

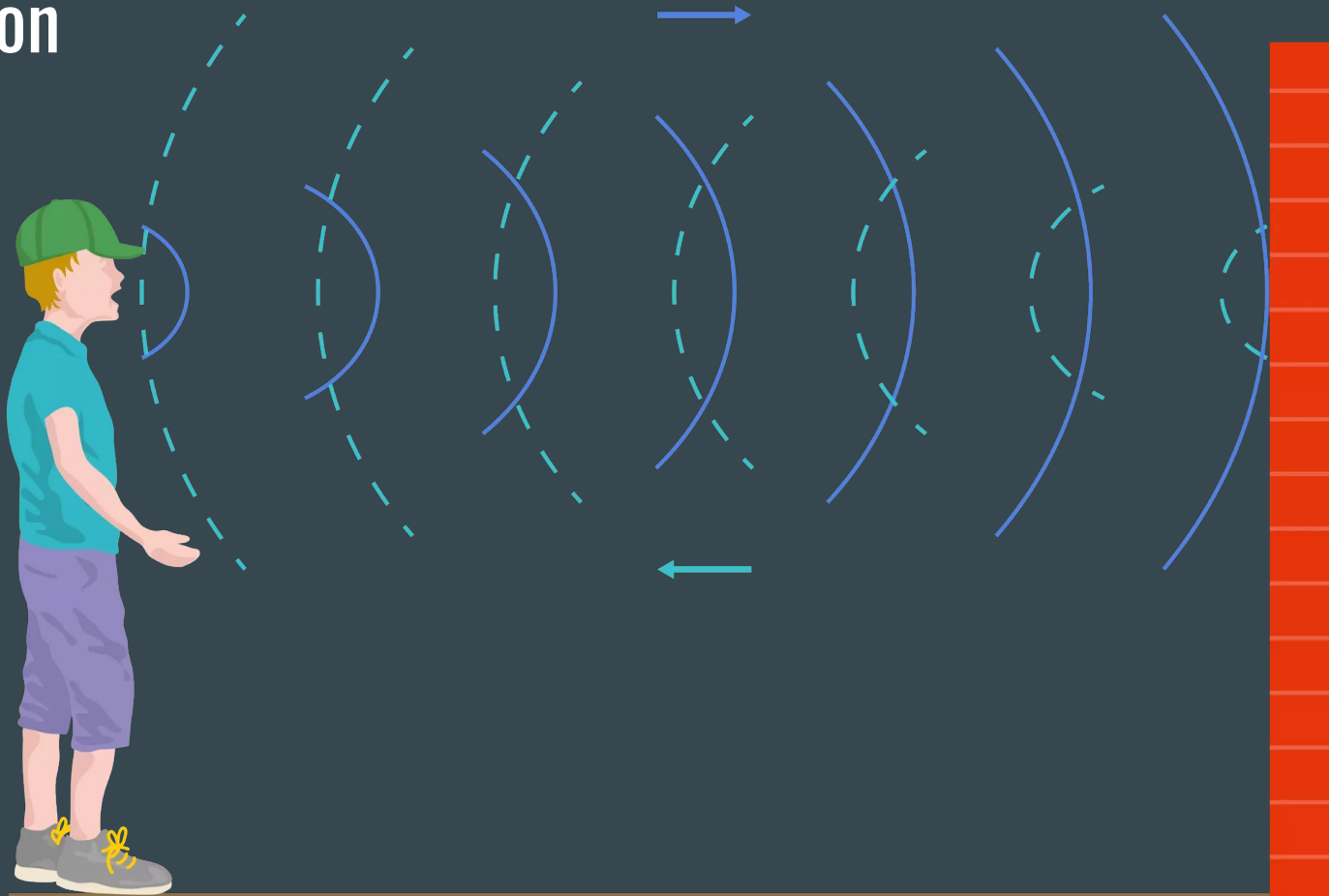
EchoTag: Accurate Infrastructure-Free Indoor Location Tagging with Smartphones



