

Protecting Health Care and Cyberphysical Systems: Wicked BiZaRrE Semiconductor Physics of Sensor Security: Sensors, Signals, Semiconductors, Software Systems



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NORTHEASTERN
1898

[USENIX Security]

Light Commands: Laser-Based Audio Injection Attacks on Voice-Controllable Systems

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[IEEE S&P]

Poltergeist: Acoustic Adversarial Machine Learning against Cameras and Computer Vision

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[IEEE Euro S&P]

WALNUT: Waging Doubt on the Integrity of MEMS Accelerometers with Acoustic Injection Attacks

Timothy Trippel, Ofir Weisse, Wenyuan Xu*, Peter Honeyman, Kevin Fu

Computer Science and Engineering, University of Michigan

*Computer Science and Engineering, University of South Carolina

<https://spqr.eecs.umich.edu/walnut/>

[IEEE SENSORS]

Why Lasers Inject Perceived Sound Into MEMS Microphones: Indications and Contraindications of Photoacoustic and Photoelectric Effects

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Embedded Security Research

[ACM CCS]

Session 10A: Cyberphysical Security CCS '19, November 11–15, 2019, London, United Kingdom

Trick or Heat? Manipulating Critical Temperature-Based Control Systems Using Rectification Attacks

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Session 10A: Cyberphysical Security CCS '19, November 11–15, 2019, London, United Kingdom

Adversarial Sensor Attack on LiDAR-based Perception in Autonomous Driving

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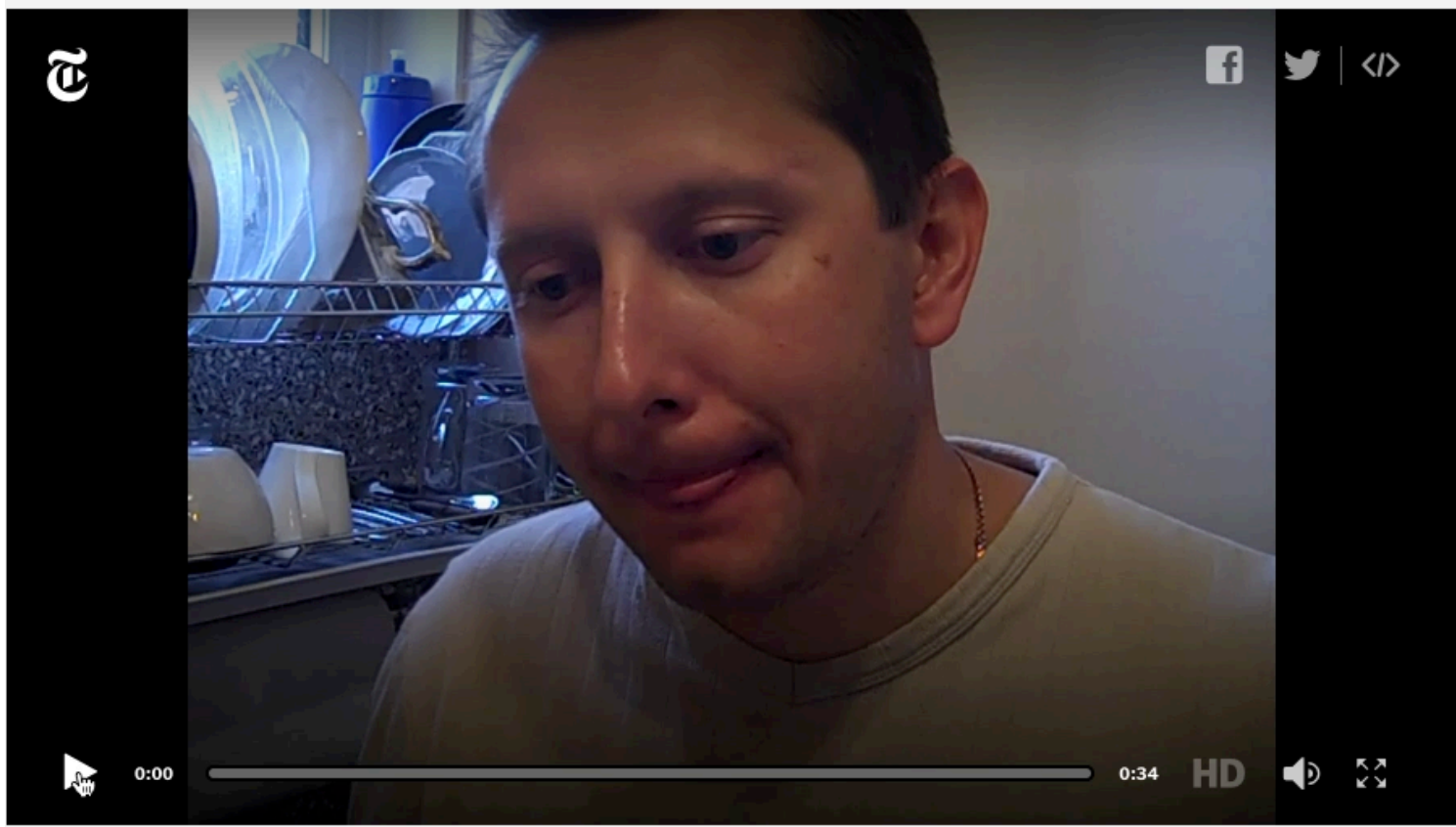
DOI:10.1145/3176402

Kevin Fu and Wenyuan Xu

Inside Risks
Risks of Trusting
the Physics of Sensors

Protecting the Internet of Things with embedded security.

Wicked Bizarre Physics of Sensor Security • k.fu@northeastern.edu • spqrlab1.github.io



VIDEO

Mobile Phone Turns On Oven

By James Dwyer | Aug. 21, 2009 | 0:34

A New Yorker explains to Jim Dwyer how his cellphone turns on his oven when it rings.

Share:

<https://www.nytimes.com/video/multimedia/1247464146747/mobile-phone-turns-on-oven.html>

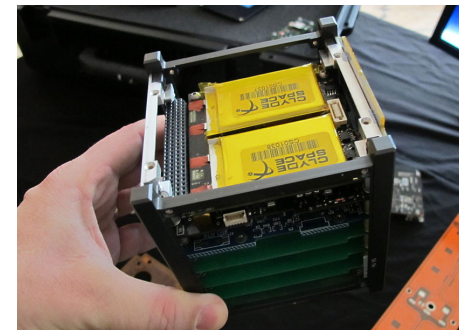
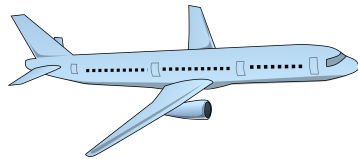
Security is hard.

Correctness is easy.



Photo by Kevin Fu

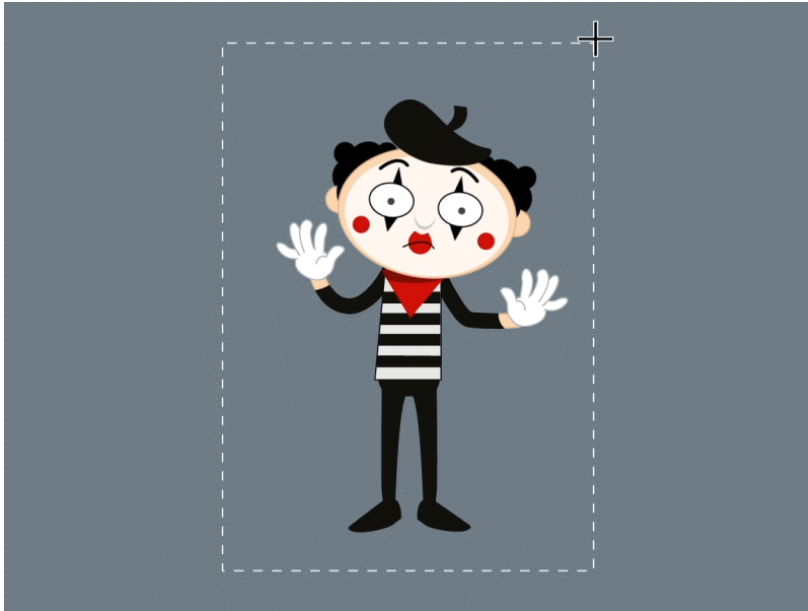
Sensors are Everywhere



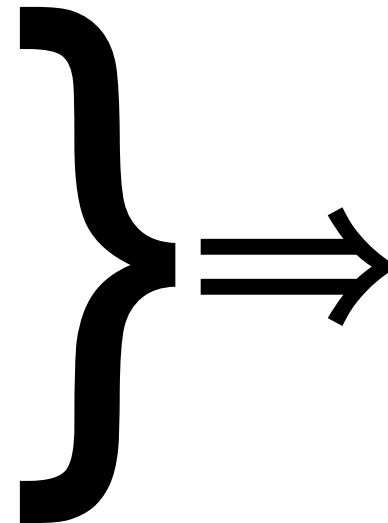
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Digital Abstraction != Force Field

intentional interference violates assumption of sensor output integrity



- Vibration
- Acoustics
- RF
- Light
- Heat



Analog
Sensor
Spoofing

Do Not Blindly Trust Sensors

Sensors are a proxy for reality

- **Thermocouple interpolates from a voltage potential**
- **Not necessarily temperature**



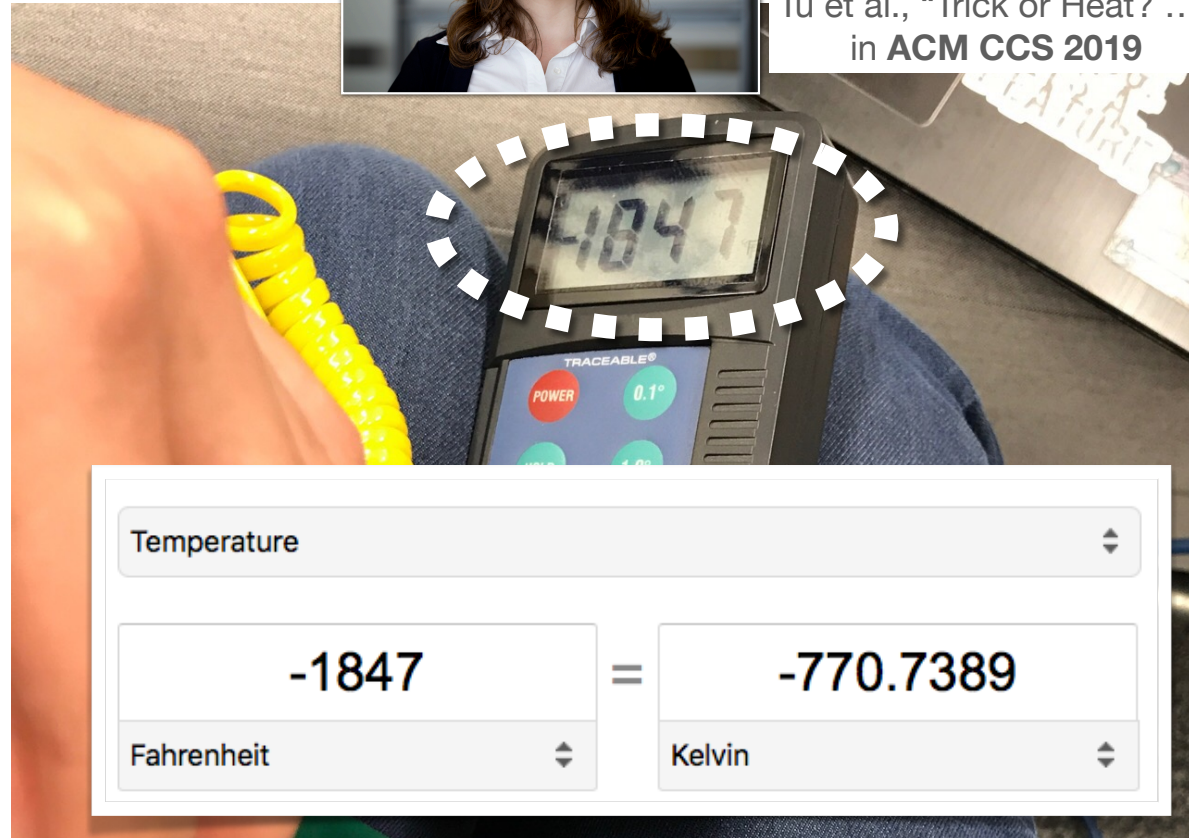
Absolute Zero Day Attack

Sensors are a proxy for reality

- ➔ Microphone measures electromagnetic potential of copper spool
- * Not necessarily sound
- ➔ MEMS accelerometer measures vibration of a tiny element
- * Not necessarily sensor acceleration
- ➔ Thermocouple interpolates from a voltage potential
- * Not necessarily temperature



Dr. Sara Rampazzi
joins UFL faculty
Tu et al., “Trick or Heat? ...”
in ACM CCS 2019



Where Do Thermocouples Matter?

The New York Times

How to Ship a Vaccine at -80°C , and Other Obstacles in the Covid Fight

Developing an effective vaccine is the first step. Then comes the question of how to deliver hundreds of millions of doses that may need to be ...



