

# Inaudible Attack on Smart Speakers With Intentional Electromagnetic Interference

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**Abstract**—This article demonstrates an inaudible attack on smart speakers using electromagnetic interference (EMI). The EMI induces voltages on the order of a few millivolts on conductors, which are then converted into baseband signals by exploiting the inherent nonlinearity of microphones. The EMI signal is specially preprocessed to minimize the useless harmonics generation at the microphone output signals, which significantly improves the recognition rate as well as nullify the previous countermeasures based on the harmonics detection. The sensitive carrier frequency found by our proposed method can improve the attack distance as well. A measurement-based methodology is applied to locate the sensitive regions for noise coupling without knowing the layout of the printed circuit board (PCB), and the transfer function is also obtained to insure the main coupling location. Our experiments show that in open space, intentional EMI under 2.5 W can inject commands at distances up to 2.5 m on smart speakers.

**Index Terms**—Hardware security, inaudible attack, intentional electromagnetic interference (IEMI), smart speaker.

## I. INTRODUCTION

A SMART speaker nowadays is not a just music player. With more and more devices connected, smart speakers such as Google Home and Amazon Echo can serve as a “home assistant” that can provide control of common household tasks, such as environmental control (thermostat), lighting, door locks, security monitoring, and more.

The security of these smart speakers can have substantial effects, such as compromised home security and information leakage. Since smart speakers are having Wi-Fi or Bluetooth connections, various attacks can be performed through the apps and networks [1]–[3]. Researchers have implemented two application-level attacks, voice squatting and voice masquerading, which impersonate the smart speakers to steal and eavesdrop the conversations [4], [5]. A security researcher from MWR Info Security has demonstrated an attack on Amazon Echo speakers by placing the malware which enables the adversary to have access to control the smart speaker [6], [7]. However, these attacks cannot be executed remotely.

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On the other hand, as the system trusts the microphone readings, a physical layer attack can readily bypass conventional security algorithm providing an unchecked entry point to the system. The application layer with software running on the smart speaker systems makes critical decisions of the input data acquired by the microphone circuit. Recently, several publications have demonstrated inaudible voice commands injection on the physical layer of the smart speakers by exploiting the nonlinearity of the microphone [8]–[17].

The dolphin attack or ultrasound attack [9]–[15] has demonstrated that a voice-enabled device can respond to inaudible ultrasound voice commands. More recently, laser pointers have been demonstrated as another tool for attacking microphone-based devices [8]. Some defense methods regarding the ultrasound attack have been investigated recently [9], [15] such as the voice signal processing method proposed in [15]. However, these attack types are limited by obstacles such as windows. The ultrasound is mechanical waves which need strong power to propagate through the window [27], and the laser pointer attack requires the device insight because the microphone of the smart speaker needs to be pointed while attacking.

In contrast to other types of attacks such as ultrasound and light command, EMI-based attack can penetrate windows with relatively low loss and does not need to have the target in sight. The high-power intentional EMI can stop electrical network such as electric cars, trains, and transformers [20]–[27], and radio communicating devices such as cellphone, computer, and other electronics will be impacted as well [20], [23]; the required high power can be deduced by [22] for a long-distance attack. The IEMI can also be applied to inject information into the analog devices which operate in the order of a few millivolts [16]–[19]. This attack with circuit easily affected is known as “back-door” interfering [17]. Since the acquisition process requires much lower energy, microphone circuit with cables or copper PCB interconnects is vulnerable to interference [21] and allows the information injection [17]. The intentional EMI has been employed to attack the headset cable of smartphones [16], and the audio signal has been injected through the electromagnetic coupling on the cable of the headset because the cable can act as an antenna which can receive the electromagnetic interference (EMI). The intentional EMI has also been employed to attack the analog sensor of the microphone of cardiac electronic devices [17]. However, in their application, the attack setup needs to be placed very close to the cardiac electronic device.

This article is the first article demonstrating the IEMI attack on the smart speakers and attempted to increase the IEMI attack distance for the smart speakers and cellphones. Different from previous cellphone attack work, we targeted at the microphones of the devices not the headset cable. The attack principles including the electromagnetic (EM) coupling and microphone nonlinearity are presented in Section II. Then, the attack signal is optimized by exploring the nonlinearity performance of the digital microphone of the smart speakers to achieve longer distance in Section III. In Section IV, since the coupling efficiency to the smart speaker will differ depending on the frequencies of the attack signal, the sensitive frequency of the attack EM wave is explored based on measurement. A measurement-based methodology is applied to investigate the sensitive location without knowing the exact layout of the board. The maximum real voice attack distance is discussed, and the required electrical field intensity to attack different devices is presented. Finally, conclusions and countermeasures are proposed in Section V.