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Intuition

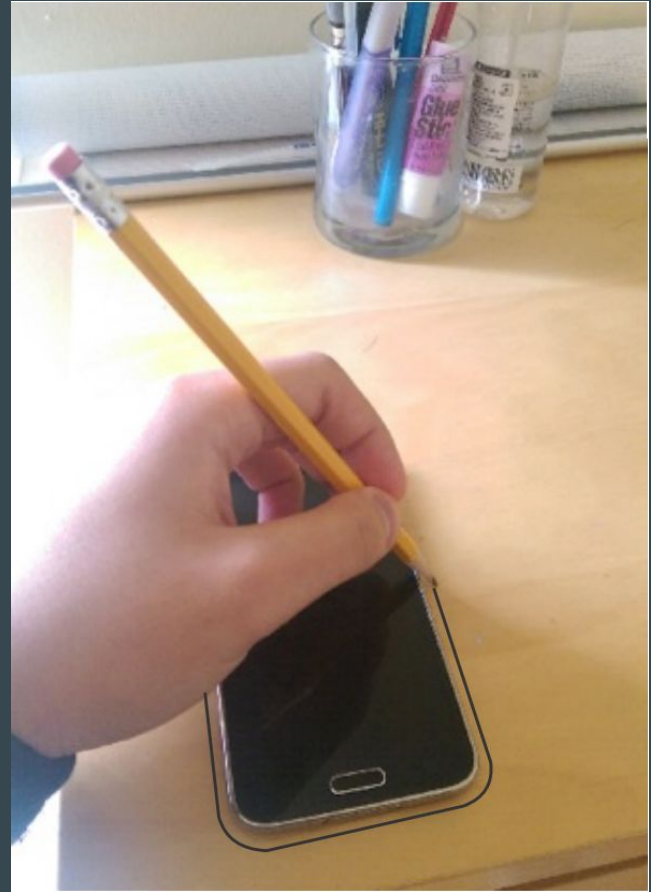


EchoTag

- Uses this intuition to implement indoor location “tagging”
- A tag is a location and associated actions
 - Phone could go to silent in class
 - Phone could automatically set an alarm in the bedroom
- Tags are created when user places phone at specific location
 - Phone “learns” location by emitting an audio signal
- Phones “remember” where they are based on received audio
 - ... and perform pre-defined actions

