

# RSAC<sup>®</sup>Conference2015

San Francisco | April 20-24 | Moscone Center

SESSION ID: MBS-F01

## Side-Channels in the 21st Century: Information Leakage From Smartphones

**Gabi Nakibly, Ph.D.**

National Research & Simulation Center  
Rafael – Advanced Defense Systems Inc.  
gabin@rafael.co.il

**Yan Michalevsky**

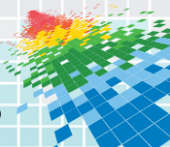
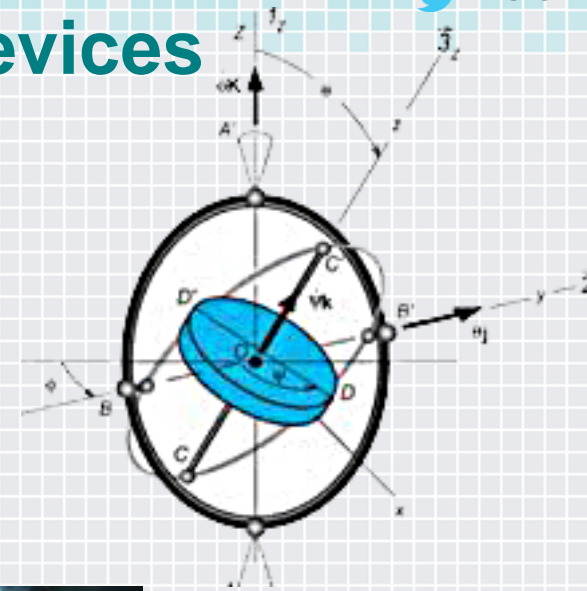
Stanford University  
yanm2@cs.stanford.edu

# CHANGE

Challenge today's security thinking

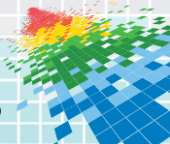


# Side-Channel Attacks on Mobile Devices



# Session's Main Points

- ◆ Mobile devices are susceptible to information leakage in weird and unexpected ways.
- ◆ Rogue applications might do harm even if they have few permissions.
- ◆ The bottom line: treat every app you install as having 'root' on the phone.
  - ◆ After this presentation you will think twice before installing a “harmless” game from an unofficial market having “zero” permissions.

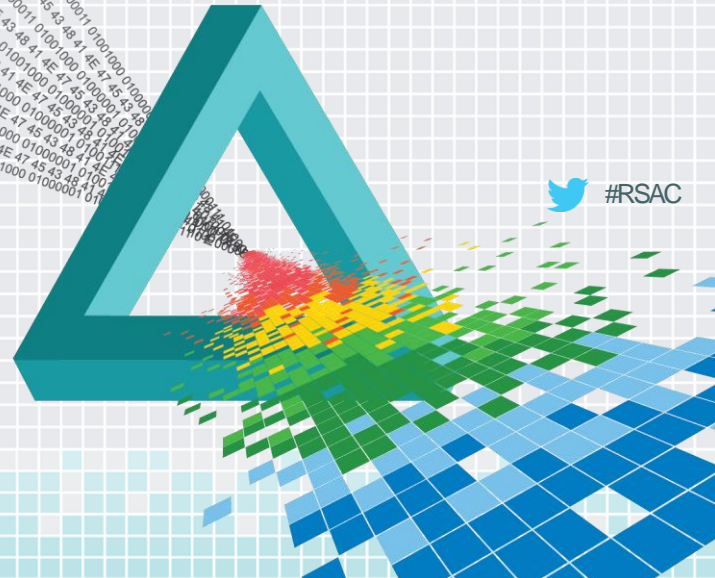
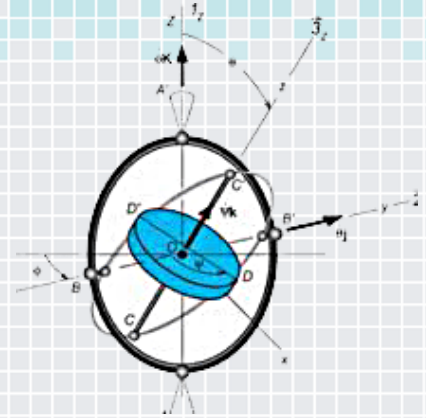


# RSA<sup>®</sup>Conference2015

San Francisco | April 20-24 | Moscone Center

## Sensor ID: Mobile Device Identification via Sensor Fingerprinting

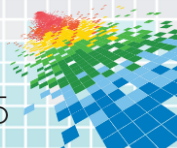
H. Bojinov, Y. Michalevsky, G. Nakibly and D. Boneh



 #RSAC

# Mobile device identification

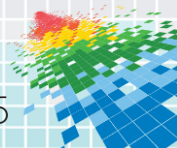
- ◆ The research question: Can an app (or a website) identify the device on which it runs?
- ◆ Answer: Yes!
  - ◆ Android: Device ID, Serial number, MAC Address, ANDROID ID.
  - ◆ iOS :UDID, identifierForVendor, advertisingIdentifier, MAC Address.





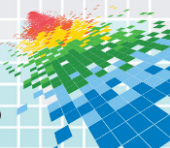
# Mobile device identification (cont.)