## II. ATTACK OVERVIEW

### A. Threat Model

We assumed that the adversary could obtain a smart speaker having the same model as the targeting smart speaker being widespread in the market. The investigation and attack experiments can be performed for the obtained smart speaker within a private home or a laboratory. In addition, the same smart speaker model can be attacked using the same setup developed in the laboratory. As the system of the smart speaker trusts the microphone readings from the physical circuit, the physical attack can bypass conventional security algorithm. The adversary can manipulate the microphone's reading by injecting voice command signals to the analog circuit of the microphone in the smart speaker; then, the injected voice command can pass the application layer algorithm of the system and be recognized by the smart speaker. Finally, the intended voice command from the adversary will be executed by the smart speaker. Meanwhile, the adversary can make the voice command signal inaudible to human but audible to the smart speaker. Generally, human ear can receive audio signals with frequency between 20 Hz and 20 kHz and, therefore, the microphones of the smart speakers are also designed to receive the audio signals in this frequency range. Thus, no one can realize that the target device is being attacked. This is a critical security issue for the smart speakers placed at homes.

Although some smart speakers can be set to recognize only the owner's voice, the owner's voice record or speeches can be found on the internet or somewhere; then, the recompose of the voice command word by word can be done through the software programming [37]–[39]. In addition, the owner's voice can also be constructed through deep learning [40], [41].

### B. IEMI Coupling Mechanism

The working smart speaker with an electrical circuit can be taken as a receiving antenna; EM waves can be coupled to conductors on the printed circuit board (PCB) [32], [33], as shown by the dashed line in Fig. 1. When the signal is

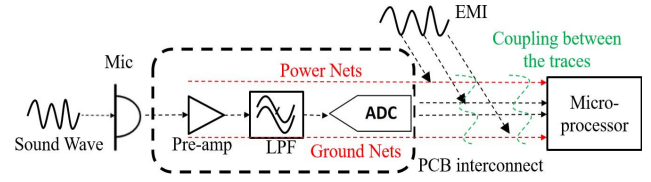


Fig. 1. Microphone circuit diagram and the anticipated coupling path.

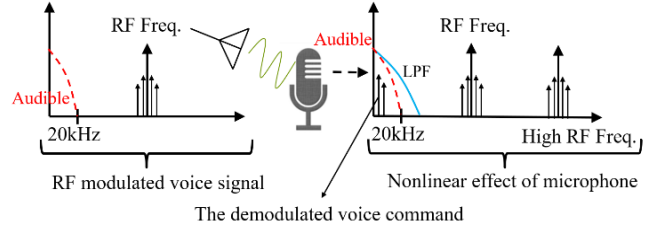


Fig. 2. Demodulation due to the inherent nonlinearity of microphones.

coupled onto the power/ground net and reached the amplifier, the induced nonlinearity can be modeled by developing the output signal equations of a simple amplifier [34], [35]. The injection path of the EMI attack is different from the previous attacks such as ultrasound commands where the commands are injected through the membrane [9]–[12]. However, the IEMI attacks are performed by injecting the signal to the electrical circuit.

All the circuits are having components that can couple the EM signal efficiently from MHz to GHz depends on the resonant frequency of the receiving circuit. Once the EM signal is coupled to the PCB, the traces on the PCB will deliver the signal to the microphone module.

### C. Nonlinearity of Microphones

A typical microphone system in a smart speaker consists of four primary blocks, as shown in Fig. 2 [9]–[12]: a microelectromechanical system (MEMS)-based microphone sensor [28]–[31], an amplifier, a low-pass filter (LPF), and an analog-to-digital converter (ADC). The acoustic waves passing through the microphone sensor induce vibrations in the membrane and are processed by the rest of the circuit. Most microphones are designed to only capture voice commands below 24 kHz. An amplifier is used because the captured voice command is too low in amplitude to be processed by the ADC. The ADC quantifies the signal levels, usually with a sampling rate of twice the maximum voice signal frequency. As a result, audio signals with frequencies greater than 24 kHz will be removed by the LPF.

As previously reported, nonlinearity is induced in the microphone circuit. The nonlinearity can be expressed by the following equation:

$$S_{out} = aS_{in} + bS_{in}^2 + \dots + dS_{in}^4 + mS_{in}^n. \quad (1)$$

In general, the coefficients of the higher order terms decrease dramatically, with the coefficients  $m \ll c \ll b$ ; hence, only the second-order coefficient needs to be considered for the nonlinearity [10], [28]. The attack signal,  $A\cos_i t$ , is multiplied with the carrier signal,  $B\cos_r t$ , to generate the

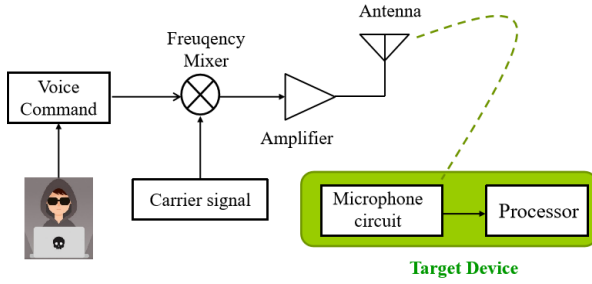


Fig. 3. Intentional EMI attack schematic.