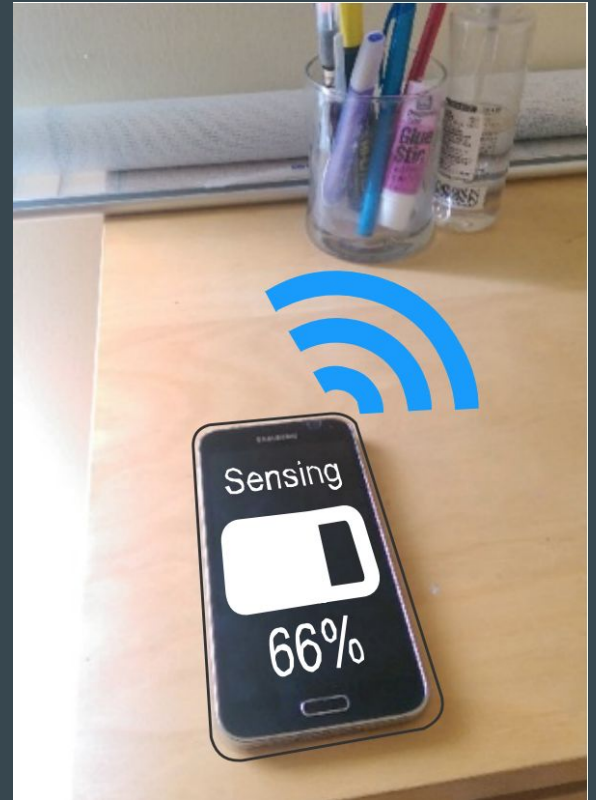
Using EchoTag - Drawing tags

- Help users remember tag location
- Can draw on paper and tape on surfaces



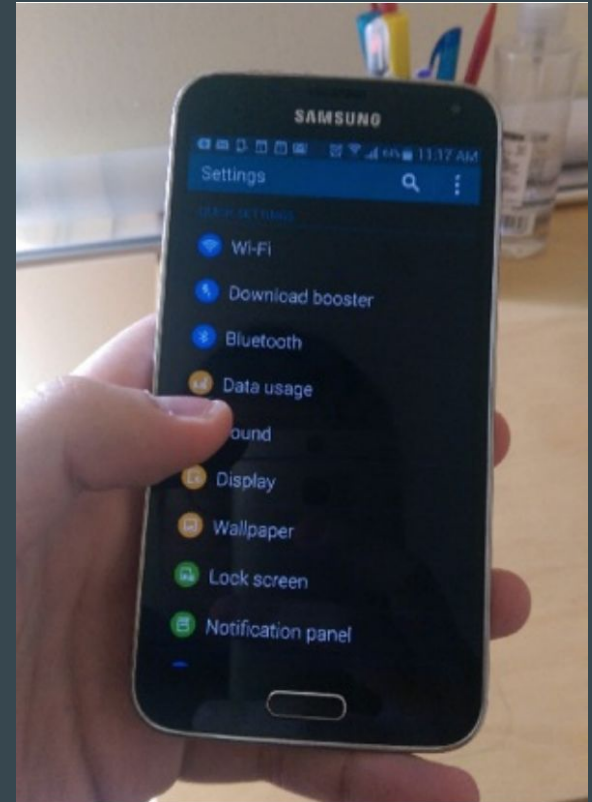
Using EchoTag - Sense acoustic signatures

- Use phone's speaker to play a sound
- Use phone's microphone to record echo



Using EchoTag - Select mapped applications

- Users specify and associate different actions with different tagged locations

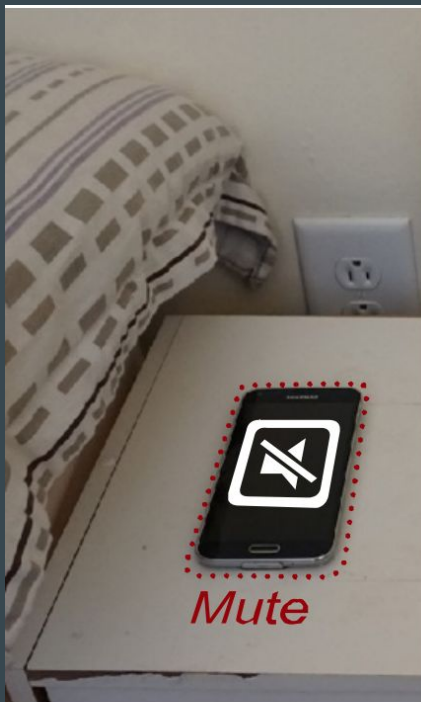


Using EchoTag - Replay recorded tags

- Put phone back to tagged location to replay actions specified by users



EchoTag: Use cases



Mute

a) Auto silent mode



Music

(b) Auto music



Timer

(c) Auto timer

(How well) does it work?

- In a “perfect” environment
 - Can distinguish 11 tags at 1cm resolution with 98% accuracy
 - Maintains 90% accuracy over a week

- Introducing foreign objects to room
 - Reduces accuracy to 56%

(How well) does it work?