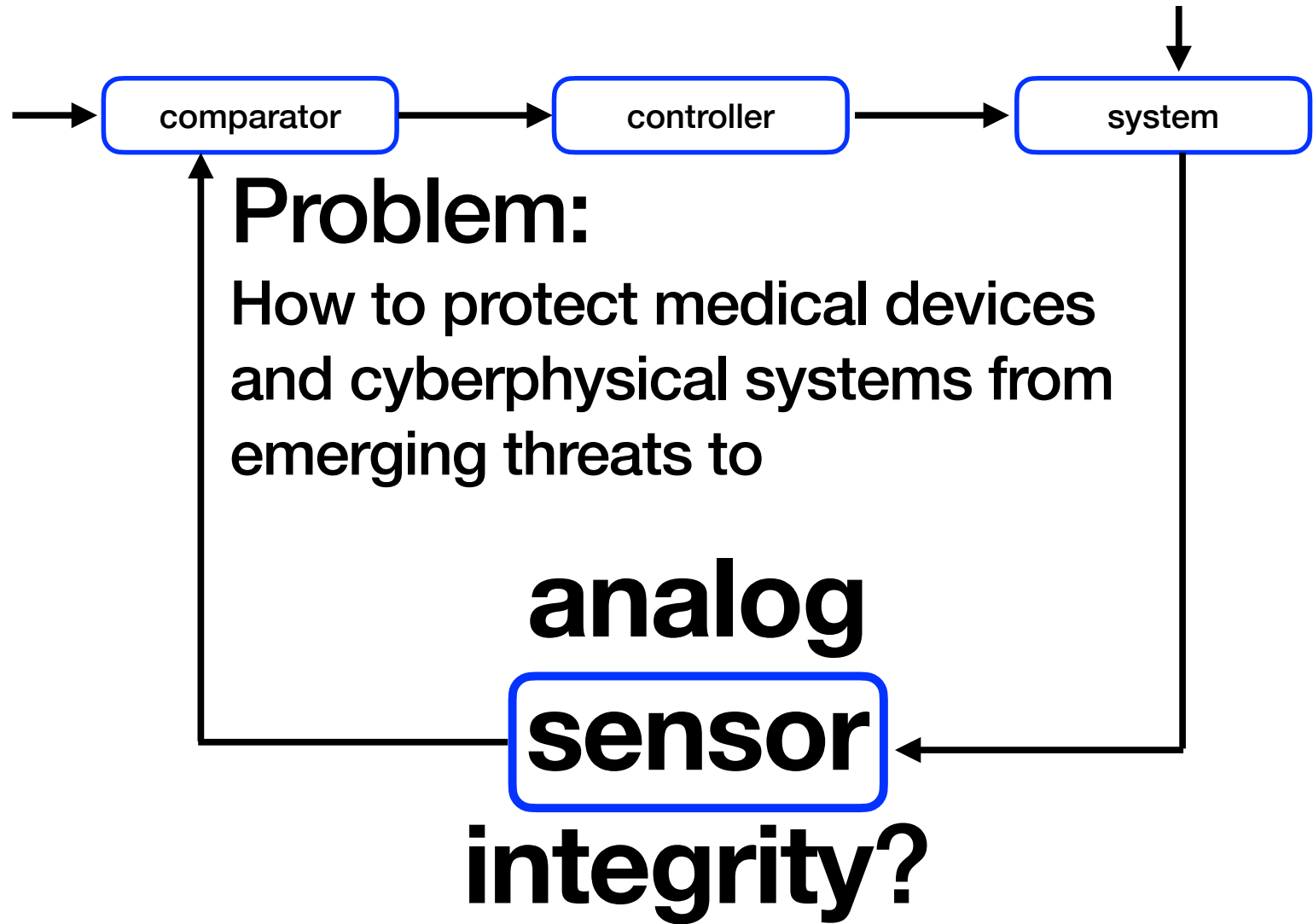
Blog / Temperature measurements and temperature control in the IVF lab are crucial for your results

Temperature measurements and temperature control in the IVF lab are crucial for your results

Posted by Jaco Geyer, Jan 26, 2016 6

At Risk: Closed-Loop Feedback Systems

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Outline: Protecting Sensor Integrity

Today: taste of sensor security research across three modalities:

- Defending against **radio-based attacks** on sensors
- Defending against **sound-based attacks** on sensors
- Defending against **light-based attacks** on sensors

Intentional Electromagnetic Interference (Or Don't Trust Your Sensors)



**“Ghost Talk: Mitigating EMI Signal Injection Attacks against Analog Sensors” by Foo Kune et al.
In Proc. IEEE Symposium on Security and Privacy, 2013.**

**Joint work with Denis Foo Kune (U. Michigan),
John Backes (U. Minnesota), Shane Clark (U. Mass Amherst),
Dr. Dan Kramer (Beth Israel Deaconess Medical Center),
Dr. Matthew Reynolds (Harvard Clinical Research Institute),
Yongdae Kim (KAIST), Wenyuan Xu (U. South Carolina)**

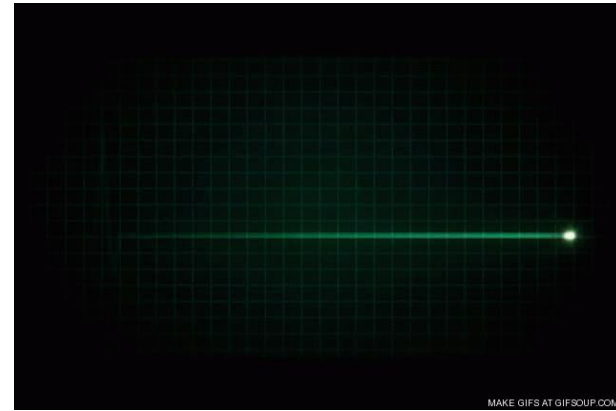


Which one is the real cardiac signal?

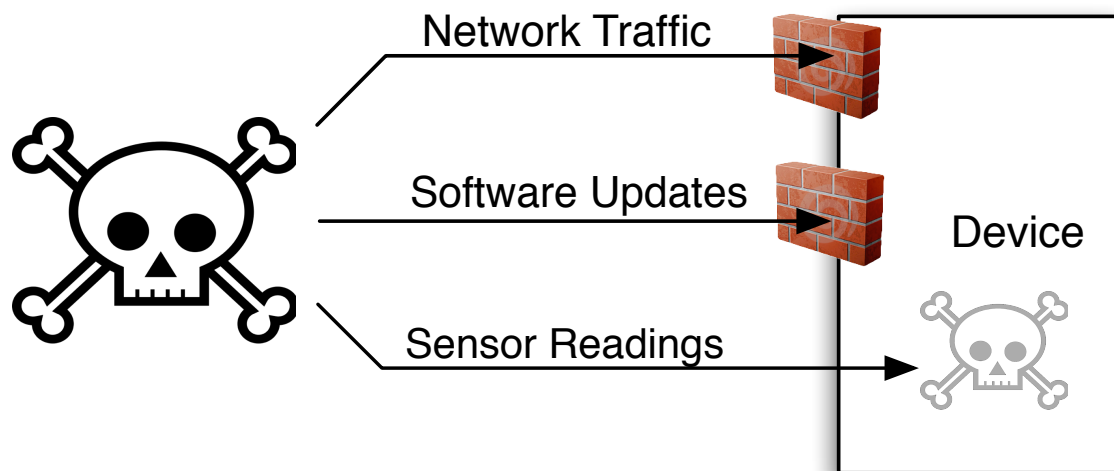
A



B

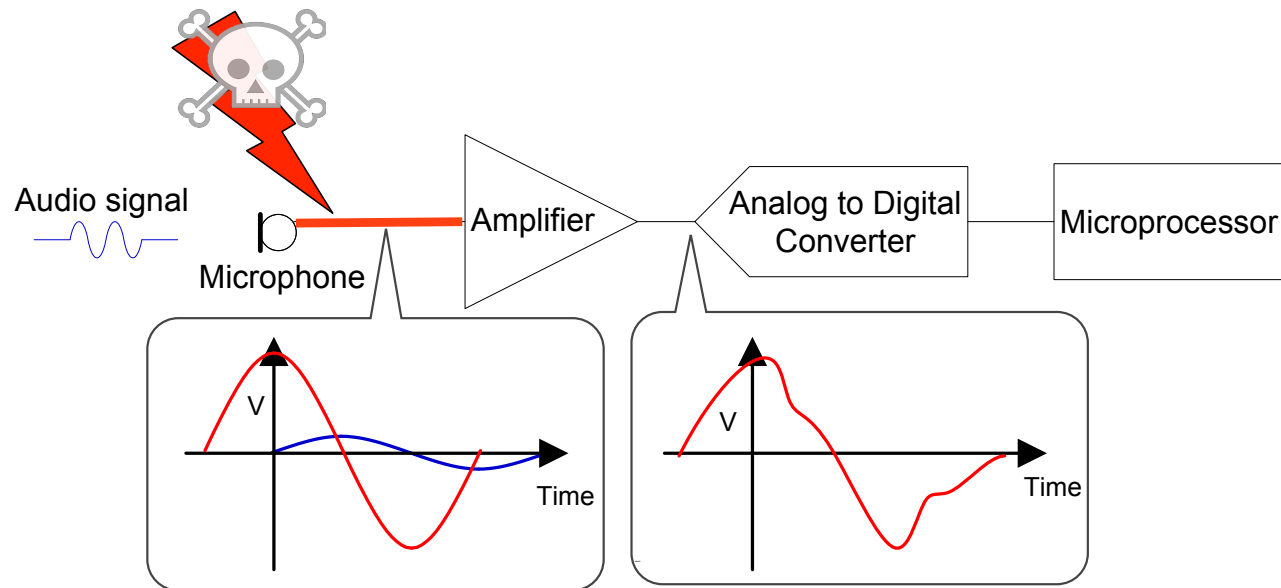


Inputs may not be trustworthy



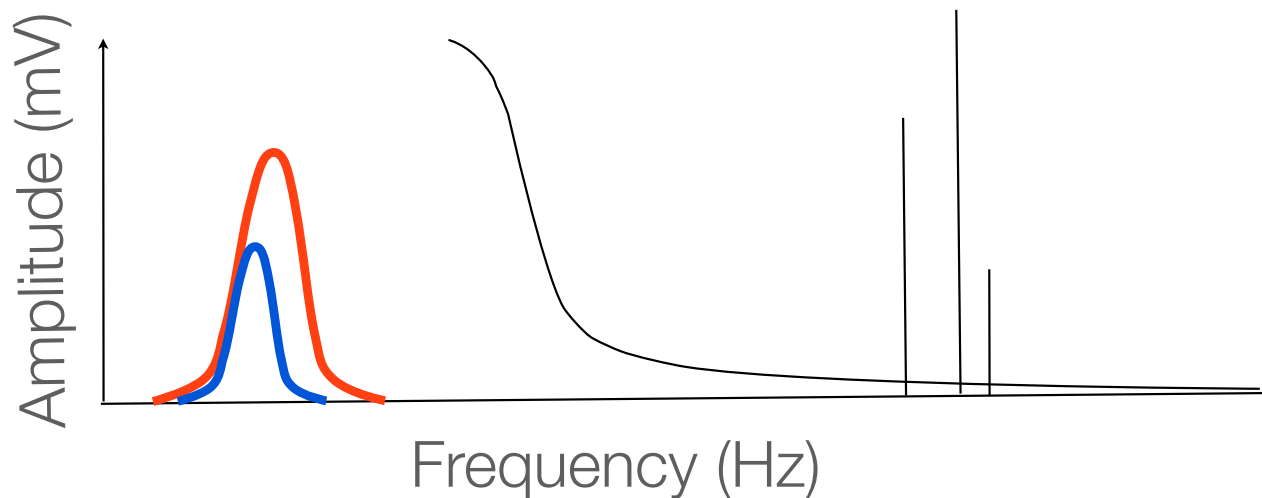
Ghost Talk: **Intentional** interference

- Conducting traces can couple to EMI (back-door).
- Sensitive analog sensors can be affected.