- ◆ However, all of these standard identifiers either
  - ◆ require the user's permission
  - ◆ can be changed by the user
  - ◆ does not survive factory reset
  - ◆ not good for all mobile device types
  - ◆ can not be used by a web application



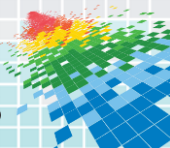
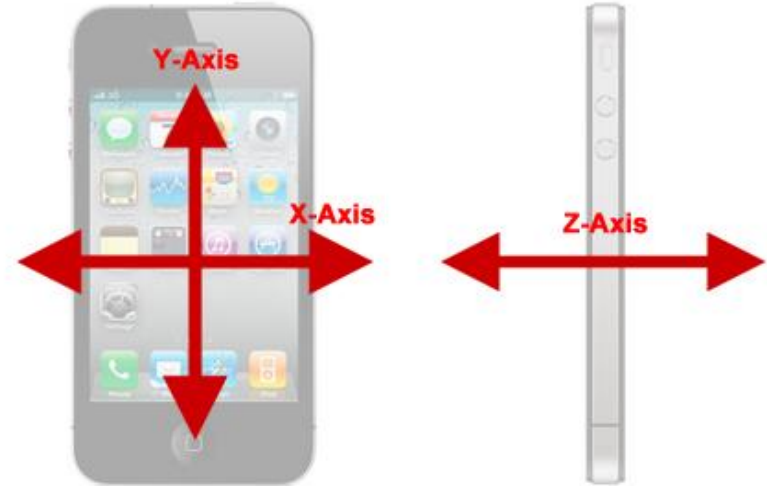
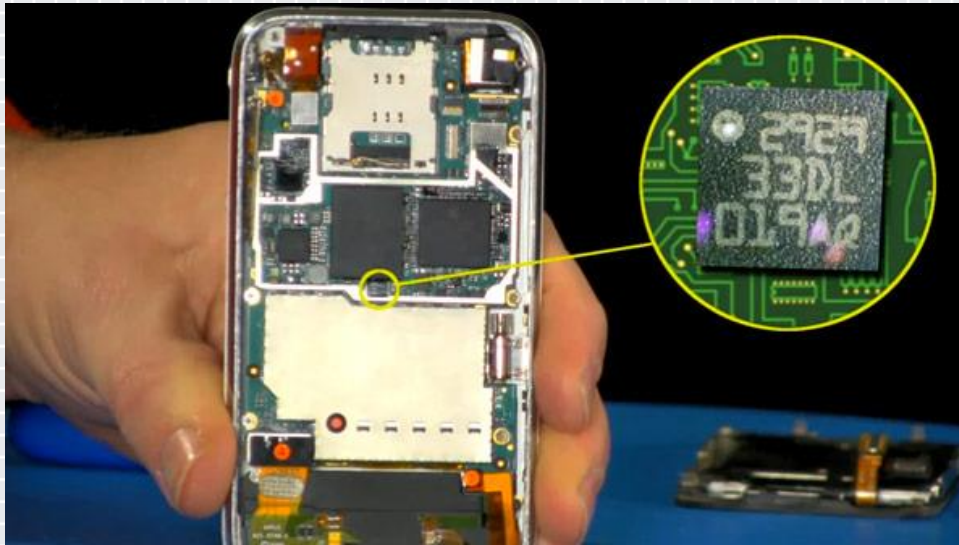
# The Basic Idea

- ◆ Each sensor has a tiny inaccuracy that is very specific to it.
- ◆ Such inaccuracies can be used to fingerprint the device.



# Accelerometer

- ◆ Measures the acceleration of the phone in all three directions.

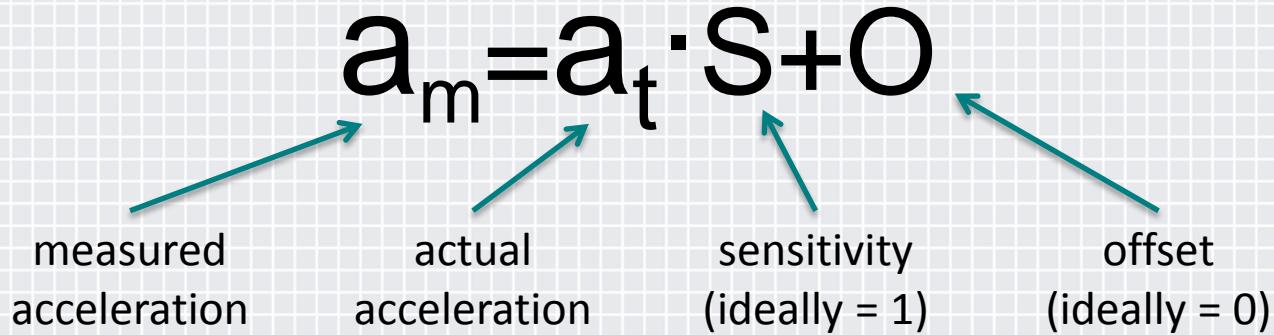




# Accelerometer Skew

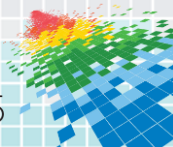
$$a_m = a_t \cdot S + O$$

measured acceleration      actual acceleration      sensitivity (ideally = 1)      offset (ideally = 0)



# But how can we measure S and O?

- ◆ We need some reference acceleration...

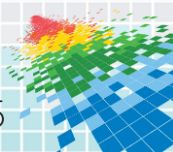
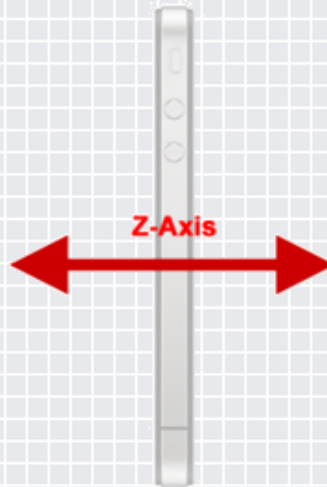


# GRAVITY



# Measuring S and O

- ◆ As a first step we tried to identify S and O for the Z axis



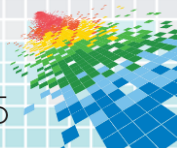
# Measuring S and O

- ◆ Measure the acceleration face up and then face down and then do some calculations