amplitude-modulated signal

$$A \cos \omega_i t \times B \cos \omega_r t = \frac{AB}{2} [\cos(\omega_r - \omega_i)t + \cos(\omega_r + \omega_i)t] \quad (2)$$

where  $A$  and  $B$  are the amplitude of the signals,  $\omega_i = 2\pi f_i$  and  $\omega_r = 2\pi f_r$  represent the angular frequency of the attack and carrier signals, and  $f_i$  and  $f_r$  are the frequencies of the attack signal and carrier signal with relation  $f_i \ll f_r$ . Due to the second-order term of (1),  $S_{in}^2$ , the manipulated voice command will be shifted to the audible range as shown in (3). Since the carrier signal is a high-frequency signal, both low and high frequency components will present after the square function. The high frequency will be removed by the LPF in the microphone circuit, therefore, only low-frequency components presented in the following equation:

$$(A \cos \omega_i t \times B \cos \omega_r t)^2 \rightarrow \frac{A^2 B^2}{4} \cos 2\omega_i t. \quad (3)$$

We assume  $f_i$  is the voice command below 10 kHz in the audible range. After the nonlinear operation of the microphone, low-frequency audible components up to 20 kHz containing the information of the voice command are generated. Since the spectrum of the audible output is doubled compared to the voice command, the voice command signal needs to be preprocessed before it is modulated into the attack signal so that the exact voice command can be recovered after this nonlinearity of the smart speaker.

#### D. Attack Setup

Fig. 3 shows a general setup of the attack. A computer or audio signal generator is used to generate the attack voice signal that is sent to a mixer. Different mixers can be applied to mix/modulate the attack signal to the carrier signal depending on the sweeping frequency band. A frequency synthesizer, vector network analyzer (VNA), or signal generator can be employed to generate the carrier signal. A power amplifier is used for amplifying the modulated signal. Since the unidirectional antennas radiate to all directions, a directional antenna is used to radiate more power in the dedicated direction of the modulated signal. The real attack setup is shown in Fig. 4.

### III. AMPLITUDE MODULATED ATTACK SIGNAL MANIPULATION

Assuming that we modulate a 2 kHz single tone signal to the carrier signal directly without pre-processing and send

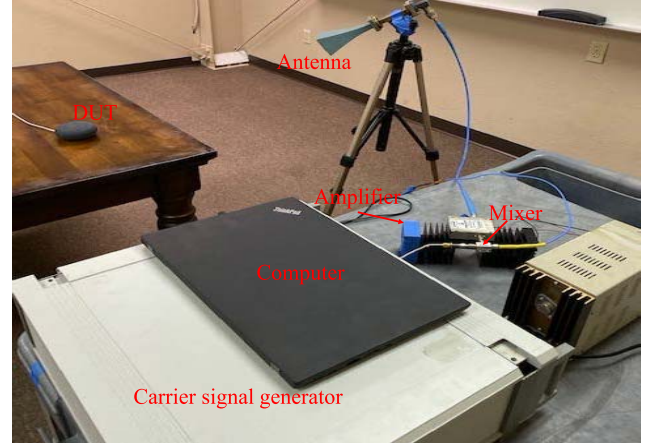


Fig. 4. Real attack setup.

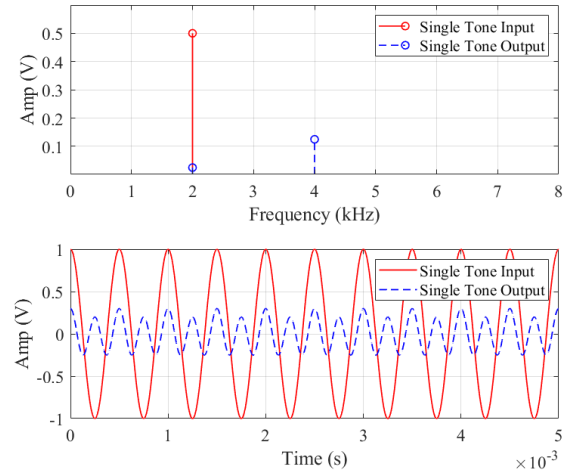


Fig. 5. Original single-tone input and its model output.

out to the microphone circuit, a square function exhibiting nonlinear behavior of the microphone circuit will be applied to the modulated signal. The resulting signal passes through the LPF in the microphone circuit, and only the low-frequency components remain. Through the mathematical derivation, the low-frequency component  $\cos(\omega_i t)$  with  $f_i = 2$  kHz and  $\cos(2\omega_i t)$  with  $2f_i = 4$  kHz are found after the LPF as shown in the equation below

$$(A \cos \omega_i t \times B \cos \omega_r t + F \cos \omega_r t)^2 \rightarrow \frac{A^2 B^2}{4} \cos 2\omega_i t + ABF \cos \omega_i t \quad (4)$$

where  $\cos(\omega_r t)$  is the feed-through component generated by the mixer due to the limited isolation of the mixer. The measurement of the modulated signal through the mixer exposed this feed through component. And this component has been applied in all the computations in the following sections. As shown in Fig. 5, the generated 4 kHz is much stronger than the 2 kHz signal. Since we want to recover the attack signal of 2 kHz after the microphone's nonlinearity, the preprocessing of the attack signal needs to be performed.

#### A. Dc Added Attack Signal Preprocessing

By adding the dc component to the attack signal, still using a 2-kHz signal as an example, the model output will change.