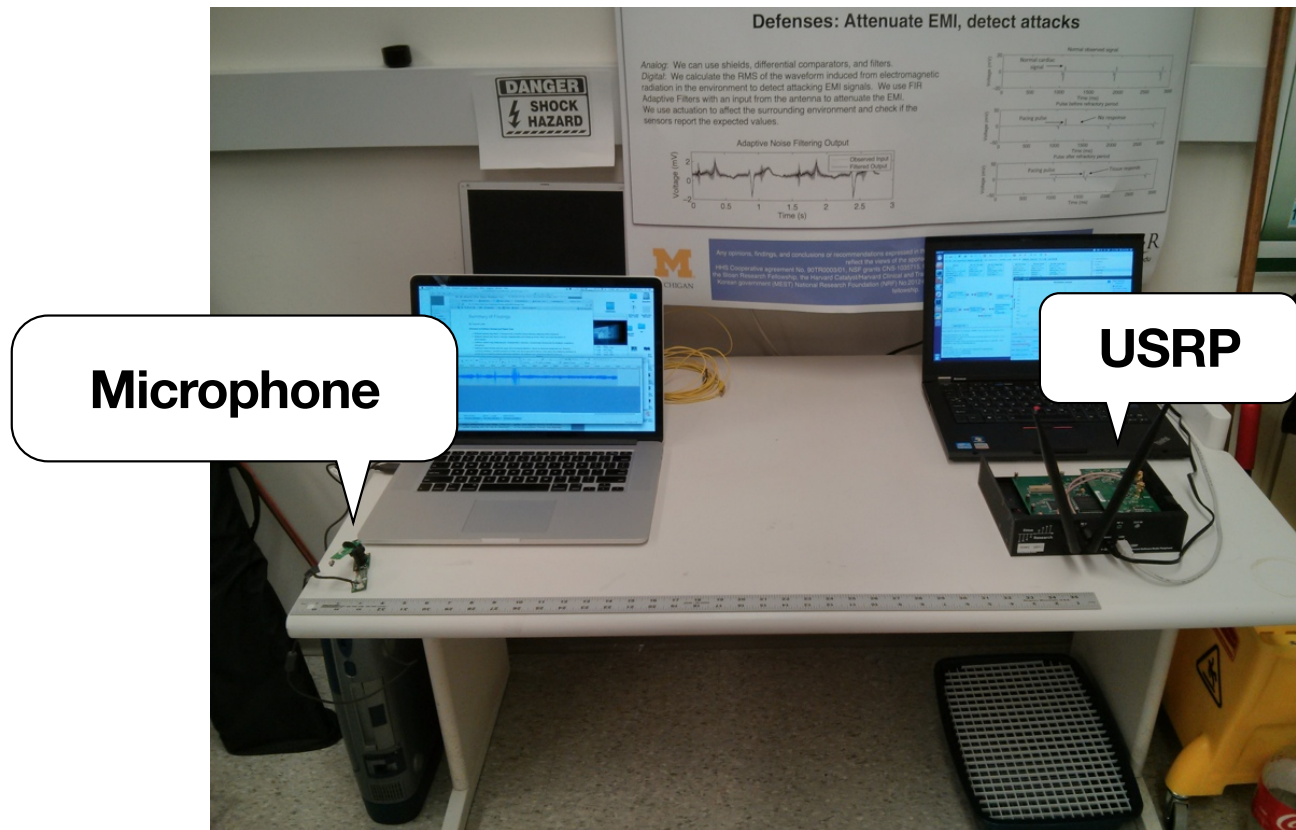


Fundamental Problem: Baseband

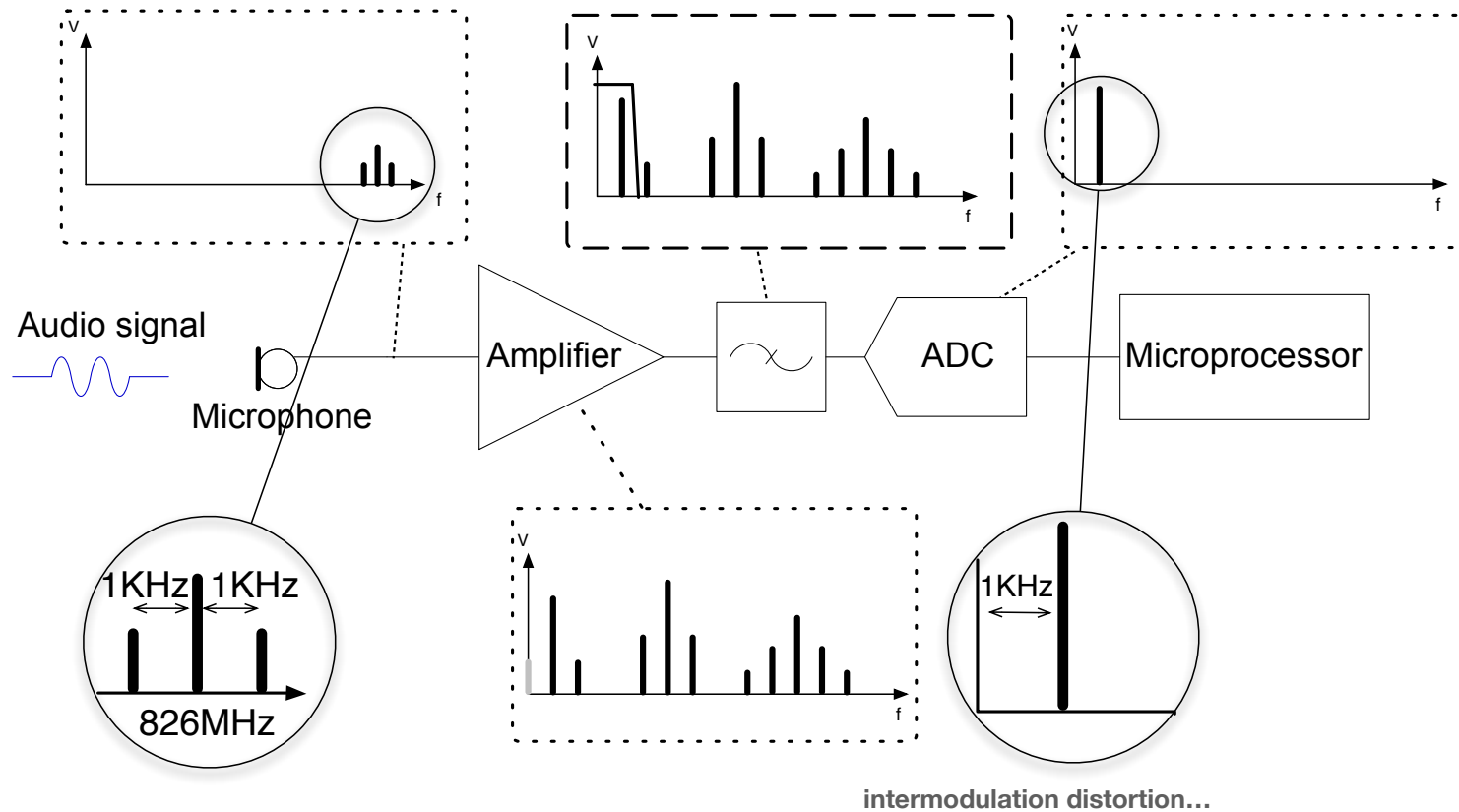
- Baseband: frequency range of desired signals.
- Interference outside the baseband is easy to filter.
- Interference in the baseband is hard to remove.



Microphone Interference with RF



Non-Linearity: Self Demodulation



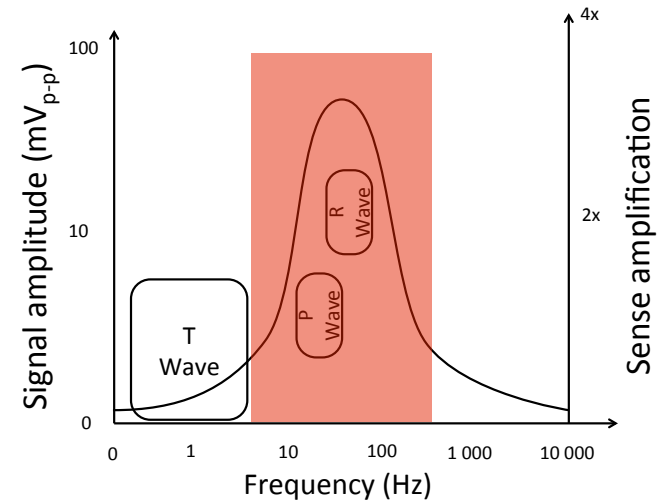
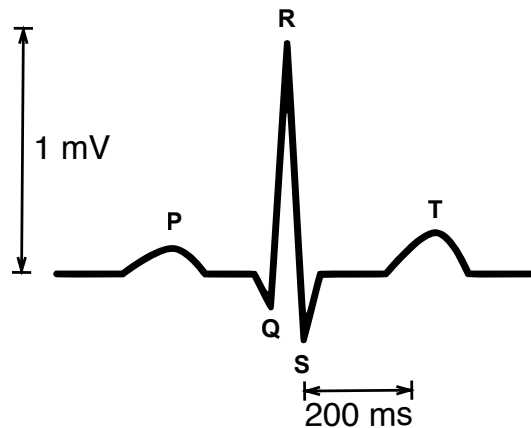
Intentional EMI on cardiac devices

- Pacemakers, defibrillators
- Electrocardiogram machines



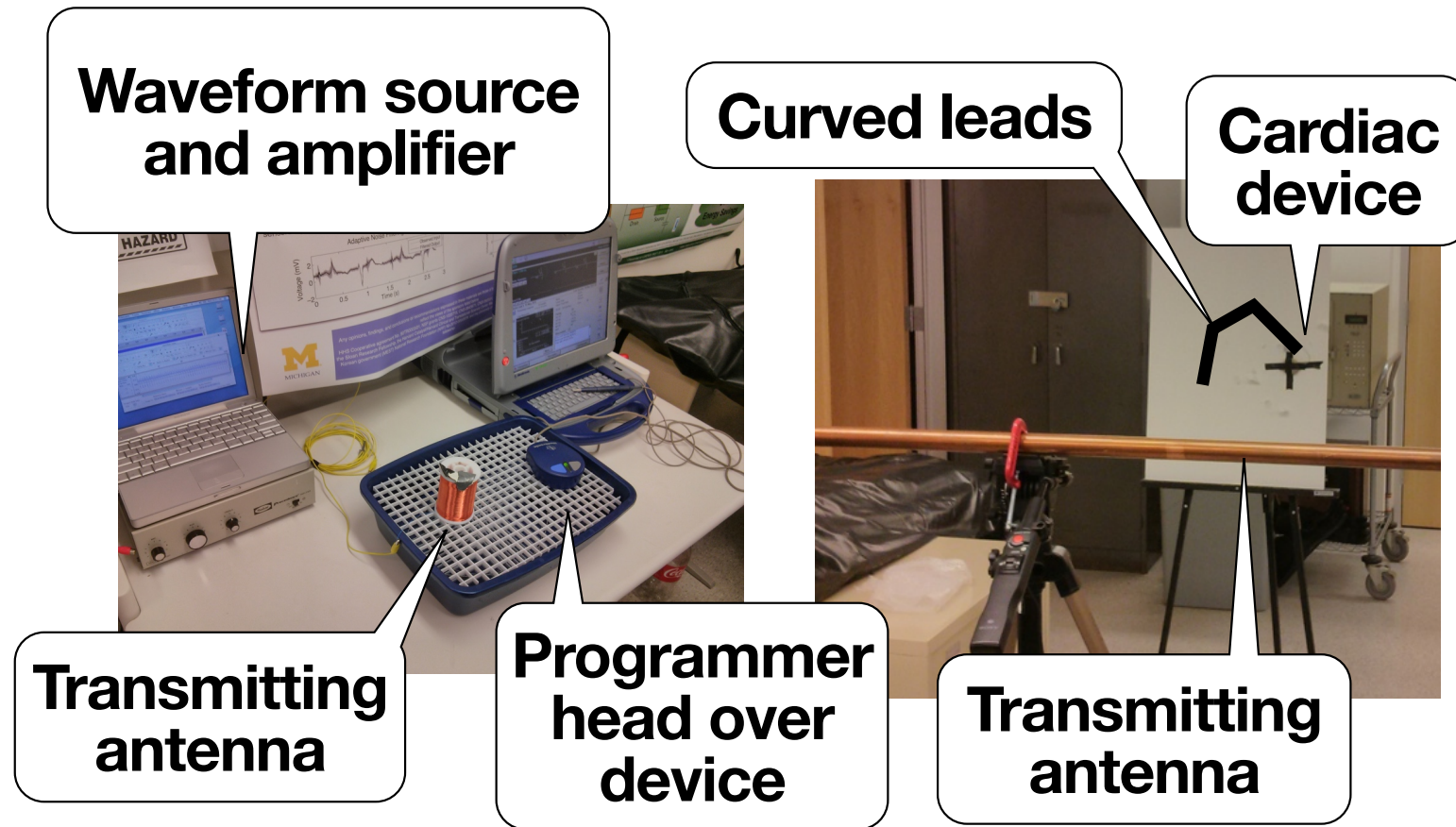
Cardiac devices vulnerable to baseband EMI

- Filter high frequency
 - 800MHz and GHz range: attenuation of up to 40dB
- Can't filter baseband



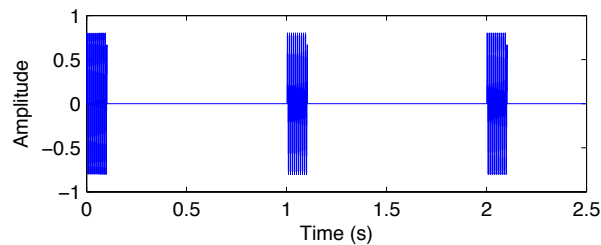
Cohan et al, 2008

Experiment: Implants & Emitters

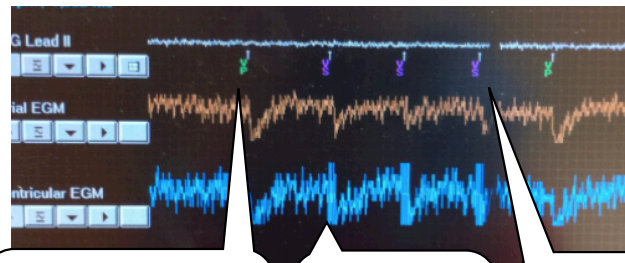
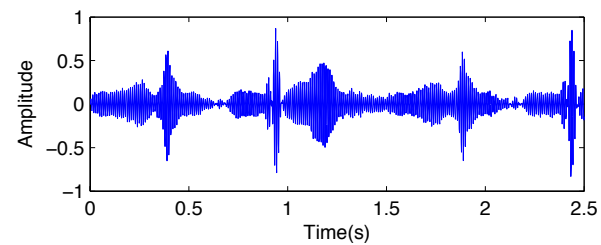