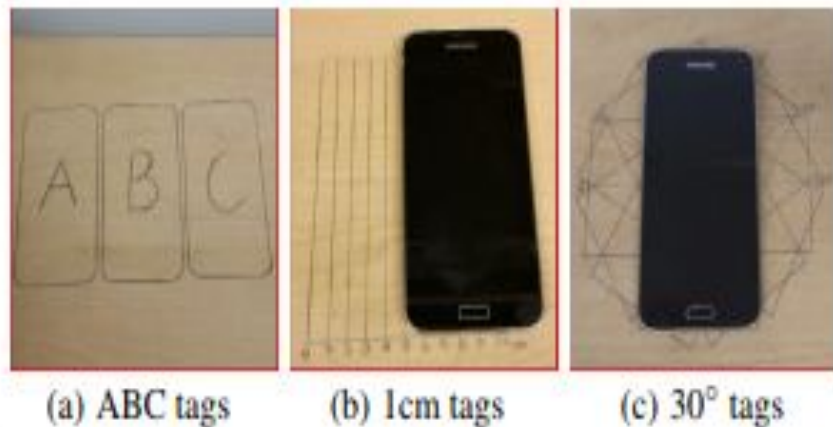


Figure 12—Tag systems. The first tag system consists of disjoint (echo) tags while the second and third tag systems are composed of overlapped tags 1cm or 30° apart.

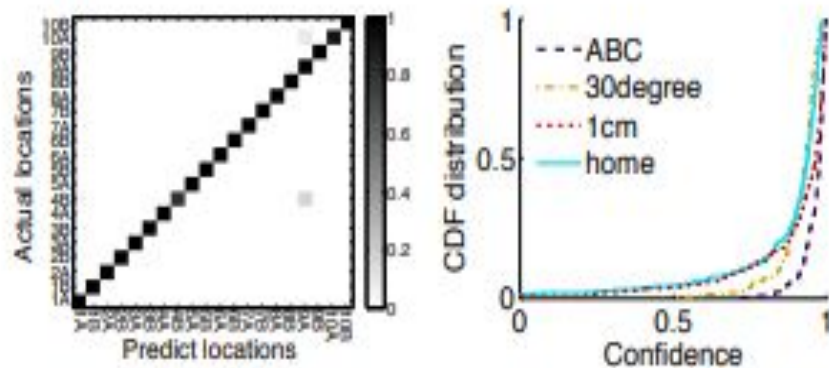


Figure 13—Result of 30min dataset. *Confidence* is defined as the prediction probability at the target location minus the largest prediction probability at the other locations.

Implementing EchoTag

- Create an acoustic signature that is unique to a location
 - Must be robust to “spacial” and temporal changes
- Classification based on signature
 - Uses generic classifiers
- Usability
 - Is EchoTag useful? Obtrusive?

Acoustic Signature

- (Ideally) unique sound features that can identify location
- Training phase; phone emits a signal, records received echo.
- Features extracted from echo, mapped into large dimensional space for classification
- EchoTag uses Frequency Domain signatures (11-22Khz)
 - Depends on differing absorption across frequencies
- Acoustic signature augmented with wifi AP visibility and tilt sensor