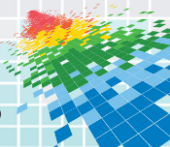
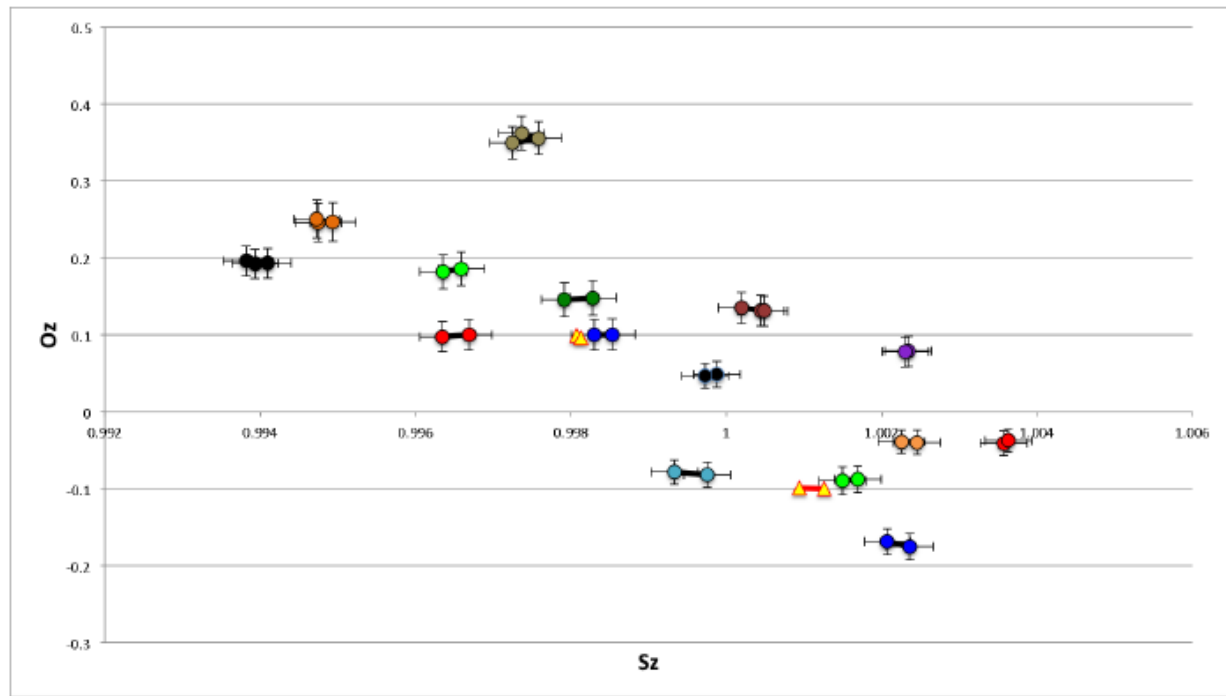
$$S_z = (z_{m^+} - z_{m^-}) / 2g$$

$$O_z = (z_{m^+} + z_{m^-}) / 2$$





# Initial Experiment for 17 iPhones

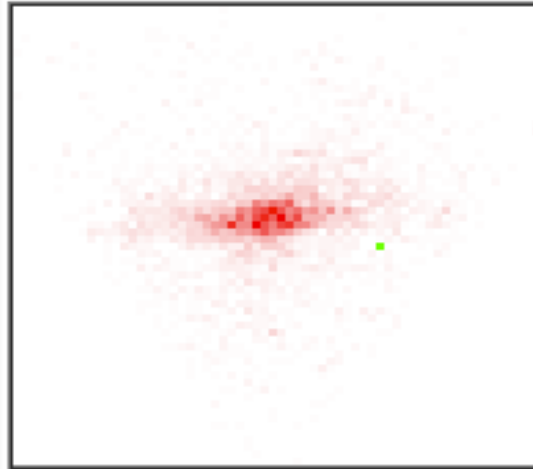


# Results for 10,000(!) phones

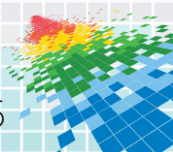
- ◆ An estimated **7.5 bits** of identification.
- ◆ If we can measure S and O for all three axes we can get  $3 \times 7.5 = \mathbf{22.5 \text{ bits}}$  of identification.

## Sensor ID Result Chart

your device ID is **(0.341178,1.007)** and it is unique among **17749** records

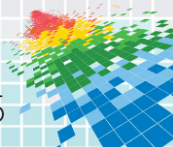


the green square marks your device's ID  
more IDs in a cell make that cell more red



# Measuring S and O for all axes

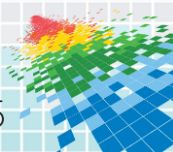
- ◆ A phone does not usually stand up...
- ◆ Alternatively, we can measure the phone in 6 resting positions.



# Measuring S and O for all axes

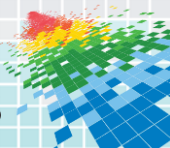
- ◆ And then do some math....

$$\left(\frac{x_m - O_x}{S_x}\right)^2 + \left(\frac{y_m - O_y}{S_y}\right)^2 + \left(\frac{z_m - O_z}{S_z}\right)^2 = g^2$$