Results: Waveforms & Responses

Pulsed sinusoid



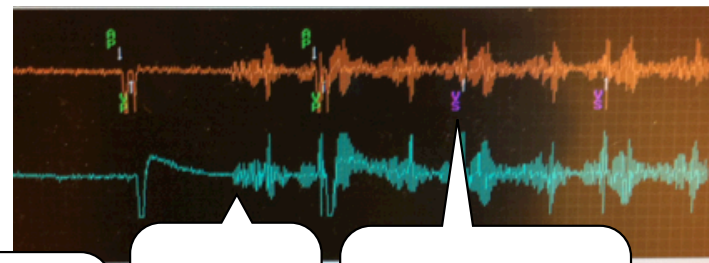
Modulated heart beat



**Ventricular
pace**

**Signal
onset**

**Ventricular
sense**

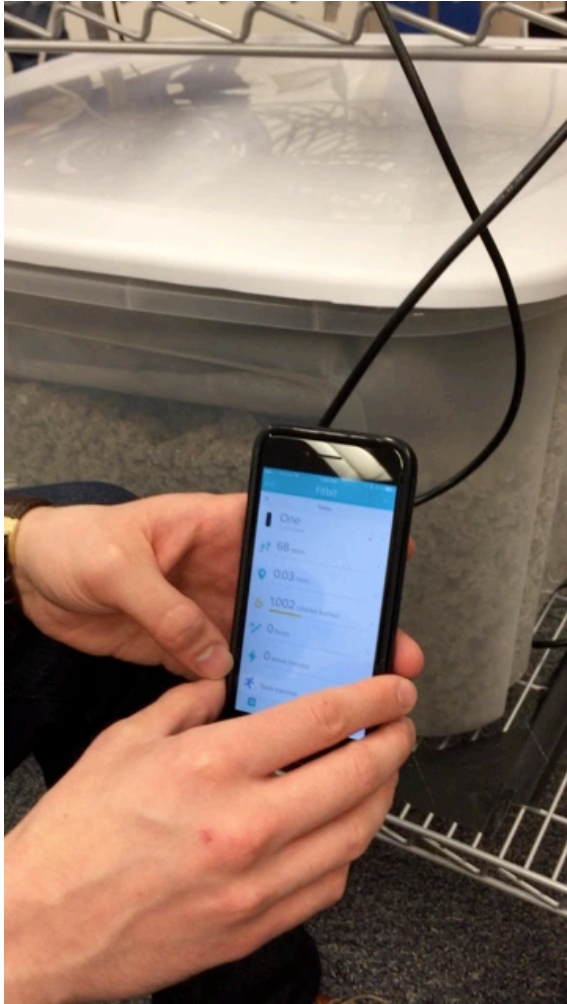


**Signal
onset**

**Ventricular
sense**

Good News: Distance

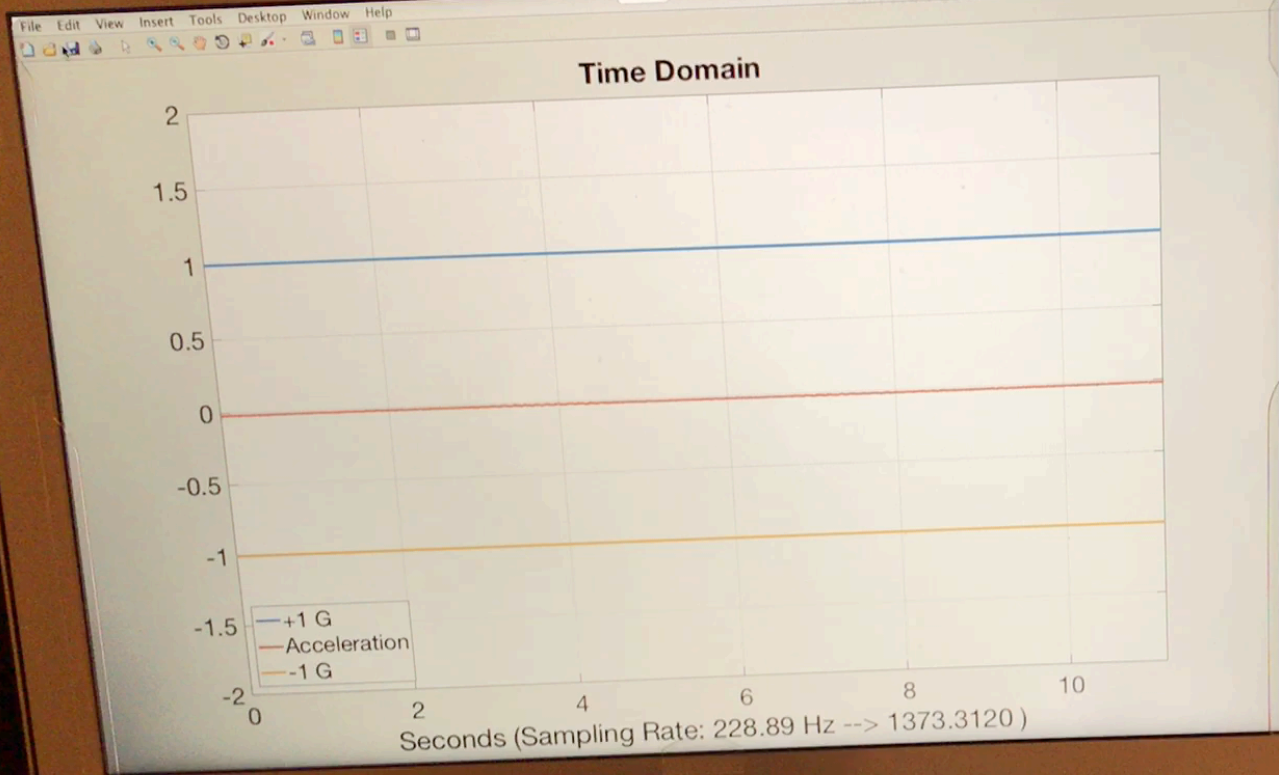
Device	Open air pacing	Open air Defib	Saline tips only	SynDaver
Medtronic Adapta	1.40m	NA	3cm	Untested
Medtronic InSync Sentry	1.57m	1.67m	5cm	8cm
Boston Scientific Cognis	1.34m	No defib	Untested	Untested
St. Jude Promote	0.68m	No defib	Untested	Untested



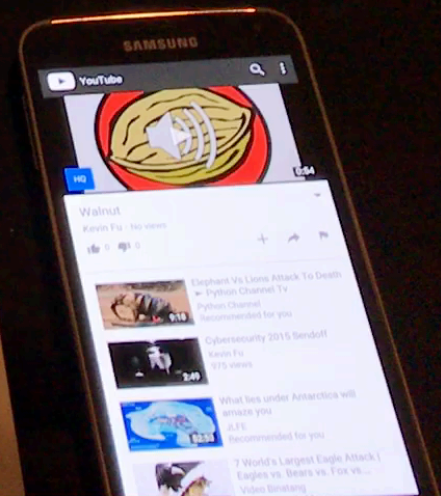
Sound and MEMS Sensor Security



[“WALNUT” by Trippel et al., IEEE Euro S&P 2017]



MacBook Air



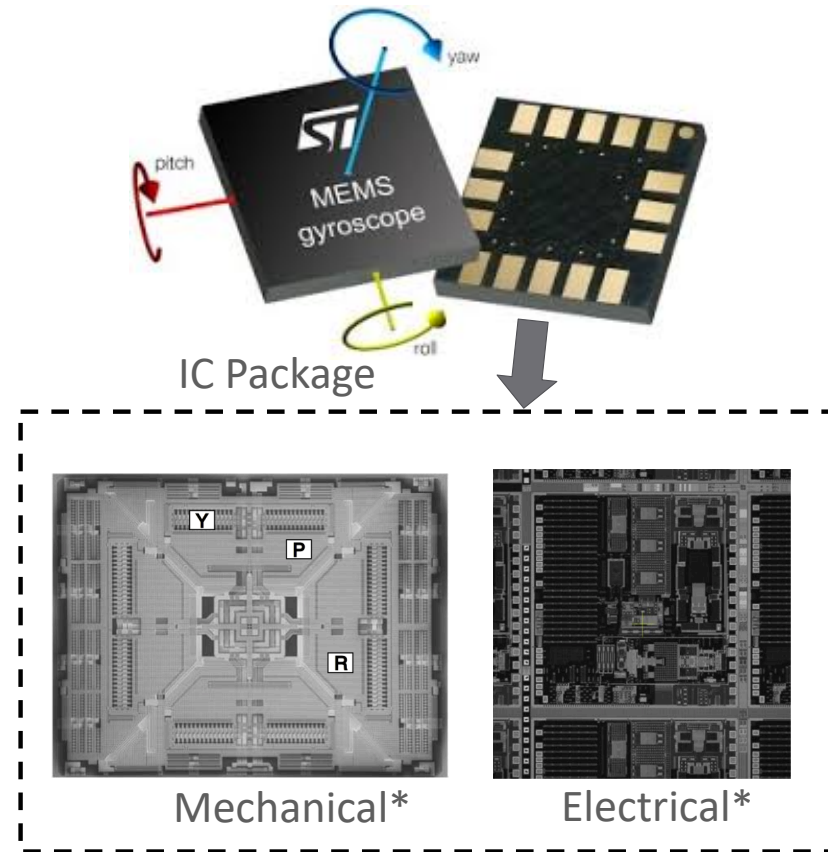
MEMS Sensors

- Micro-Electro-Mechanical Systems

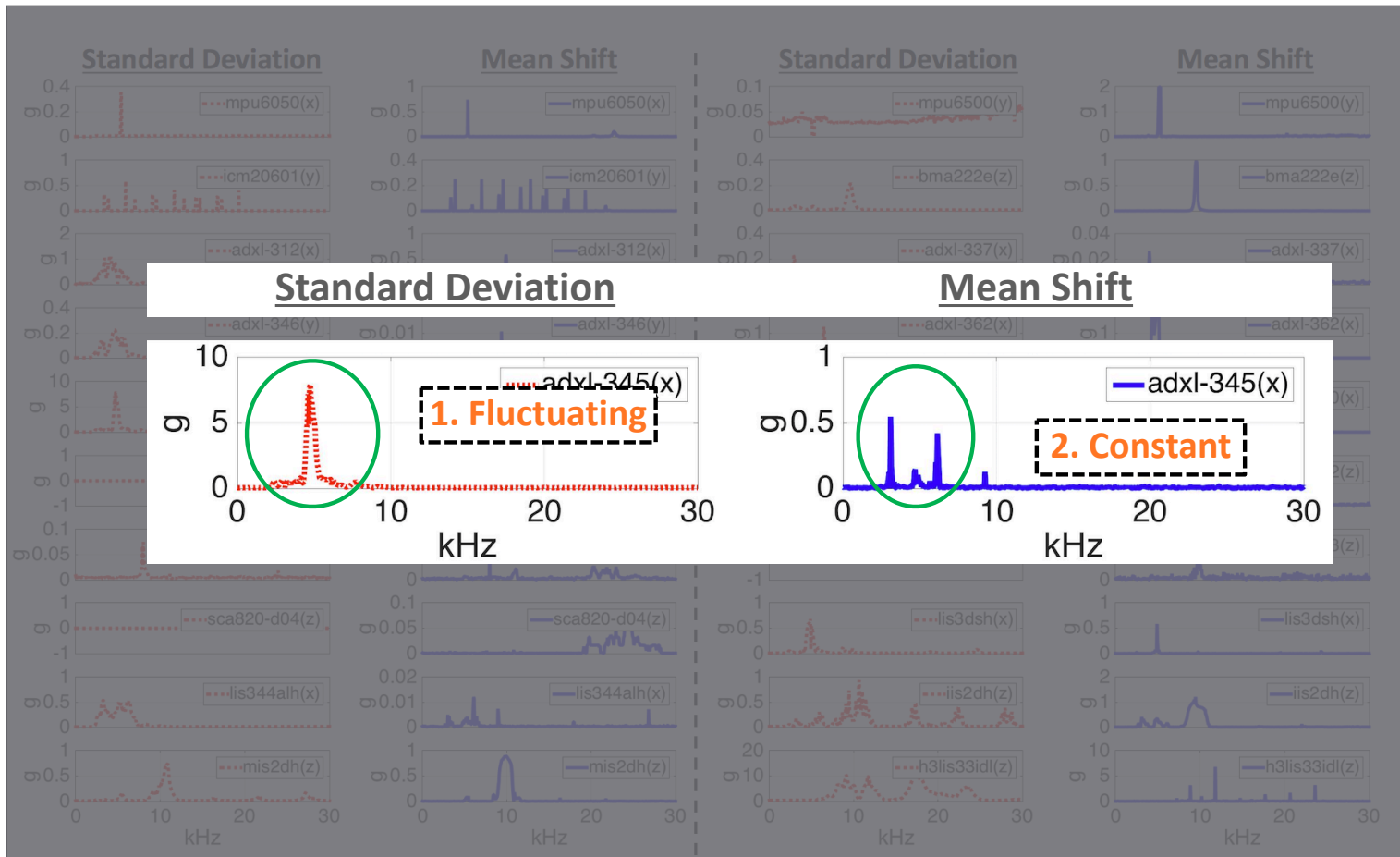
- Accelerometers
- Gyroscopes
- Clocks

- Advantages

- Low cost
- Low power
some < 1 mA
- Small integrated circuit



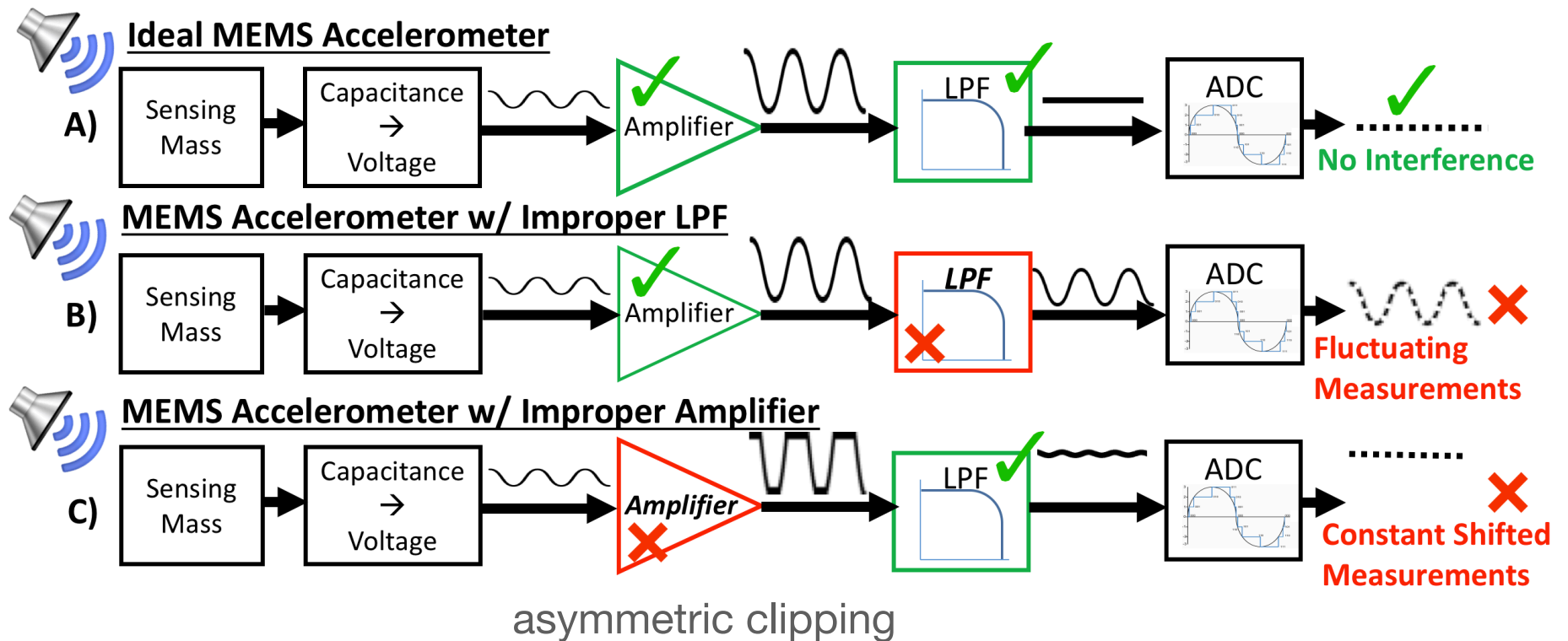
*Photos courtesy of "Everything about STMicroelectronics' 3-axis digital MEMS gyroscopes – Technical Report", by STMicroelectronics.