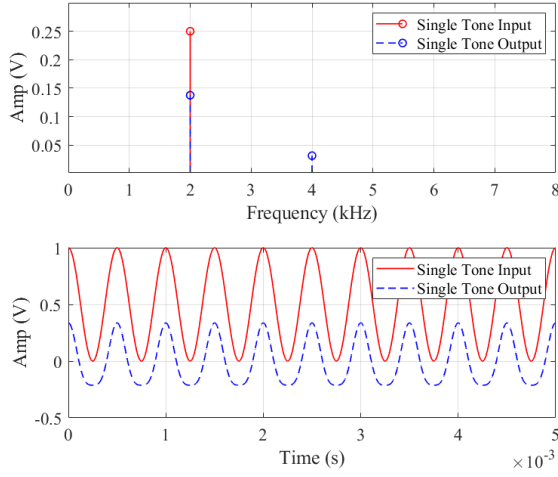


Fig. 6. Dc-added single-tone input and its model output.

As shown in (5) below, where  $C$  is the amplitude of the dc component, after the LPF, we still have both  $\cos\omega_i t$  and  $\cos 2\omega_i t$ , in other words, the 2- and 4-kHz signals, as shown in Fig. 6. But now the 2-kHz signal has a higher amplitude compared to the previous case. Notably, the time domain output waveform is deformed compared to the original signal waveform shown as the red solid curve in Fig. 5

$$\begin{aligned} & ((A \cos \omega_i t + C) \times B \cos \omega_r t + F \cos \omega_r t)^2 \\ & \rightarrow \frac{A^2 B^2}{4} \cos 2\omega_i t + (ACB^2 + ABF) \cos \omega_i t. \end{aligned} \quad (5)$$

To ensuring that the coefficient of the  $\cos 2\omega_i t$  component is much smaller than the coefficient of the  $\cos \omega_i t$  component, as shown in (5), the relation in (6) can be developed

$$\frac{A^2 B^2}{4} \ll (ACB^2 + ABF) \Rightarrow AB \ll 4CB + 4F. \quad (6)$$

$4(CB + F)/AB \gg 1$  should be the condition to minimize the  $\cos 2\omega_i t$  component.

### B. Square-Rooted Attack Signal Preprocessing

Since the nonlinearity is represented as the square term as shown in (1), a square root of the signal can be first performed. Therefore, after the square function of the signal, the original signal can be recovered. Since the computer can only output the real number of the signals, the dc value has to be added first before square root to avoid generating complex values. Continuing to preprocess the attack signal, the operation shown in the following equation can be performed:

$$\begin{aligned} & (\sqrt{(A \cos \omega_i t + C)} \times B \cos \omega_r t + F \cos \omega_r t)^2 \\ & \rightarrow \frac{A^2 B^2}{2} \cos \omega_i t + BF \sqrt{(A \cos \omega_i t + C)}. \end{aligned} \quad (7)$$

As shown in Fig. 7, by applying this operation to the attack signal, we still have the  $\cos 2\omega_i t$  signal (4 kHz), but it is much lower in amplitude and has less effect on the original signal,  $\cos \omega_i t$ . Moreover, the shape of the time domain output curve is well recovered compared with the dc added case. Therefore, the square-rooted injection signal is a better attack signal recovered in the smart speaker.

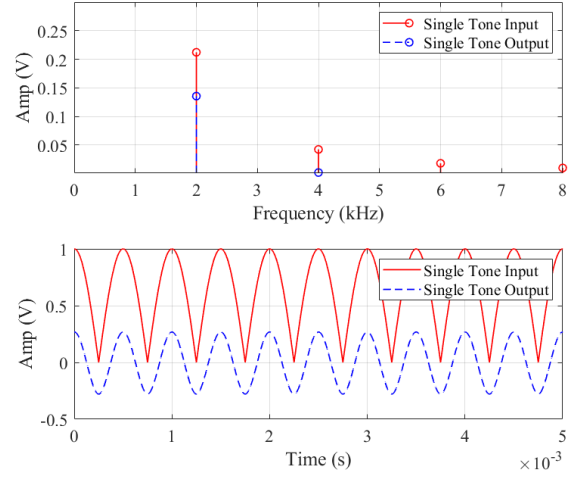


Fig. 7. Square-rooted single-tone input and its model output.

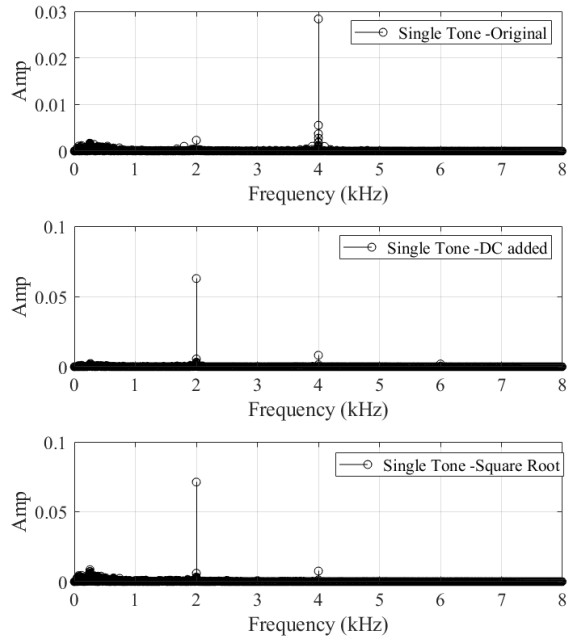


Fig. 8. Single-tone injection signal measurement.