# Accelerometer is not alone...

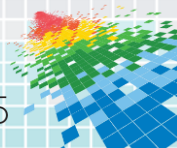
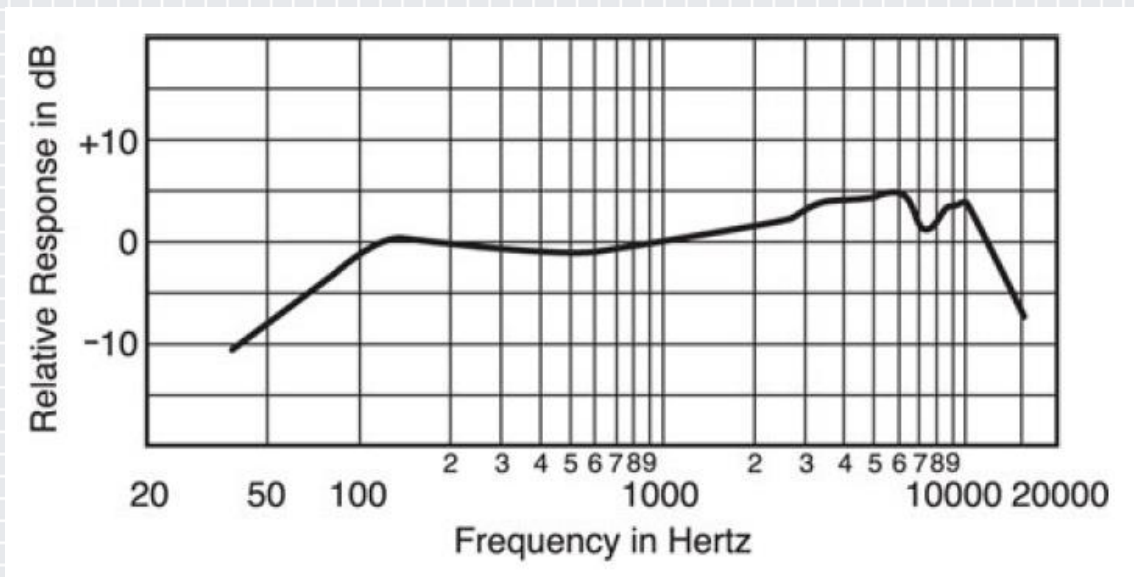
- ◆ Other sensors can also be fingerprinted
- ◆ For example, the microphone





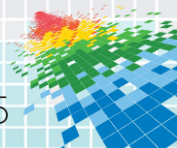
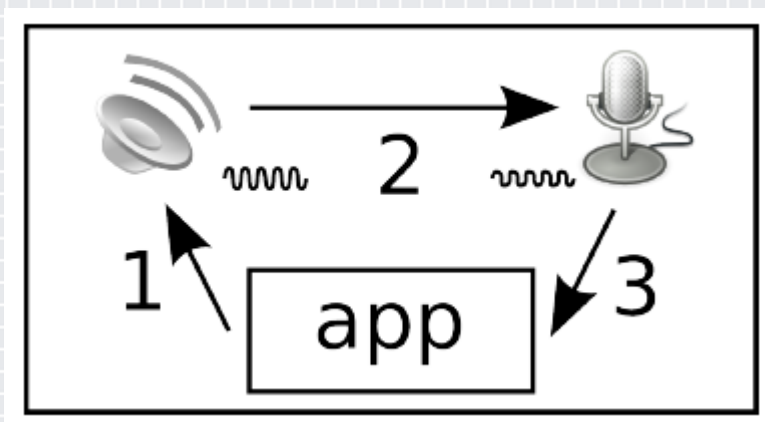
# Microphone

- ◆ Each microphone has a characteristic frequency response curve

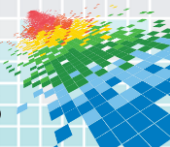
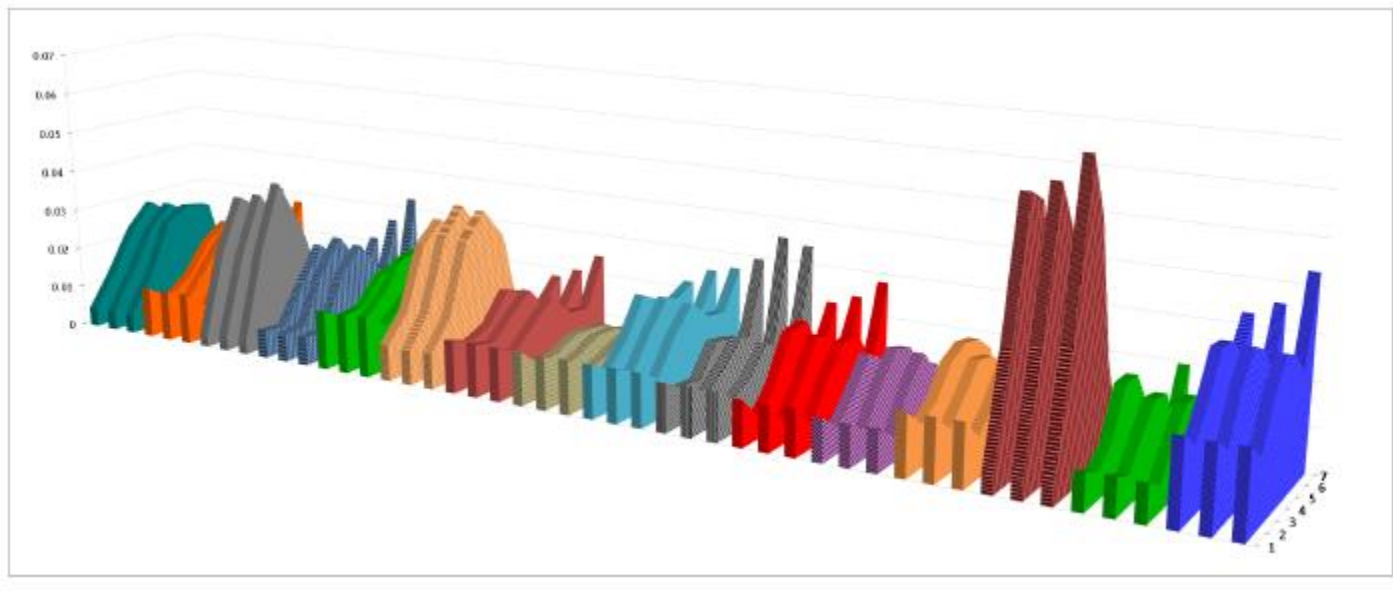


# How can we fingerprint a microphone?

- ◆ We need some audio reference....
- ◆ We can use ....the phone's speaker

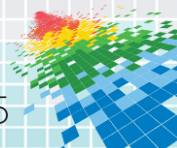


# Experiment for 16 Motorola Droids



# SendorID – Conclusions

- ◆ We have found ways to construct a device ID by sensor fingerprinting.
- ◆ All the sensors' fingerprints may sum up to enough bits to identify all devices.
- ◆ It is hardware dependent.
- ◆ It can be used by web application.