Signal Distortion

Two types of spoofed acceleration

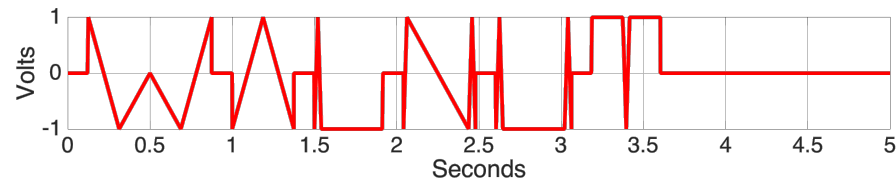
- Fluctuating accelerometer output
- Constant accelerometer output



Output Control Modulation

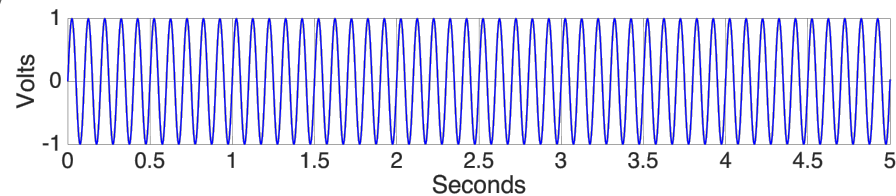
Desired Accelerometer
Output Signal

+

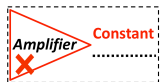
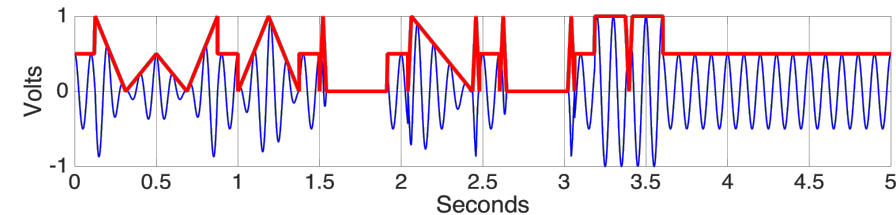


MEMS Resonant Frequency
(Carrier Signal)

=



Modulated Acoustic
Attack Signal



ANALOG DEVICES ADVISORY TO ICS ALERT-17-073-01

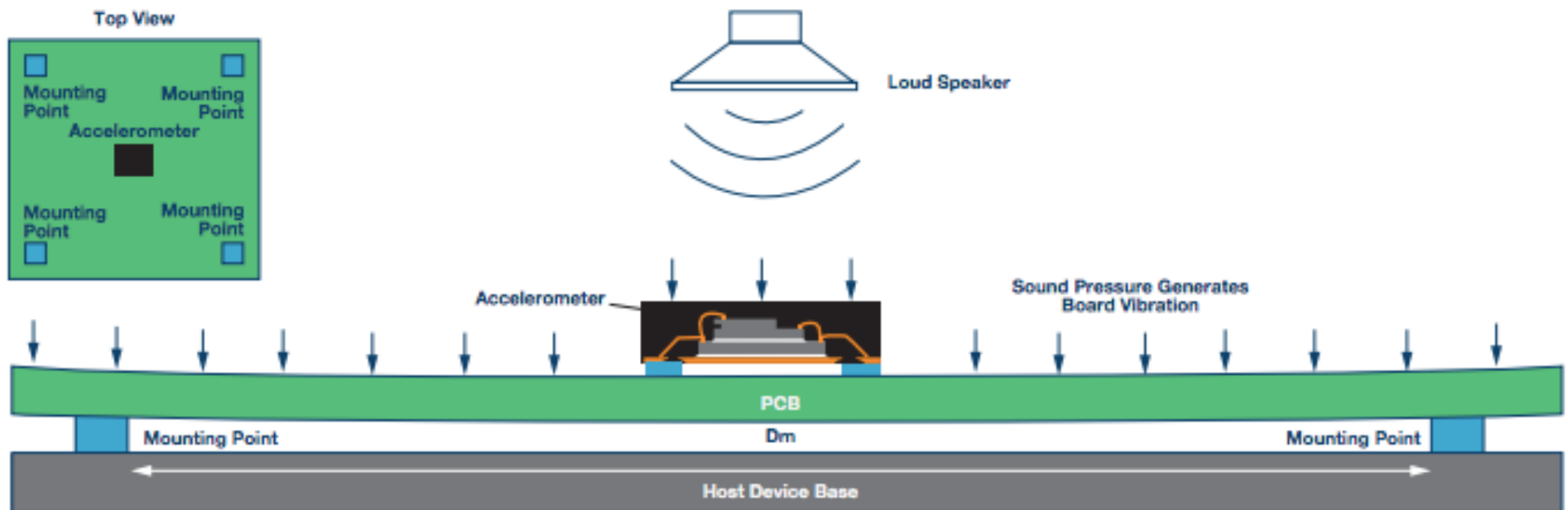


Figure 1. MEMS accelerometer board and mounting with acoustic vibration from off-board speaker.

ANALOG DEVICES ADVISORY TO ICS ALERT-17-073-01

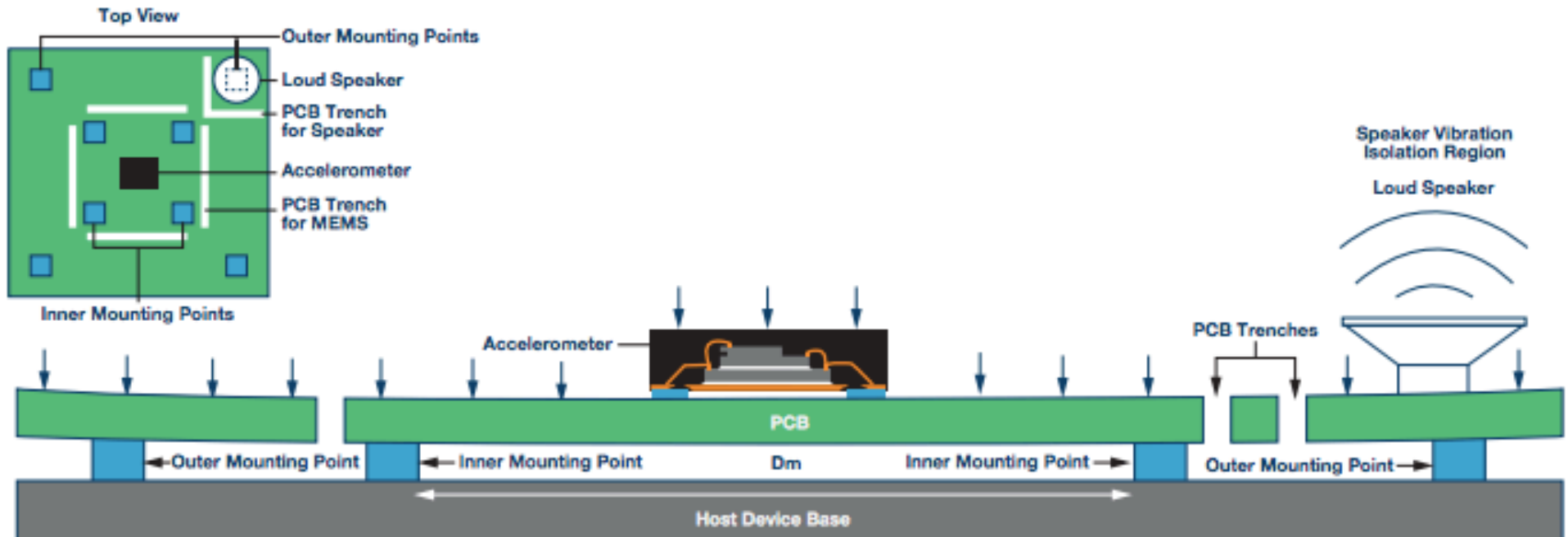


Figure 2. MEMS accelerometer board and mounting with acoustic and mechanical vibration from on-board speaker.

ICS-CERT is also working with several of the cooperative vendors to identify a list of affected devices that contain vulnerable capacitive MEMS accelerometer sensors.

The following MEMS Accelerometer sensors may be affected:

- Bosch BMA222E,
- STMicroelectronics MIS2DH,
- STMicroelectronics IIS2DH,
- STMicroelectronics LIS3DSH,
- STMicroelectronics LIS344ALH,
- STMicroelectronics H3LIS331DL,
- InvenSense MPU6050,
- InvenSense MPU6500,
- InvenSense ICM20601,
- Analog Devices ADXL312,
- Analog Devices ADXL337,
- Analog Devices ADXL345,
- Analog Devices ADXL346,
- Analog Devices ADXL350,
- Analog Devices ADXL362,
- Murata SCA610,
- Murata SCA820,
- Murata SCA1000,
- Murata SCA2100, and
- Murata SCA3100.



ANALOG DEVICES ADVISORY TO ICS ALERT-17-073-01

The following derivations based on a single periodic sound frequency can be used to relate the board deflection to acceleration level.

The board harmonic deflection can be defined as:

$$deflection = d_{bd} \times \sin(\omega \times t) \quad (1)$$

where d_{bd} is the amplitude of the board deflection under the sound pressure and ω is the frequency of the sound.

The acceleration produced by the harmonic deflection is:

$$acceleration = d_{bd} \times \omega^2 \times \sin(\omega \times t) \quad (2)$$

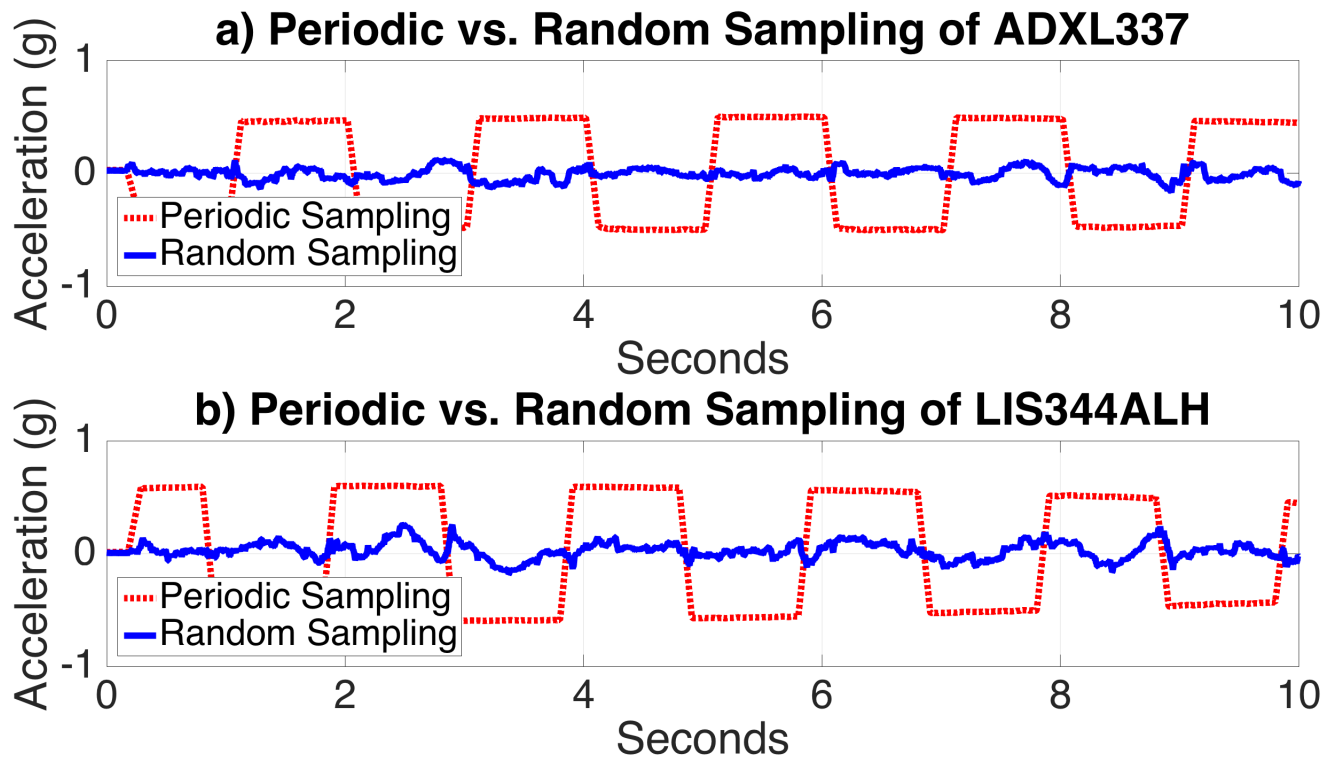
In the case where the sound frequency matches the board resonant frequency, the deflection will be amplified by the quality factor (Q_{bd}) of the board and Equation 2 will be modified as:

$$acceleration \text{ at board resonance} = Q_{bd} \times d_{bd} \times \omega^2 \times \sin(\omega \times t) \quad (3)$$

By inspecting Equation 3, one can find the following methods to mitigate the board acceleration effect. These methods have been either implemented in Analog Devices' accelerometer products or advised to the customers for system design considerations, whichever is applicable.