### C. Verification of the Preprocessed Attack Signal

To verify the preprocessing procedure, three types of injections are applied to a target device by applying the setup shown in Fig. 8. All the attacks in this section are performed at the sensitive frequency of the smart speaker, which will be explained in Section IV-A. After the frequency analysis of the recorded files, the results shown in Fig. 8 are obtained. The results from Fig. 8 prove that the square-rooted attack signal has the best performance when injected into the device.

### D. Real Voice Command Attack Analysis

To be more confident on the attack signal preprocessing, some real voice command attacks were performed. One attack voice command was “What time is it?,” and the target device responded with the current time. The command was sending continuously. Fig. 9 shows that the recorded voice signal matches well with the original signal. The square-root function of the original signal was applied to form the attack signal,

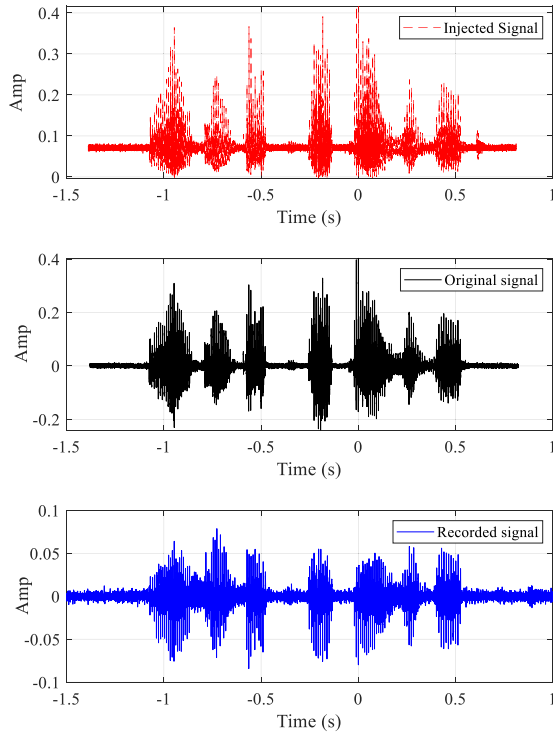


Fig. 9. Real voice command injection measurement.

TABLE I

MAXIMUM ATTACK DISTANCE BASED ON THE CURRENT SETUP WITH DIFFERENT ANTENNAS

Product	<i>Smart Speaker 1</i>	<i>Smart Speaker 2</i>	<i>Smart Speaker 3</i>	<i>Cellphone 1</i>
Maximum attack distance	2.5 m	40 cm	40 cm	20 cm
Minimum attack power density	2.94 Watts/m	39.3 Watts/m	39.3 Watts/m	157.2 Watts/m
Minimum attack electrical field intensity	150 dBuV/m	161.7 dBuV/m	161.7 dBuV/m	167.7 dBuV/m

and the resulting signal was then injected to the target device to ensure better signal recovery in the recorded file. However, without the preprocessing, the target device could not understand the voice command because the frequency of the signal changed due to the nonlinear effect.

At the maximum attack distance shown in Table I, the target devices can barely recognize the voice command. Therefore, we can analyze the efficiency of the different preprocessed attack signals with the peak-to-peak value normalized to 1. The recognition rates of the various preprocessed attack signals for different products are compared in Fig. 10, indicating that the square-rooted input has the best attack performance. The recognition rates are determined from the execution times of the target devices over ten attacks for each preprocessed attack signal.

#### IV. EXPERIMENTAL FEASIBILITY ANALYSIS

In this section, the sensitive carrier frequency is analyzed first; then, the sensitive location for EM coupling is studied based on the measurement without knowing the PCB layout.

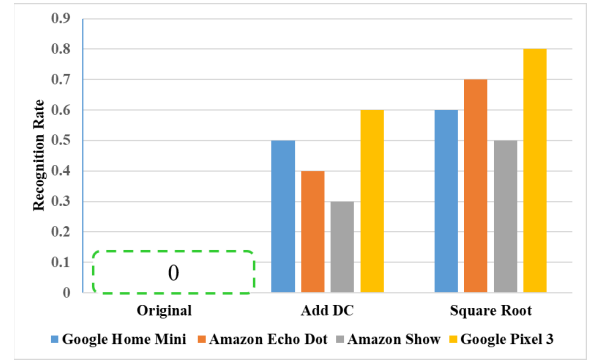


Fig. 10. Recognition rate for different preprocessed input signals and products.

The maximum attack distances are investigated for different smart speakers. And then the required electrical field densities are analyzed for different device under tests (DUTs).

#### A. Sensitive Carrier Frequency Analysis

The most sensitive frequency of the carrier signal needs to be identified to have efficient energy coupled to the smart speaker. In addition, attacking at the sensitive frequency can increase both the attack distance and the success rate. The following process can be applied to find the most sensitive frequency of the carrier signal for implementing an attack on a smart speaker.