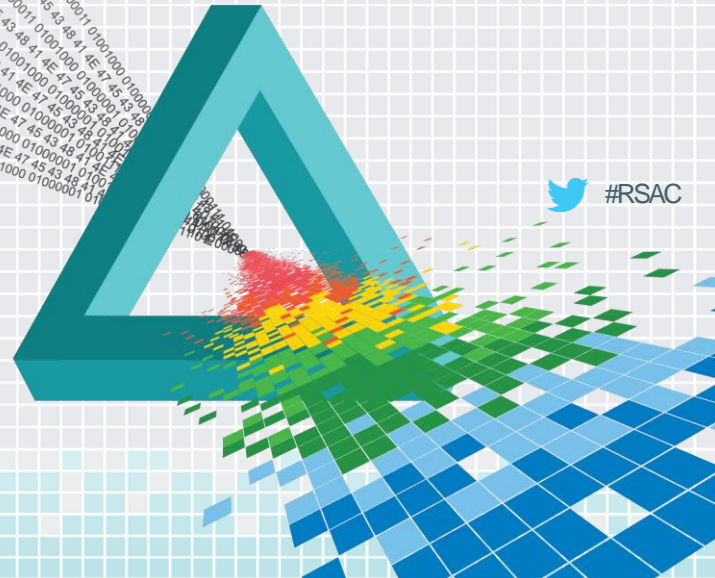
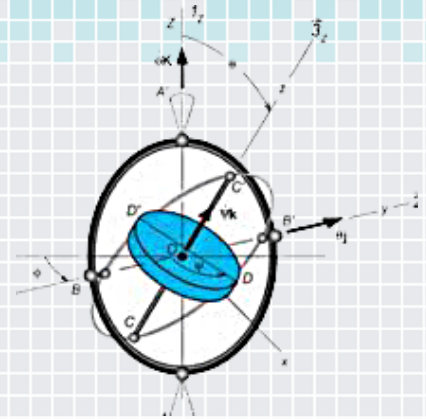


# RSA<sup>®</sup>Conference2015

San Francisco | April 20-24 | Moscone Center

## Gyrophone: Recognizing Speech from Gyroscope Signals

Y. Michalevsky, G. Nakibly and D. Boneh

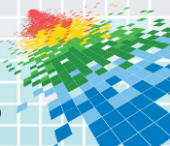
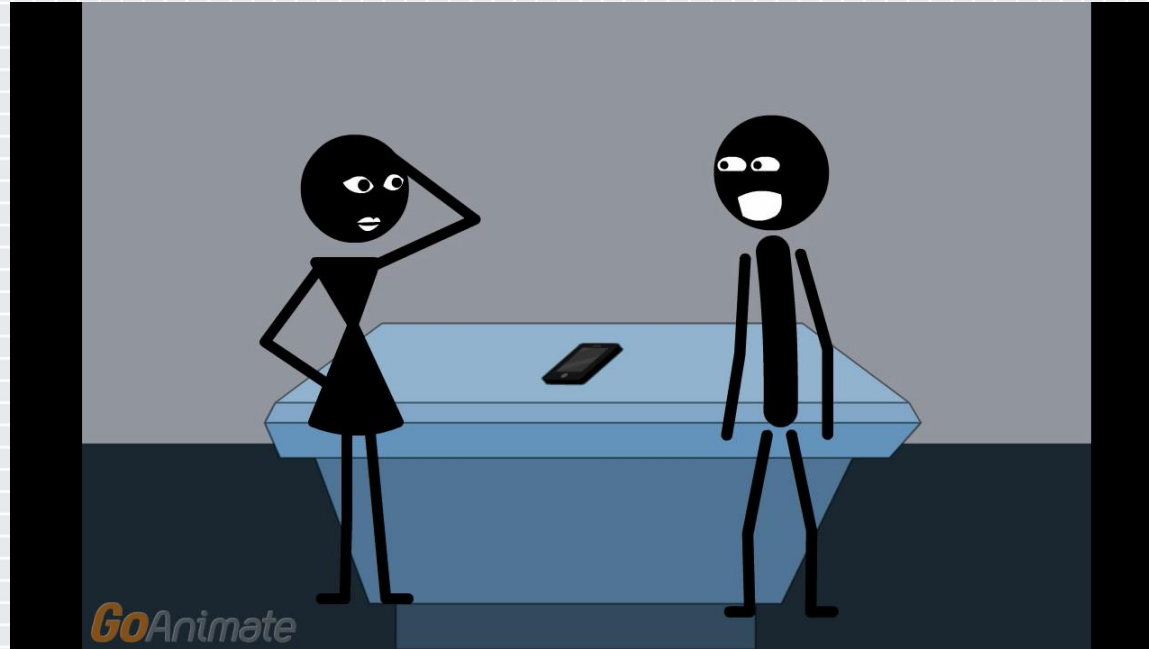