Randomized Sampling

- Destroy predictability of sampling regime
- Randomize delay at each sampling interval



["WALNUT" by Trippel et al., IEEE Euro S&P 2017]

Lasers & Sensor Security



The New York Times [@nytimes](#) · Nov 5, 2019

"This is so basic." Researchers say they have found a way to take over voice-controlled digital assistants like Apple's Siri and Amazon's Alexa — and all it took was a cleverly pointed light.

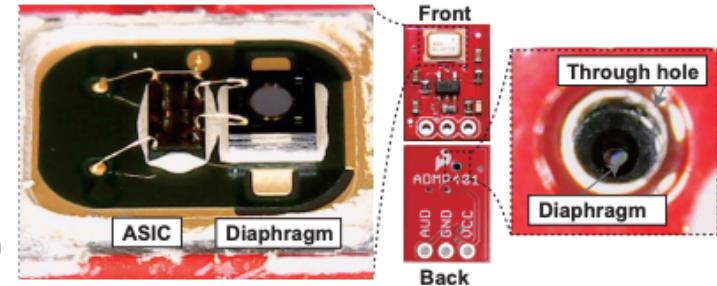
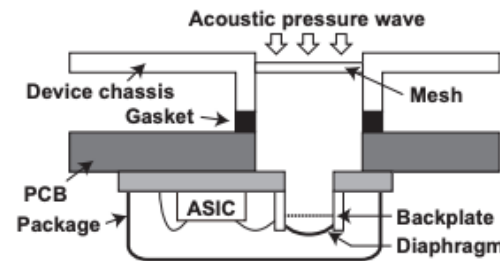


NBC Nightly News with Lester Holt [@NBCNig...](#) · Nov 4, 2019

The smart speaker in your home may not be as secure as you think.

Researchers discovered that Amazon's Alexa, Apple's Siri and Google Home can be hacked by laser pointers and flashlights.

[@jolingkent](#) has the details.



Sugawara et al., Light Commands: Laser-Based Audio Injection Attacks on Voice-Controllable Systems, USENIX Security 2020



["Light Commands" by Sugawara et al., USENIX Security 2020]



LIGHT COMMANDS

[“Light Commands” by Sugawara et al., USENIX Security 2020]



CSE COMPUTER SCIENCE
AND ENGINEERING
UNIVERSITY OF MICHIGAN

["Light Commands" by Sugawara et al., USENIX Security 2020]

So, you depend on sensors?



Creating Trustworthy Sensors

Demystify analog sensor attack surface

👉 Test to security **FAILURE**, not test to $\neg(\text{ツ})\neg$

👉 Unwrap abstractions of electrical engineering, mechanical engineering, materials science

Ad-hoc security \Leftrightarrow measurable science

👉 Physically de-risk **intentional interference** with more deliberate HW specs & design (e.g., resonance)

Rethink ICs and hardware-software APIs

👉 Convey to SW stack **WHY** trust sensor output

👉 HW should expose **HINTS** of trustworthiness

Analog Cybersecurity Risks

- Computers have always been vulnerable to analog cybersecurity threats
- What's changing?
 - Degree of connectedness and dependence
 - From human-in-the-loop to automated consequences
 - Increased risks to availability and integrity
- Maybe it's a not a good idea to put a computer in everything unless there's a good reason



Embedded Security References

- CRA's Grand Challenges for Embedded Security Research in a Connected World
- Back door acoustic injection
 - Gyroscopes: Drone DoS [Son et al., USENIX Sec '15], Dolphin Attacks: Ultrasound voice recognition injection [Zhang et al., ACM CCS'17]
 - Walnut: Acoustic injection on MEMS accelerometers [Trippel et al., IEEE Euro S&P'17]
- RF, IR, EMFI injection
 - Tire pressure sensors: [Rouf et al., USENIX Sec '10], Infusion pumps [Park et al., USENIX WOOT '16], BADFET [Cui & Housley, USENIX WOOT '17]
 - Ghost Talk: RF injection on microphones, pacemakers [Foo Kune et al., IEEE S&P '13], GSMem [Guri et al., USENIX Sec '15]
- Lasers and MEMS injection
 - Light Commands [Sugawara et al., USENIX Security 2020]

Research Vision:

A world where science-based security is built-in by design to all embedded systems:

- medical devices
- healthcare delivery
- autonomous transportation
- manufacturing
- the Internet of Things (IoT)
- Why important now?
 - Consumers need confidence in security and privacy before they can trust innovative medical devices and other emerging technology

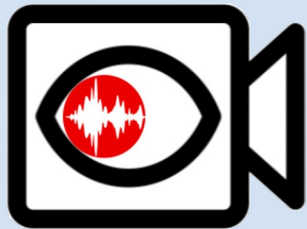
Conclusions

- ✦ **Microprocessors should not blindly trust sensors**
- ✓ **Protect emerging devices with SW that leverages physics**
- ✓ **Focus on trustworthy systems, rather than just secure components**



Seeking PhD applicants in CS, ECE, or bioengineering.
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our values on our website!

Side Eye: Characterizing the Limits of POV Acoustic Eavesdropping from Smartphone Cameras with Rolling Shutters and Movable Lenses



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Pirouz Naghavi², Blas Kojusner²,
Sara Rampazzi², Kevin Butler², Kevin Fu³

¹ University of Michigan, EECS



² University of Florida, CISE



³ Northeastern University, ECE & CS



Northeastern
University

Sound Vibrations

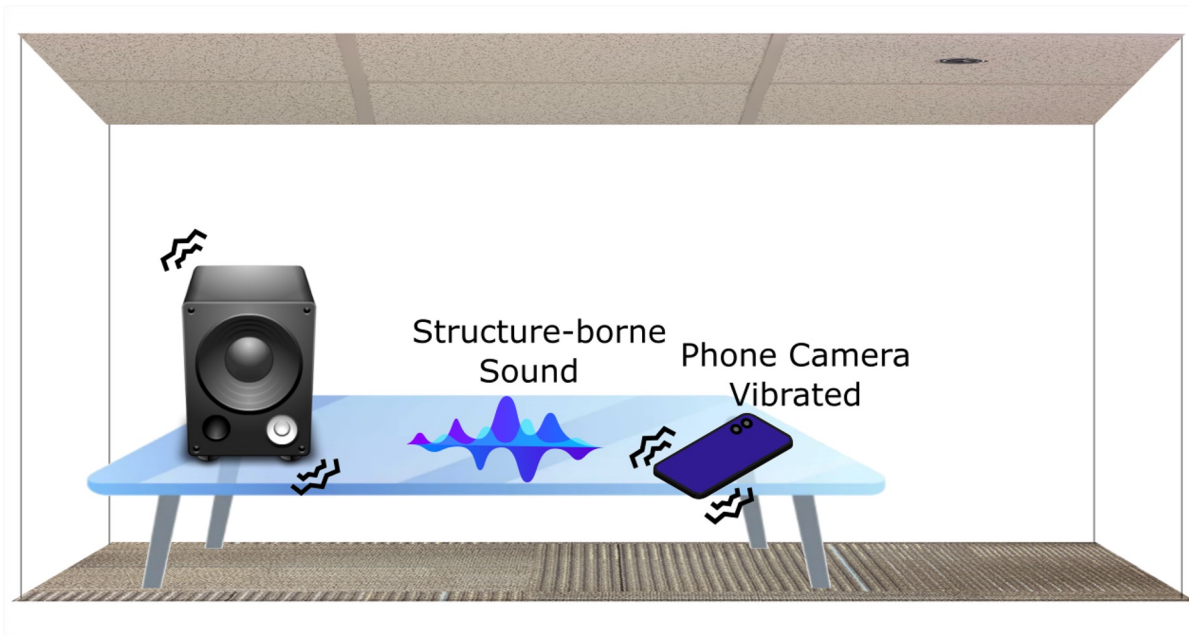
Air-borne



Structure-borne



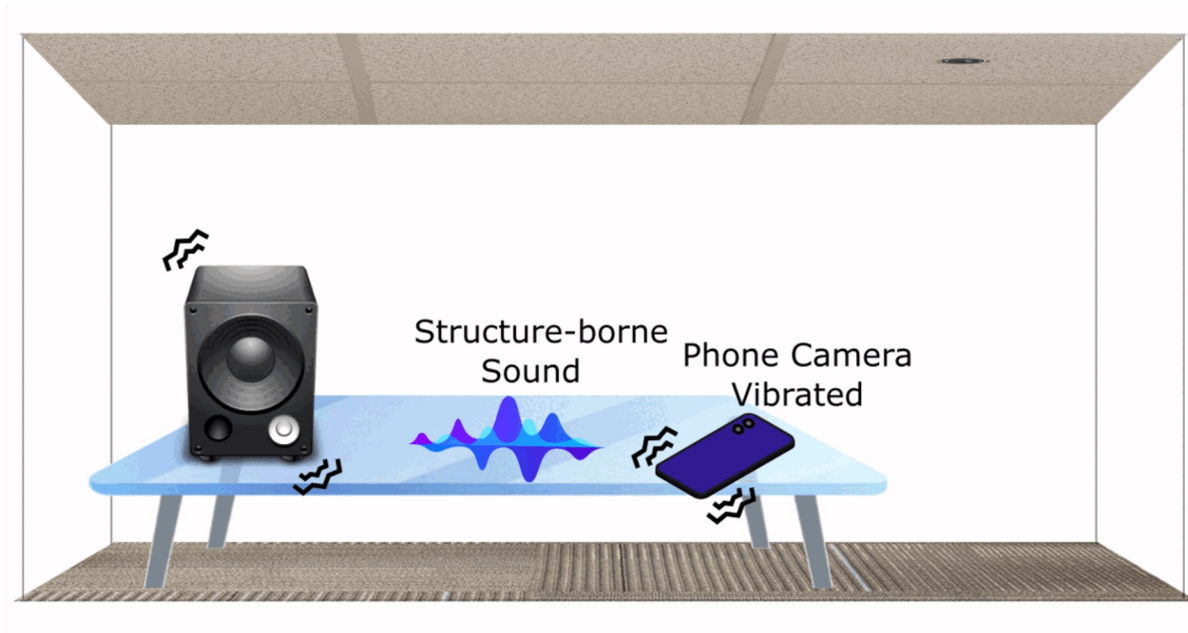
Overview



Adversary:

- Eavesdrop on sound
- Camera access
- No microphone access

Overview



- Acoustic signals leak into muted videos
- 2 Gender (99.67%)
- 20 Speaker (91.28%)
- 10 Spoken digits (80.66%)

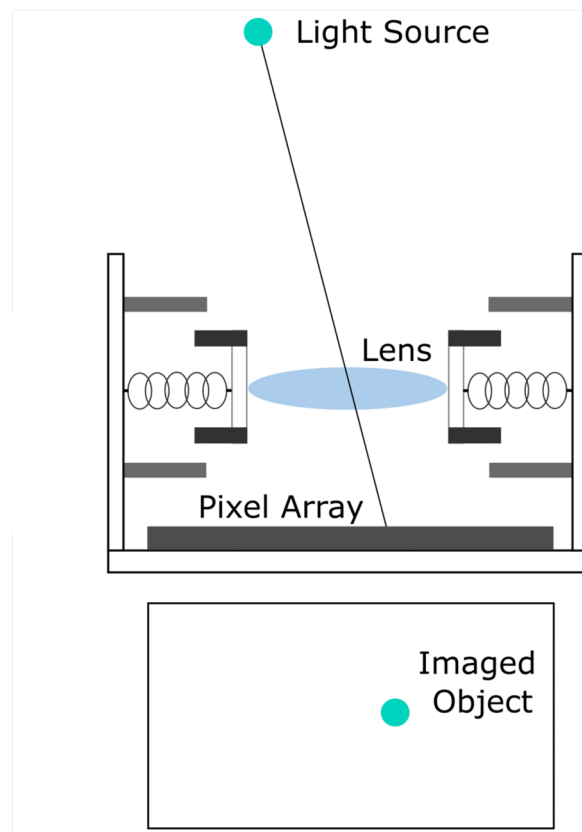
Fundamental Principles

Side Eye: Characterizing the Limits of POV Acoustic Eavesdropping from Smartphone Cameras