- 1) A single-tone audible signal (2 kHz was chosen in this article; other single-tone signals within the audible frequency band can be chosen as well) modulated to the carrier signal is applied for the attack for simplicity, because a real voice command is a signal with multiple tones, which is difficult to define the amplitude because there might be some other noise in the recorded file.
- 2) Then, sweep the frequency of the carrier signal with the attack setup and send the modulated signal to the smart speaker with an activation voice command to wake up the device. Alternatively, let the device make a voice call to a phone which can record before sending the modulated signal.
- 3) The record file can be downloaded from the cloud because most smart speakers upload the voice command to the cloud automatically. Alternatively, the recorded file on the phone can be transferred to the computer for analysis.
- 4) Finally, the recorded file can be analyzed through Fast Fourier Transform (FFT) to determine whether the frequency harmonic at 2 kHz is present. By comparing the amplitude of the harmonic at 2 kHz, the sensitive frequency can be determined.

The frequency of the carrier signal was swept from 1 to 18 GHz with 1-GHz frequency step using the setup shown in Fig. 11. When the setup is fixed, the sweeping process was automated by programming the signal generator.

Fig. 12 shows the ratio of the power of the recorded 2-kHz component to the power of the attack signal at the antenna output for two different products, and the ratio is representing the transfer function from the antenna output to the record

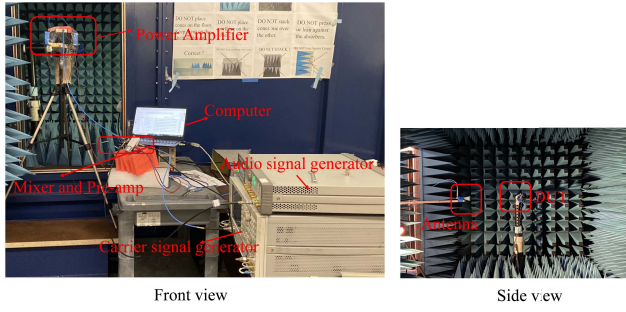


Fig. 11. Sensitive frequency measurement setup.

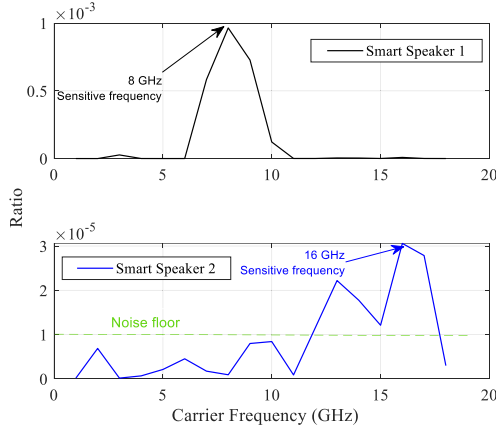


Fig. 12. Sensitive carrier signal frequency analysis for two different types of smart speakers.

file output. Four main propagation paths are included in this ratio: air propagation, coupling path, demodulation process, and record file. The same distances, 50 cm for *Smart Speaker 1*, 20 cm for *Smart Speaker 2*, are maintained for the different frequencies of the carrier signal. The sensitive frequencies of these two products are obtained at 8 and 16 GHz, respectively. From the amplitude ratios of the two products, we observe that the *Smart Speaker 1* can easily be coupled at 8 GHz. Since the environmental noise may contain the audible signals that can be recorded by the devices, this may impact the final obtained results. Thus, the experiments need to be performed in a quiet room to have reliable results. The sensitive carrier signal frequency is found at 16 GHz for the *Smart Speaker 2*. Although the ratio is very low, the attack still succeeded because the application layer of different smart speakers has different decisions on the input signal level.

### B. Measurement-Based Sensitive Location Analysis

To explore how the EM wave was coupled to the smart speakers, a measurement-based methodology is proposed. The *Smart Speaker 1* was used as the DUT to present our methodology. A near field injection technique was applied, as shown in Fig. 13. The field probe was used to inject the modulated EM signal which is different from the normal near-field scan that measures the electromagnetic field component at a scanning location. The rest of the setup is the same as in Figs. 4 or 11; only the antenna was replaced with a high-frequency field probe which was used to inject the modulated EM signal.

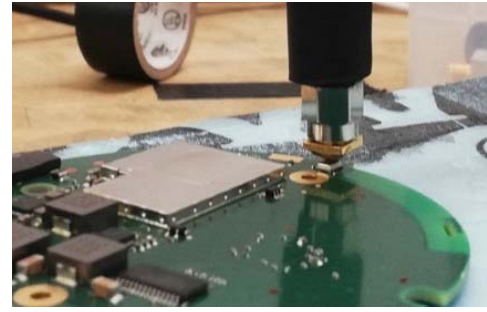


Fig. 13. Near-field injection setup.