 #RSAC



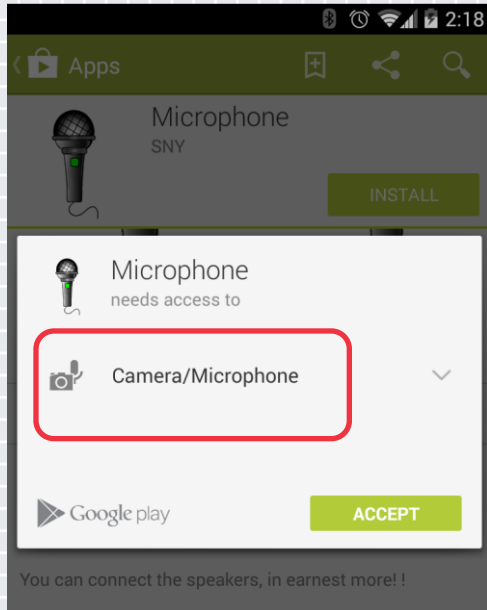
## Scenario

People are talking in the vicinity of a mobile device

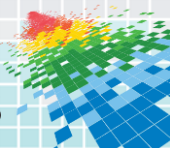
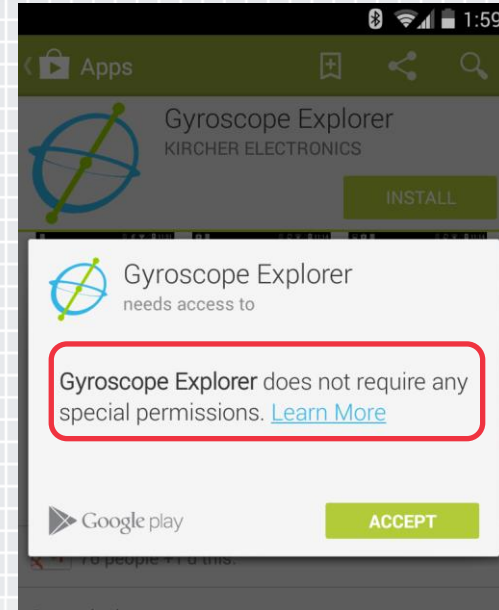


# Microphone vs. Gyroscope Access

Requires permission

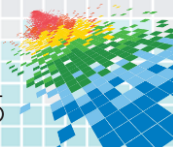
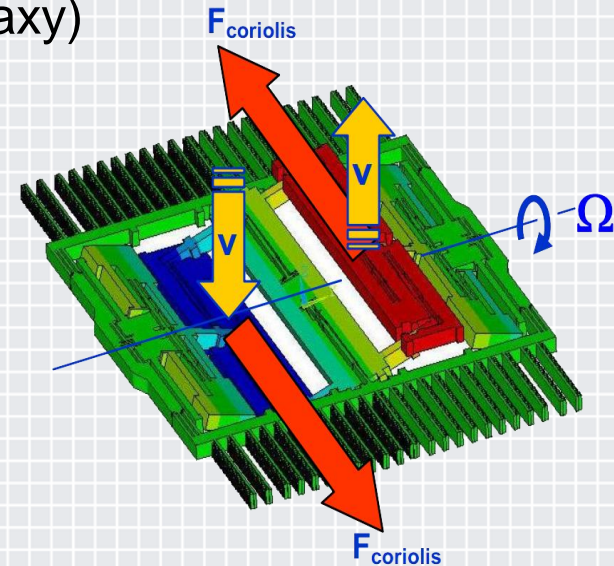


Does **NOT** require permission



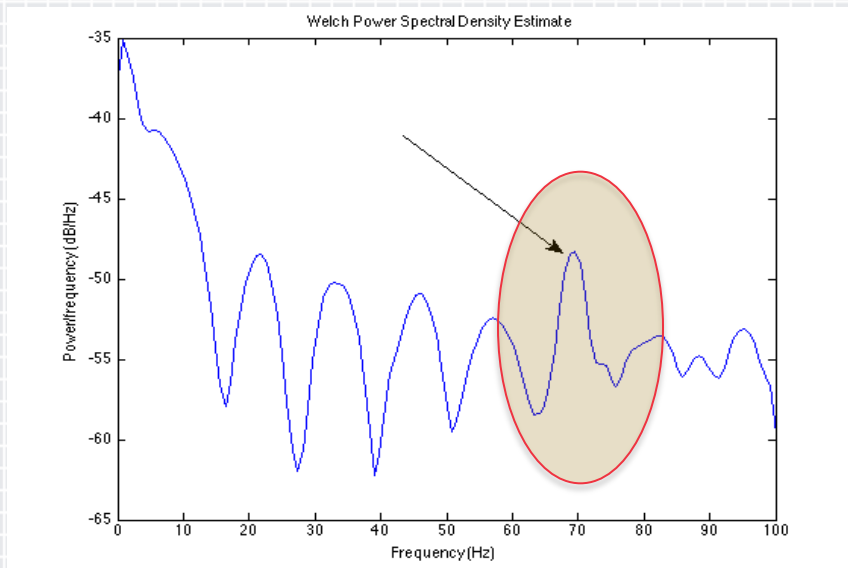
# MEMS Gyroscopes

- ◆ Major Vendors:
  - ◆ STMicroelectronics (Samsung Galaxy)
  - ◆ InvenSense (Google Nexus)