Yan Long | yanlong@umich.edu

Camera POV Variations

Camera
Sensor

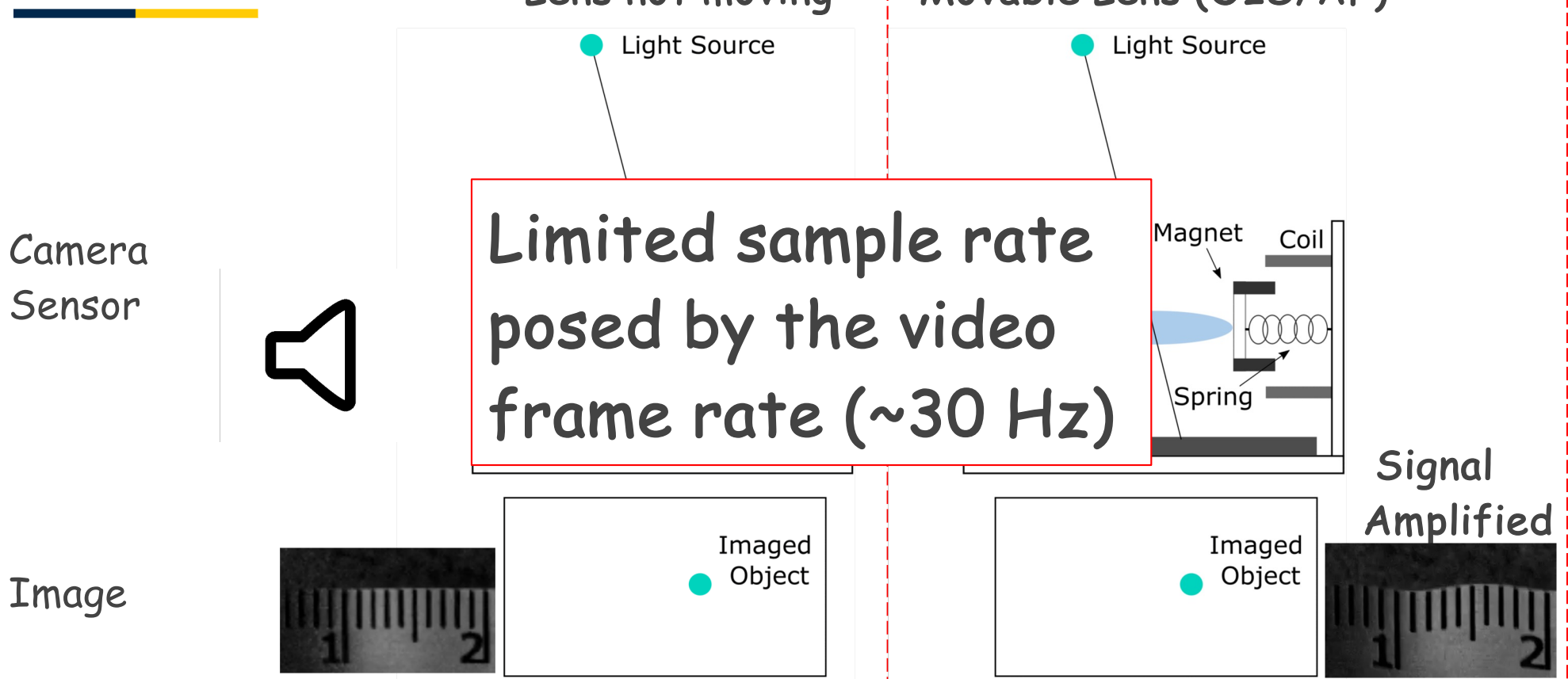


Image

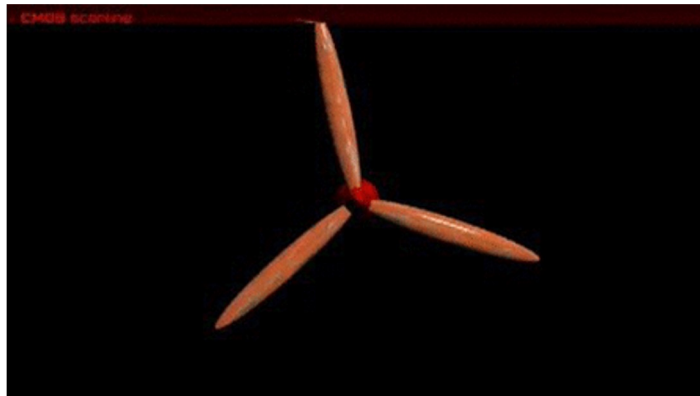
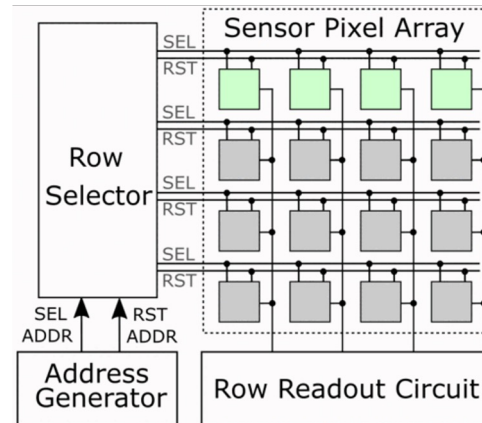
$$f(\text{Sound}) = \text{Image}$$

$$g(\text{Image}) = \text{Sound}$$

Movable Lens



Rolling Shutter



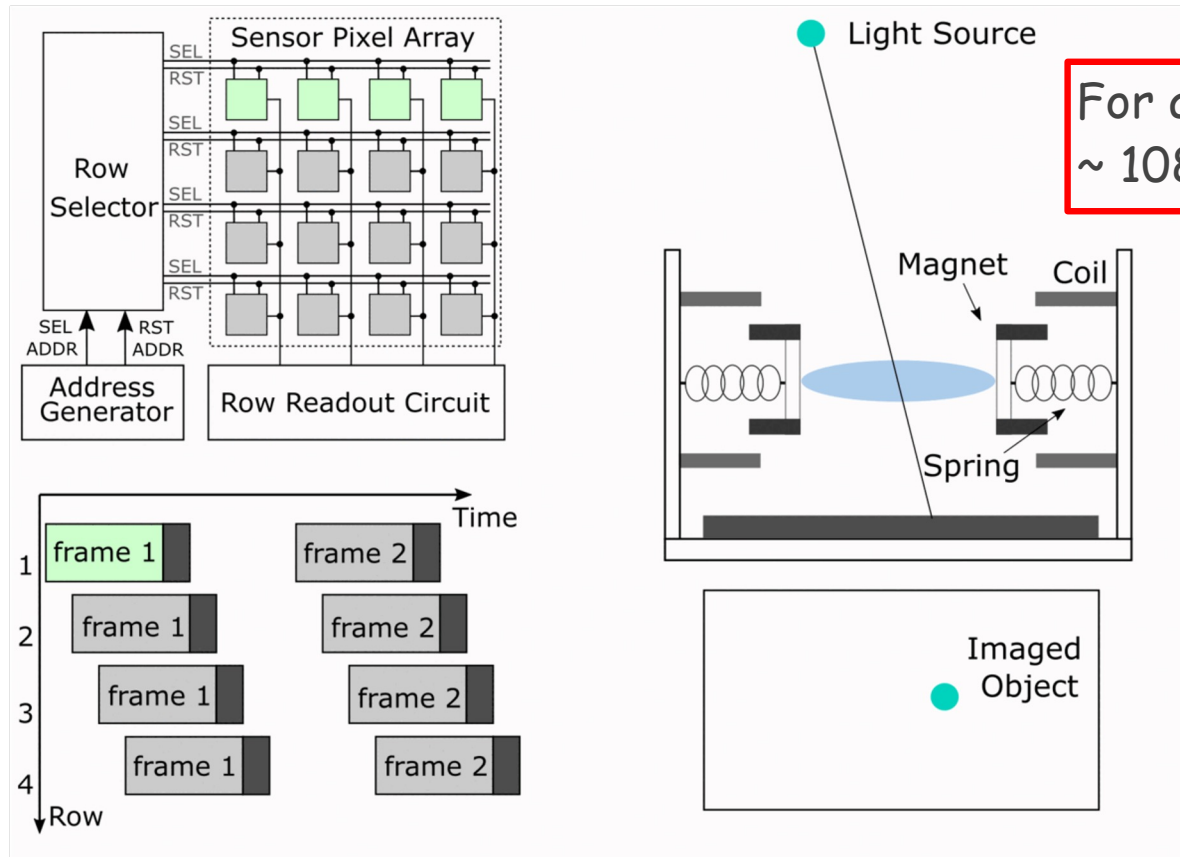
Rotational Motion



[Photo by David Adler]

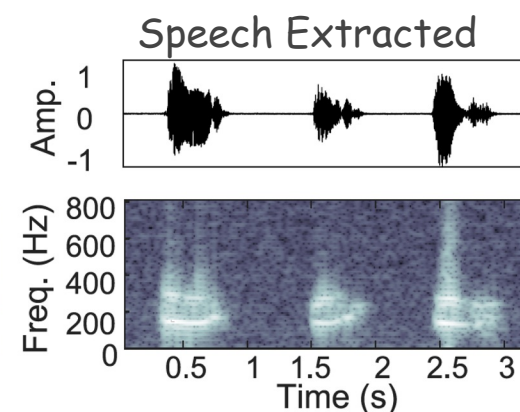
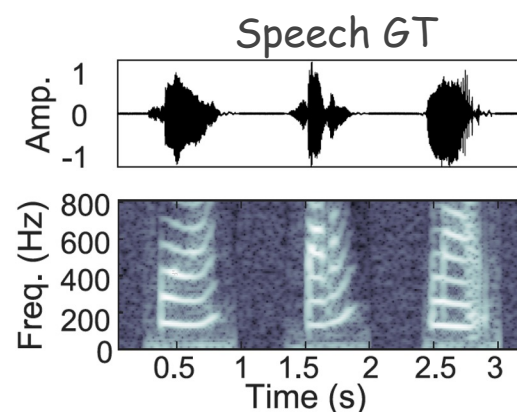
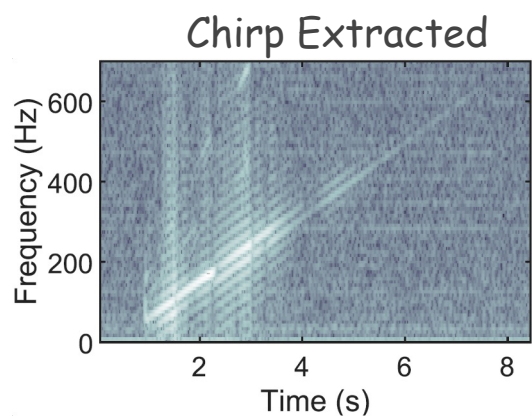
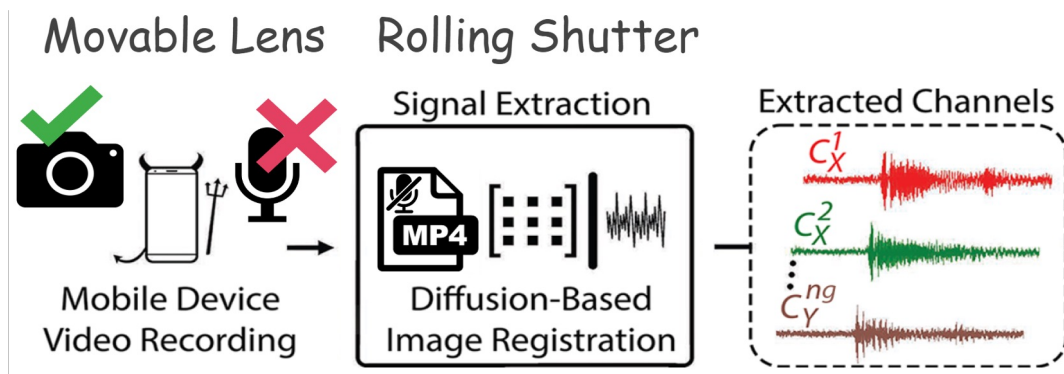
Horizontal Motion

Rolling shutter



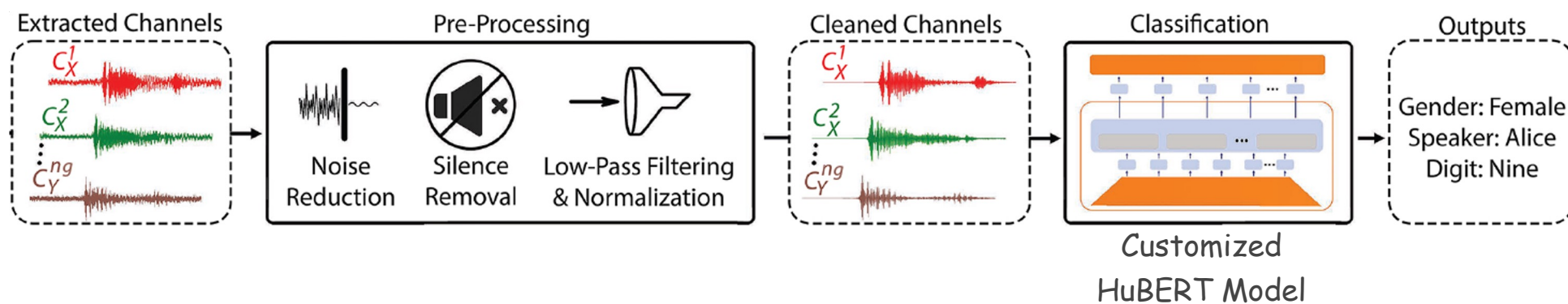
System Design & Results

Attack Mechanism



Audio Samples: <https://sideyeattack.github.io/Website/>

Attack Mechanism



Same-surface Scenarios

(random guess)

Scenario	Case	Avg. SNR	Avg. STOI	(random guess)		
				>50%	>5%	>10%
				G (%)	S (%)	D (%)
Volume	85 dB	18	0.51	99.87	91.02	79.69
	75 dB	11	0.44	99.80	89.13	76.95
	65 dB	4	0.18	98.83	76.11	68.16
	55 dB	2.4	0.13	80.27	34.77	27.67
	45 dB	2.3	0.15	54.49	8.92	13.28
	35 dB	2.3	0.14	54.23	6.84	15.95
Wooden CR TBL, Distance, Volume	10 cm, 85 dB	8.8	0.33	99.02	79.82	66.6
	10 cm, 65 dB	2.4	0.19	76.76	42.58	32.49
	200 cm, 65 dB	2.3	0.19	70.75	33.53	26.43
	300 cm, 65 dB	2.6	0.19	83.2	41.86	30.99



2 genders, 20 speakers, 10 spoken digits
<https://github.com/soerenab/AudioMNIST>

Different-surface Scenarios



Scenario	Avg. SNR	Avg. STOI	>50%	>5%	>10%
			G (%)	S (%)	D (%)
Monitor Stand 85 dB	11	0.45	99.09	80.53	60.42
Monitor Stand 65 dB	2.6	0.09	84.05	42.32	32.1
Two Desks 85 dB	2.6	0.08	75.72	19.6	14.26
Two Rooms 85 dB	2.3	0.06	66.93	15.17	15.17
Shirt Pocket 85 dB	2.5	0.19	95.9	66.37	45.7
Bag Pocket 85 dB	4.1	0.23	93.1	40.1	55.34

Mitigation

- User-based: (1) lower-quality cameras, (2) larger distances, (3) dampening
- Address rolling shutters (RS): (1) higher RS frequency, (2) random-coded RS
- Address movable lens: (1) tougher springs, (2) lens locking

Defense	Gender (%)	Speaker (%)	Digit (%)
None (Baseline)	99.87	91.02	79.69
① Rubber Mat Dampening	98.64	80.11	65.36
② Higher RS Freq. (648 kHz)	93.29	62.89	48.89
③ Random-coded RS	98.18	76.56	60.22
①+②	75.65	43.88	33.14
①+③	72.66	46.03	37.63
④ Tough Spring/Lens Locking	65.23	16.73	16.67
②+④	53.91	8.66	16.73
③+④	54.36	8.46	13.93

2-class 20-class 10-class

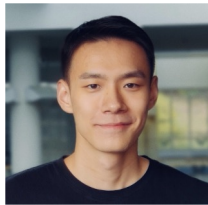
Applications

- Digital forensics on photographs for prosecution or defense
 - Opto-acoustic equivalent to DNA profiling
 - Exonerate by showing absence of opto-acoustic fingerprint
 - Implicate by showing presence of opto-acoustic fingerprint
- Law enforcement and private investigation
 - De-anonymize disguised or muted voices from kidnapping videos and propaganda
 - Determine likely gender and identity associated with voices modulated into pixels
 - Line up of the usual suspects: Determine statistical likelihood of presence or absence of a speaking individual from a recorded off-camera scene
 - Determine probability of phrases spoken behind the camera in dubbed videos (TikTok, IG, etc.)
- Detecting deep fakes or synthetic media in political videos, propaganda, etc.
- Advertising based on posted photographs that contain hints of conversations and individual interests

Summary

- Acoustic signals can leak into muted videos/image streams as POV variations
- Movable lens amplifies signals, rolling shutter increases sampling rates.