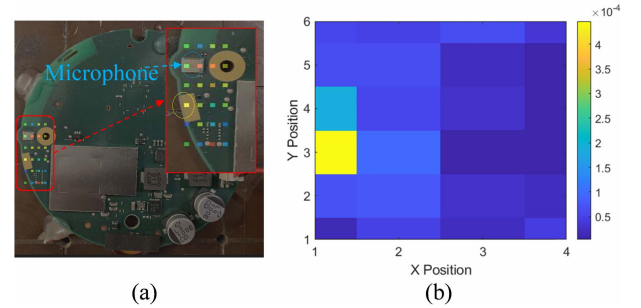


Fig. 14. Near-field injection region and results.

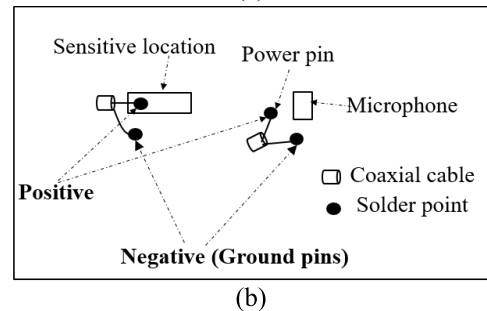
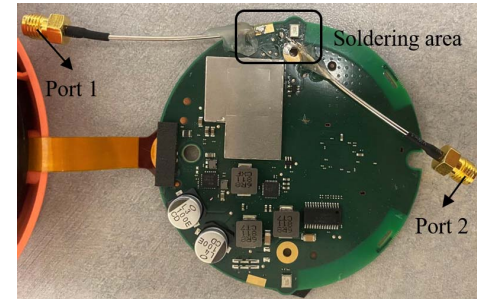


Fig. 15. (a) S-parameter measurement setup for the *Smart Speaker 1*. (b) Soldering area description.

At each position, the injection was performed and the recorded file was analyzed to monitor the strength of the 2 kHz signal. The injection area is the red-colored region shown in Fig. 14(a), where the microphone is located. Fig. 14(b) represents the 2-kHz magnitudes received in the recorded file at different locations. The results indicate that the yellow circled region is the most sensitive location.

To support that the sensitive location results in the highest noise level coupled to the microphone, the coupling path transfer function was obtained between the power pin of the microphone and the sensitive location. Fig. 15(a) shows

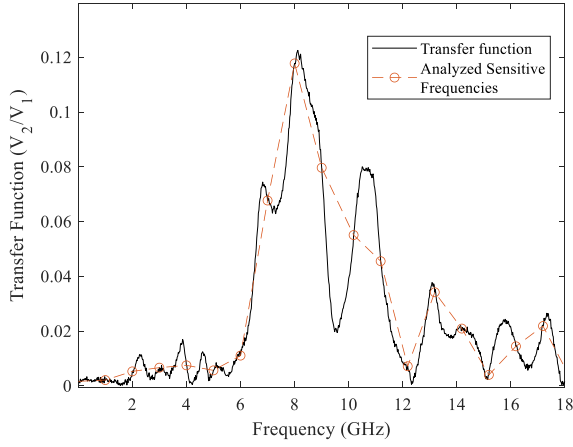
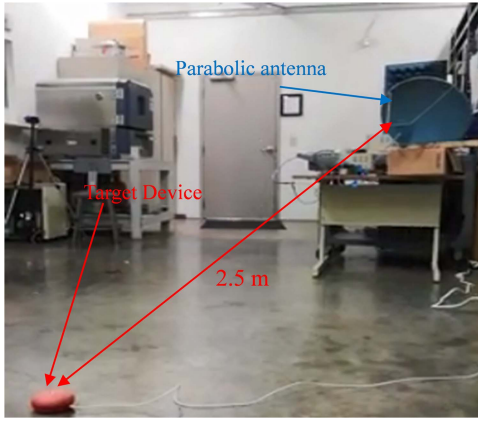


Fig. 16. Transfer function of the sensitive location.

Fig. 17. Voice command attack on a *Smart Speaker 1* with a parabolic antenna.

the 2-port S-parameter measurement setup of the DUT. The positive terminals of the two identical coaxial cables were soldered on the sensitive location and the power pin of the microphone; the negative terminals are soldered on the adjacent ground pins, as shown in Fig. 15(b). The measured 2-port S-parameter data were transformed into the ABCD matrix to obtain the transfer function, as shown in Fig. 16. The plot in circles in Fig. 16 represents the analyzed sensitive frequencies in Fig. 12. It can be seen that the strongest coupling happens at around 8 GHz which meets our expectation based on the previous results shown in Fig. 12.

### C. Maximum Attack Distance for Real Voice

The maximum attack distances for different target devices were achieved with a square-rooted attack voice command. The maximum distance reached for *Smart Speaker 1* is 2.5 m with a parabolic antenna, as shown in Fig. 17. The antenna is giant, but the adversary can drive a truck on the street and sent the attack messages to the smart rooms, and we can also change the antenna to smaller ones. For different products, we obtained varying maximum attack distances based on the current setup with different antennas, as shown in Table I. The maximum attack distance varied from 20 cm to 2.5 m for different target devices with an output power of only 2.5 W, and the antenna gain varies from 15 to 22 dBi. Despite the