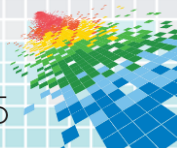
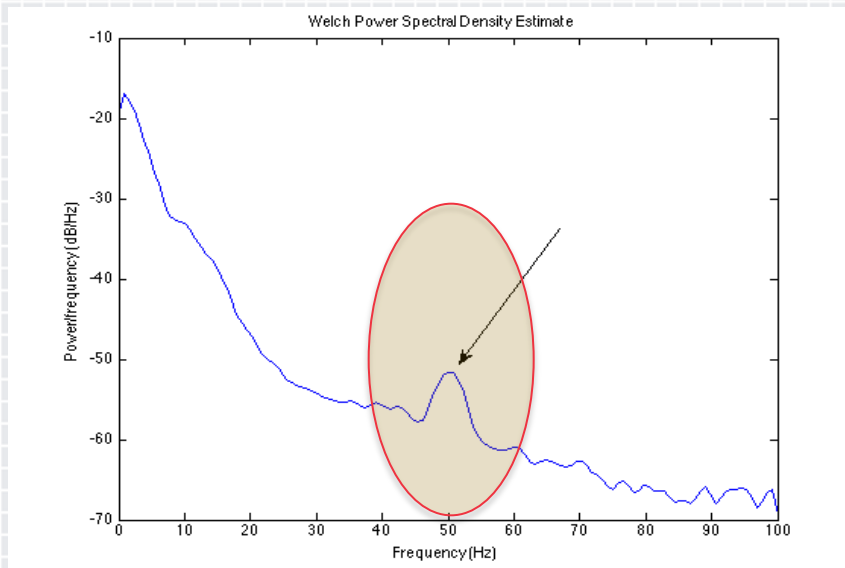


# Gyroscopes are susceptible to sound

## 70 Hz tone PSD

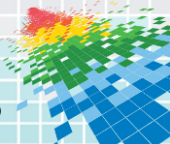


## 50 Hz tone PSD



# Gyroscopes are (lousy, but still) microphones

- ◆ Hardware sampling frequency:
  - ◆ InvenSense: up to 8000 Hz
  - ◆ STM Microelectronics: 800 Hz
- ◆ Software sampling frequency:
  - ◆ Android: 200 Hz
  - ◆ iOS: 100 Hz
- ◆ Very low Signal-to-Noise ratio (SNR)
- ◆ Acoustic sensitivity threshold: ~70 dB  
Comparable to a loud conversation
- ◆ Sensitive to sound angle of arrival
- ◆ Directional microphone (due to 3 axes)

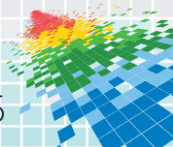




# Browsers allow gyroscope access too

WebApp based browsers

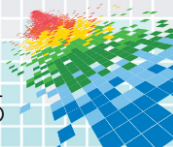
		Sampling Freq. [Hz]
Android 4.4	application	200
	Chrome	25
	Firefox	200
	Opera	20
iOS 7	application	100
	Safari	20
	Chrome	20



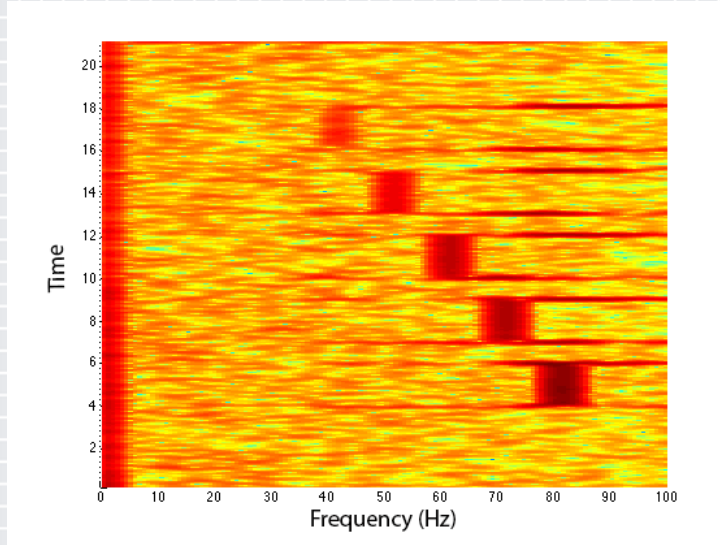
# Problem: How do we look into higher frequencies?

## Speech Range

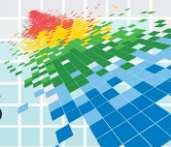
Adult Male	85 – 180 Hz
Adult Female	165 – 255 HZ



# We can sense higher frequencies signals Due to aliasing

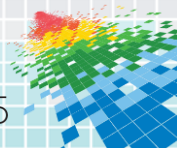


Recording tones between 120 to 160 Hz on a Nexus 7 device



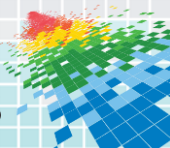
# Experimental setup

- ◆ Room. Simple Speakers. Smartphone.
- ◆ Subset of TIDIGITS corpus
- ◆ 10 speakers  $\times$  11 samples  $\times$  2 pronunciations = 220 total samples



# Speech analysis using a single Gyroscope

- ◆ Gender identification
- ◆ Speaker identification
- ◆ Isolated word recognition
  - ◆ Speaker independent
  - ◆ Speaker dependent

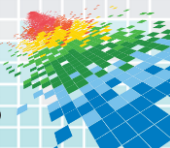


# We can successfully identify gender



Nexus 4	84%
Galaxy S3	82%

Random guess probability is 50%



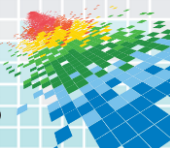


# A good chance to identify the speaker



Nexus 4	Mixed Female/Male	50%
	Female speakers	45%
	Male speakers	65%

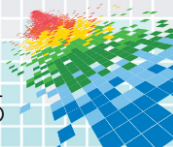
Random guess probability is 20% for one gender, and 10% for a mixed set



# Isolated word recognition (speaker independent)

Nexus 4	Mixed Female/Male	17%
	Female speakers	26%
	Male speakers	23%

Random guess probability is 9%



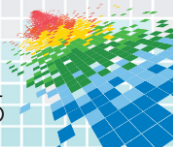
# Isolated word recognition (speaker dependent)