Team



Yan Long



Pirouz Naghavi



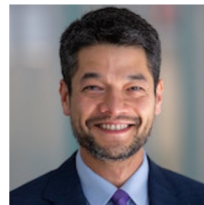
Blas Kojusner



Kevin Butler



Sara Rampazzi



Kevin Fu

<https://sideeyeattack.github.io/Website/>

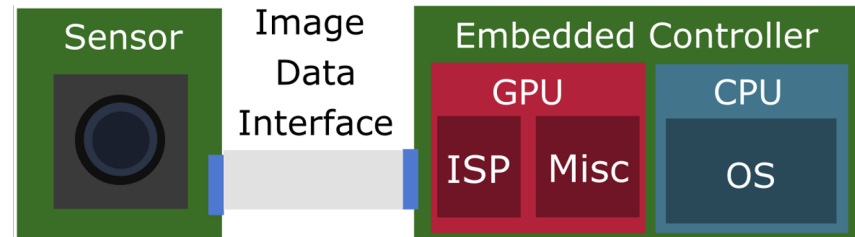
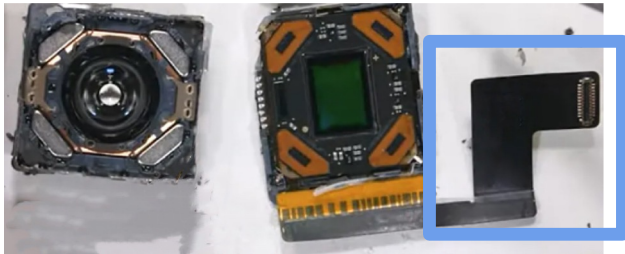


P21_23: EM-Eye: Limiting the Optical-Electromagnetic Side Channel Leakage of Smartphone Cameras

PI: Dr. Kevin Fu, Northeastern University

Student Researchers: Yan Long, Nina Shamsi





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Accelerating MIPI CSI-2 Adoption in Automotive

August 15, 2023 by [Rambus Press](#) — [Leave a Comment](#)

By Joe Rodriguez | Product Marketing Manager, Interface IP

LOW POWER-HIGH PERFORMANCE

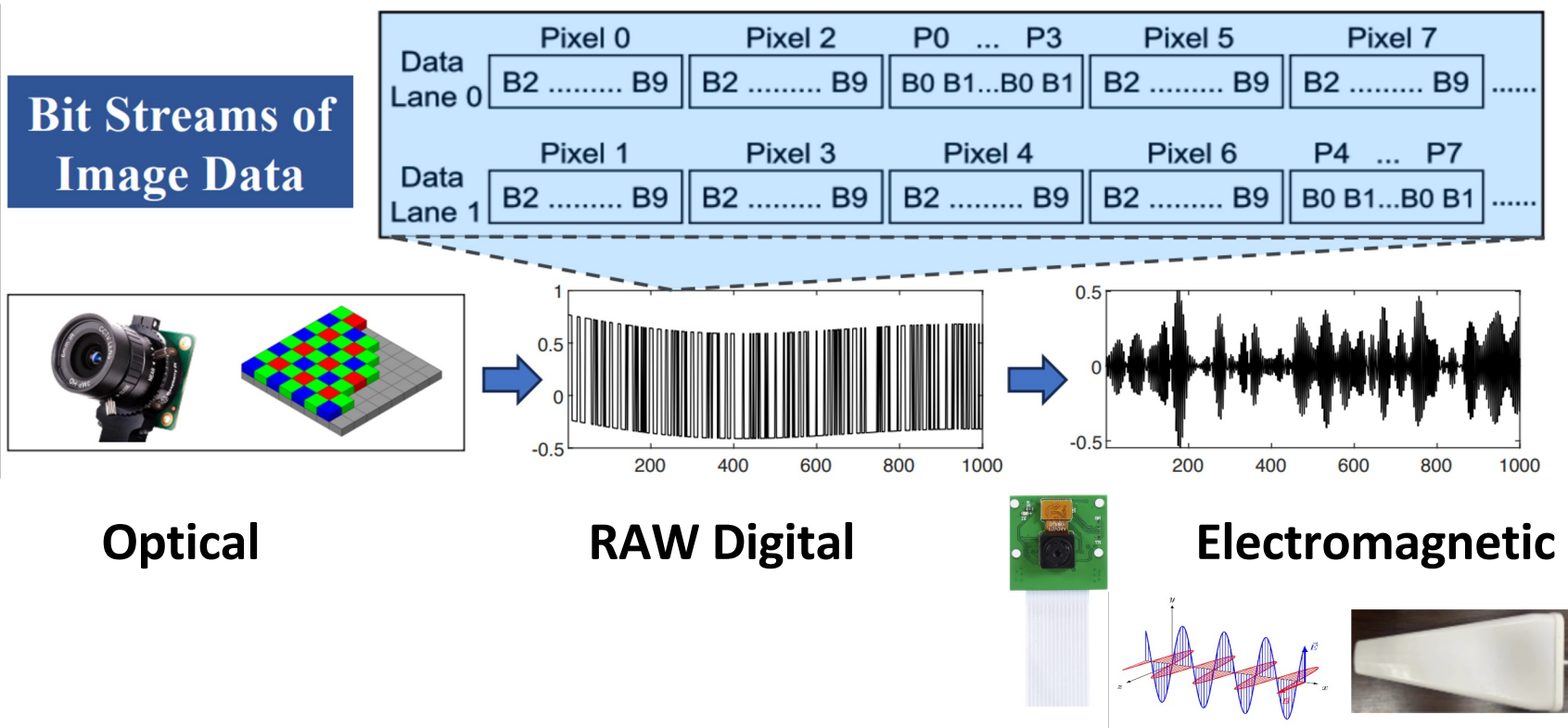
MIPI Standards Gaining Traction In New Markets

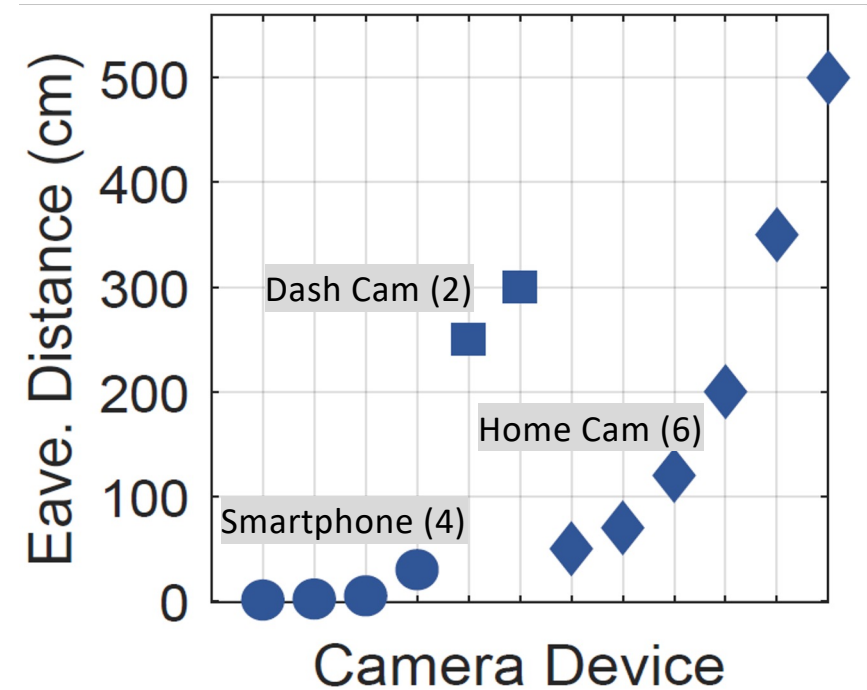
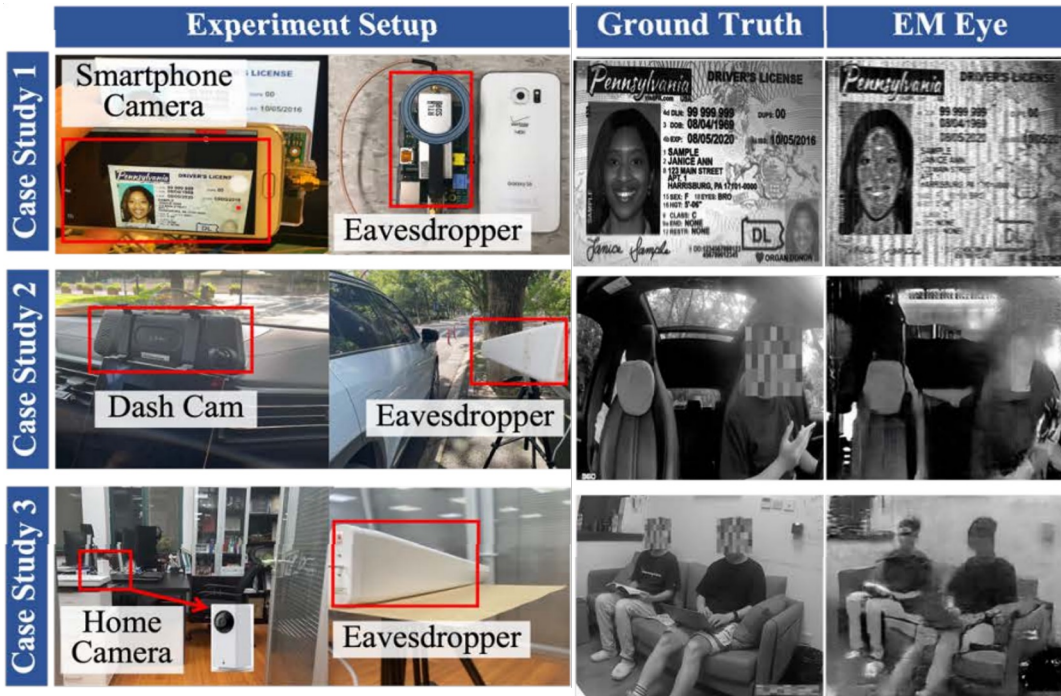
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Convergence of vision and AI is driving adoption of MIPI standards beyond just mobile phones.

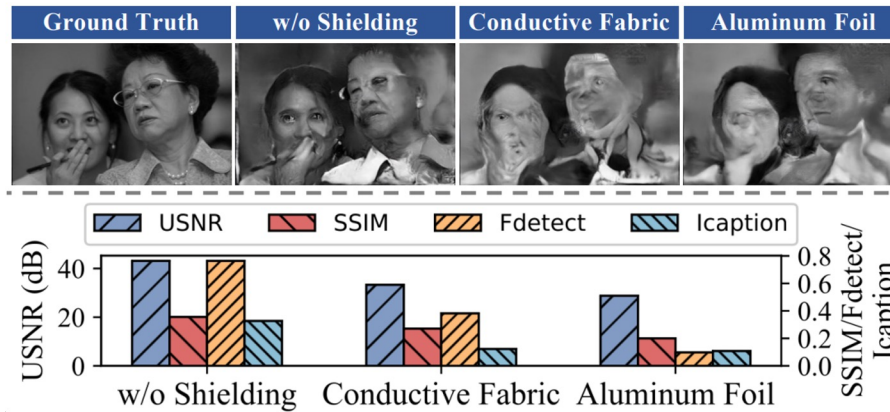
JANUARY 26TH, 2022 - BY: ANN MUTSCHLER



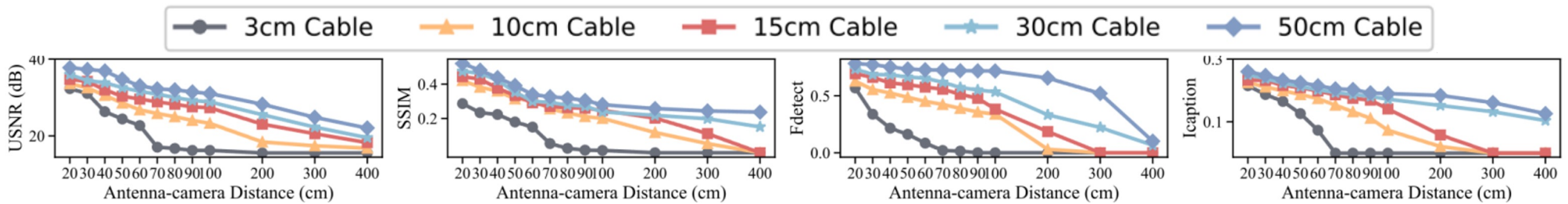




Impact of Shielding Transmission Cables



Impact of Transmission Cable Length and Distance



Prediction of EM Reconstruction

Camera RAW Ground-truth Capture

$$I_{EM}^{[l,h]} = \mathcal{R}_{base} \left\{ z + b_{clk} + \mathcal{F}_{filt} \left[l, h, \mathcal{F}_{data}(I_{GT}) \right] \right\}$$

