frequency bands by FCC: 24-GHz band and 76-GHz band. Currently, the design of REITS is based on the 24-GHz frequency band. It is also meaningful to scale the design to 76-GHz frequency. At the higher frequency, the dimension of antenna is smaller. Therefore it can capture less energy with a single antenna. Besides, the high-frequency signal typically has a higher loss in the transmission line. In order to achieve the same performance gain, the surface at 76-GHz needs to have more antennas compared with the surface at 24-GHz. Therefore, we need a design that can deploy more antennas on a surface.

Multiple reflective surfaces The current design of REITS focuses on improving the robustness of radar detection for a single REITS surface. But in the real world, there might be multiple REITS surfaces in the view of a self-driving vehicle. For two reflective surfaces that have enough separation in space, the radar system can detection both of than by looking at the difference of angle of arrival and time of flight. However, it is challenging to identify two object when they are too close to each other, when the object identification signal will interfere with each other. We need to continue to investigate how to distinguish those multiple surfaces even if they are located close to each other.

Modulate Identification Information REITS is able to transmit object identification information with on-off-keying. However, the modulation of the information also affects detection accuracy of the radar system. For example, if the frequency of constructive beamforming state is too sparse, the radar will receive a few chirps, which reduces the accuracy to compute the Doppler effect of the target object.

Doppler effect. The Doppler effect is introduced by the speed difference between the target and the self-driving vehicle. It might corrupt the object identification information. Since the radar chirp time is typical short (e.g. 0.5 millisecond), REITS can switch between constructive beamforming state and destructive beamforming state at a frequency where the Doppler effect is neglectable. For example, we can set the switch interval to one millisecond. The maximal acceleration of a typical car is around 3-6 m²/s. The speed difference between two states is at most 0.003-0.006 m/s.

7 Conclusion

In this paper, we propose and verify the design of REITS, a system to improve the robustness of detection of objects for autonomous vehicles by maximizing the RCS of the objects. REITS adopts a multi-antenna design, which enables constructive blind beamforming to return an enhanced radar signal in the incident direction. REITS can also let the existing radar system to sense identification information by switching between constructive beamforming state and destructive beamforming state. Results of a preliminary evaluation show that REITS enhances the return signal strength by 11.2 dB and extends the detection distance by 3.63x.

Acknowledgments

We thank the anonymous reviewers and our shepherd, Haitham Hasanieh for their insightful comments. This work is supported by the National Science Foundation under grant CNS-1617161, Semiconductor Research Corporation, Center for Brain-inspired Computing, and Princeton Program in Plasma Science and Technology (PPST). Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the our funding agents.

References

[1] IGLOO nano FPGAs. Webpage.
[2] Tesla Deaths as of 10/24/2020. Webpage.
[3] DADASH, M. S., HASCH, J., AND VOIINGESCU, S. P. A 77-ghz active millimeter-wave reflector for fmcw radar. In IEEE RFIC (2017).
[4] KALRA, N., AND PADDOCK, S. M. Driving to safety: How many miles of driving would it take to demonstrate autonomous vehicle reliability? Transportation Research Part A: Policy and Practice (2016).
[5] KUDERER, M., GULATI, S., AND BURGARD, W. Learning driving styles for autonomous vehicles from demonstration. In IEEE ICRA (2015).
[6] LIANG, X., WANG, T., YANG, L., AND XING, E. Cirl: Controllable imitative reinforcement learning for vision-based self-driving. In ECCV (2018).

[7] LORRAIN, P., CORSON, D. R., AND LORRAIN, F. Fundamentals of electromagnetic phenomena. 2001.

[8] MAILLOUX, R. J. Phased array theory and technology. Proceedings of the IEEE (1982).

[9] MAQUEDA, A. I., LOQUERCIO, A., GALLEGOS, G., GARCÍA, N., AND SCARAMUZZA, D. Event-based vision meets deep learning on steering prediction for self-driving cars. In CVPR (2018).

[10] MAZAHERI, M., CHEN, A., AND ABARI, O. Millimeter wave backscatter: Toward batteryless wireless networking at gigabit speeds. In HotNets (2020).

[11] NASSI, B., NASSI, D., BEN-NETANEL, R., MIRSKY, Y., DROVIN, O., AND ELOVICI, Y. Phantom of the adas: Phantom attacks on driver-assistance systems.

[12] PENG, Y., SHANGGUAN, L., HU, Y., QIAN, Y., LIN, X., CHEN, X., FANG, D., AND JAMIESON, K. Plora: A passive long-range data network from ambient lora transmissions. In Sigcomm (2018).

[13] Rohlfling, H., AND MEINECKE, M.-M. Waveform design principles for automotive radar systems. In 2001 CIE International Conference on Radar Proceedings (2001), IEEE.

[14] SCHEINER, N., KRAUS, F., WEI, F., PHAN, B., MANNAN, F., APPENRODT, N., RITTER, W., DICKMANN, J., DIETMAYER, K., SICK, B., ET AL. Seeing around street corners: Non-line-of-sight detection and tracking in-the-wild using doppler radar. In CVPR (2020).

[15] SHARP, E., AND DIAB, M. Van atta reflector array. IRE Transactions on Antennas and Propagation (1960).

[16] SHEPARD, C., YU, H., ANAND, N., LI, E., MARZETTA, T., YANG, R., AND ZHONG, L. Argos: Practical many-antenna base stations. In MobiCom (2012).

[17] STROBEL, A., CARLOWITZ, C., WOLF, R., ELLINGER, F., AND VOSSEK, M. A millimeter-wave low-power active backscatter tag for fmcw radar systems. IEEE Transactions on Microwave Theory and Techniques (2013).

[18] TALLA, V., HESSAR, M., KELLOGG, B., NAJAFI, A., SMITH, J. R., AND GOLLAKOTA, S. Lora backscatter: Enabling the vision of ubiquitous connectivity. Mobisys (2017).

[19] TUNG, F., CHEN, J., MENG, L., AND LITTLE, J. J. The raincouver scene parsing benchmark for self-driving in adverse weather and at night. IEEE Robotics and Automation Letters (2017).

[20] WANG, J., ZHANG, J., SAIHA, R., JIN, H., AND KUMAR, S. Pushing the range limits of commercial passive rfid. In NSDI (2019).