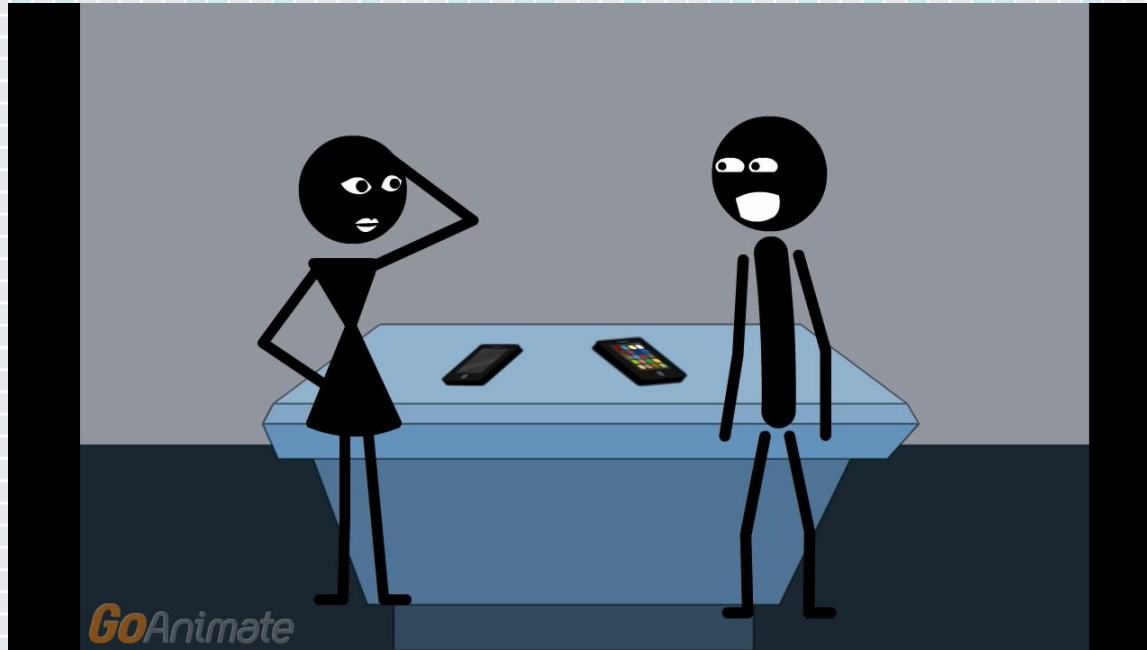
Nexus 4

Male speaker

65%

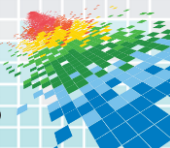
Random guess probability is 9%





## Can we use multiple devices to improve the method?

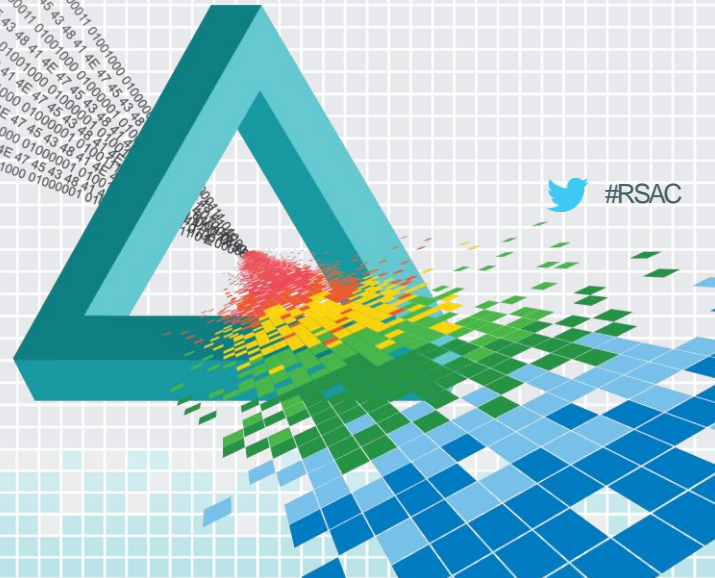
Answer: Yes. Raising speaker dependent recognition rate to 77%.



# RSA<sup>®</sup>Conference2015

San Francisco | April 20-24 | Moscone Center

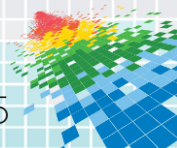
## Defenses



 #RSAC

# Software Defenses

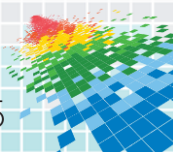
- ◆ Low-pass filter the raw samples
- ◆ 0-20 Hz range might be enough for browser based applications (learning from Web-Kit's example)
- ◆ Access to high sampling rate should require a special permission
- ◆ Can possibly be applied by software providers / open-source community





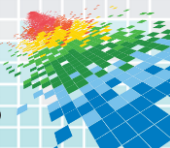
# Hardware Defenses

- ◆ Hardware filtering of sensor signals (not subject to configuration)
- ◆ Acoustic masking
- ◆ Can possibly be applied by vendors



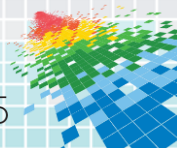
More details can be found here

[crypto.stanford.edu/gyrophone](https://crypto.stanford.edu/gyrophone)



# To conclude

- ◆ We believe this is only the beginning
- ◆ Many more unexpected information leakages will be found in coming years.
- ◆ Treat every app you install as having ‘root’ on the phone!
- ◆ Now we know you will think twice before installing that “harmless” game ....



# Questions?

- ◆ Yan Michalevsky – [yanm2@cs.stanford.edu](mailto:yanm2@cs.stanford.edu)
- ◆ Gabi Nakibly – [gabin@rafael.co.il](mailto:gabin@rafael.co.il)