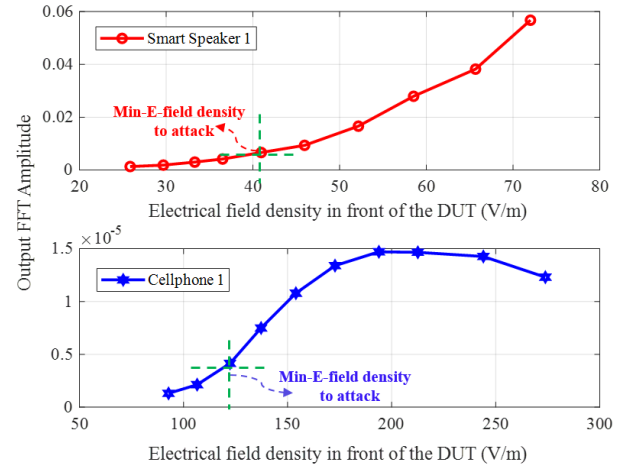


Fig. 18. Relation between input and output power.

attack distances are small for some products compared to previous works [8], the attack distance can be increased by employing a high power amplifier.

### D. Attack Electrical Field Density Analysis

In this analysis, the attack distance is fixed, and different attack powers are applied to generate different electrical field densities in front of the DUTs.

The power density in front of the smart speakers can be derived from the Friis transmission equation, as shown in the following equation:

$$P_D = \frac{P_t G_t}{4\pi d^2}. \quad (8)$$

The electric field strength at a given location can be obtained as follows [42]:

$$E = \sqrt{P_D Z_0} = \sqrt{120\pi P_D} \quad (9)$$

where  $P_t$  is the transmitter power (either the peak or average power),  $G_t$  is the gain of the antenna,  $d$  is the distance, and  $Z_0$  is the air impedance. In this case, the electric field strength in front of the device can be characterized. The minimum required power density and electrical field intensity in front of the smart speakers are listed in Table I.

The gain of the antenna is 18 dBi at 8 GHz for the *Smart Speaker 1* attack and 22 dBi at 18 GHz for the *Cellphone 1* attack. The single-tone audible output spectrum is obtained in the recorded files. The relation between the  $E$ -field density in front of the DUTs and the obtained single-tone audible output is shown in Fig. 18. The green lines indicate the minimum  $E$ -field densities needed for the different target devices to recognize a real voice command. The different target devices exhibit varying limits and coupling strengths; for example, to attack *Smart Speaker 1*, the required minimum  $E$ -field density in front of the device is around 40 V/m, with a distance of 20 cm. However, for *Cellphone 1*, the requirement is around 125 V/m. In addition, the recognition level varies due to the noise cancellation technique applied by the *Cellphone 1*. The coupling efficiency which is the ratio between the input and output power can be obtained by calculating the slope of the curve.



## V. CONCLUSION

This article presents an optimized EM attack process and the sensitivity analysis. The mechanism of the nonlinearity in the microphone circuit was described. The attack signal was preprocessed to increase the probability of a successful attack based on the nonlinearity characteristics, and measurements were performed for the single-tone signal attack to illustrate the effectiveness of the attack signal preprocessing. In addition, a methodology for sensitivity frequency analysis was discussed in order to find the most sensitive carrier frequency of a given product. The coupling sensitivity is studied based on near-field injection technique, and the transfer function from the sensitive location to the microphone under test is measured. The real voice commands were also successfully injected and executed by the target devices. Different maximum distances have been reached for different target devices. Generally, the maximum distance is depending on the output power of the antenna and types of DUT. A model can be built to estimate the required attack power (output power from the antenna or the power density in front of the device). Thus, the designer can optimize their device based on their standards regarding attackable distance and power. To eliminate the attack possibility, some countermeasures are proposed below for future studies:

### A. Layout Optimization

Most EM threats arise due to an unintentional antenna structure [33], [34], [36] associated with the PCB layout design. Additional efforts to minimize exposed traces in the outer layers can reduce EM coupling. Moreover, the unintentional antenna structure near the microphone can act as an antenna to receive the I-EMI signal and conduct it to the microphone, allowing the microphone to demodulate the voice command.

### B. Shielding Technique

Because the EM field must travel to the microphone circuit, a full structure shielding technique can be integrated into the device by exposing only the necessary parts, for example, by including a small hole for the microphone. An outer metal shield will prevent the field from coupling to the interconnects of the microphone circuit. Although the cost will increase, security risks can be minimized.

### C. Inaudible Voice Command Detection

RF modulated signals operate at high frequencies; thus, another circuit can be added to detect the high-frequency component, in parallel to the microphone circuit. If modulated RF signals are detected, the circuit can give a signal to the microphone to stop listening. Thus, the smart device will not execute the attack command.

The future investigation will conduct in the protection technique development.

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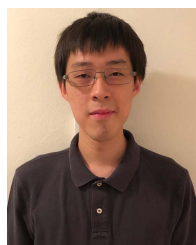
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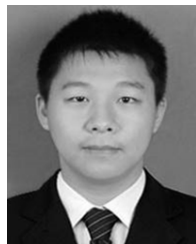
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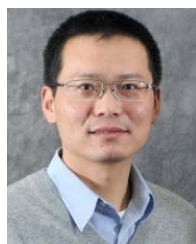
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