Acoustic Signature

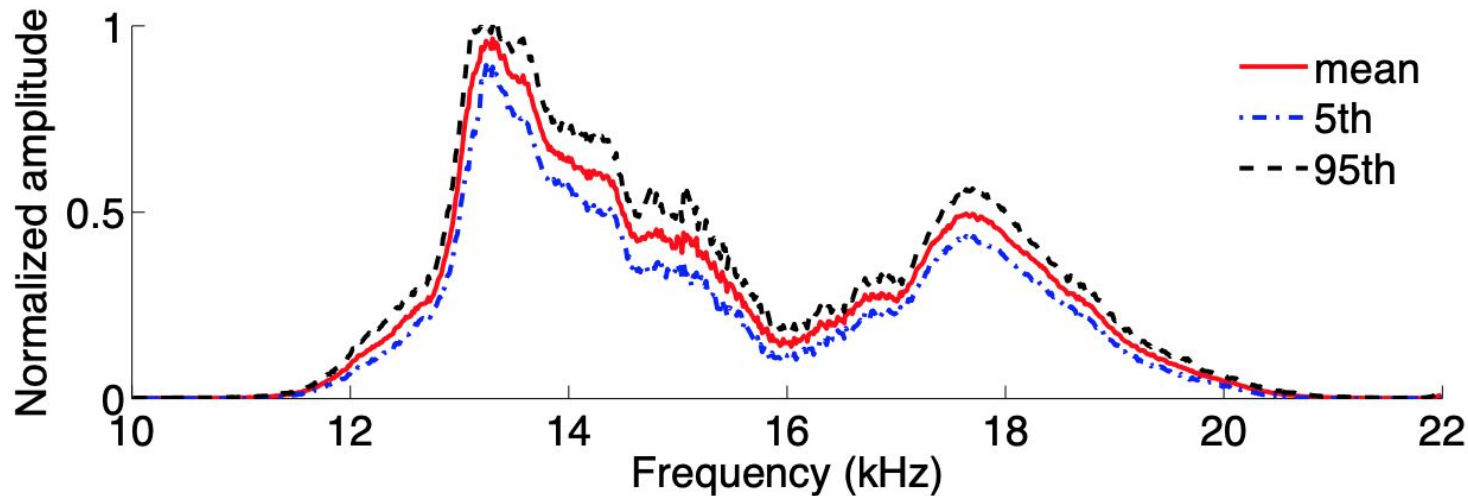
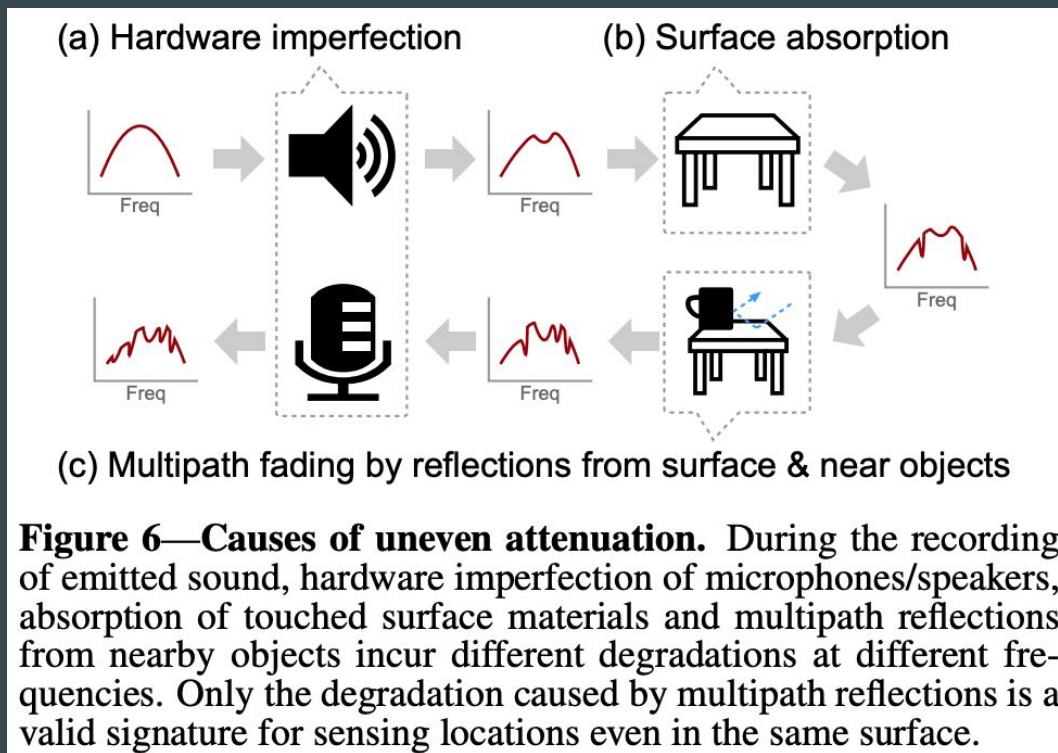


Figure 4—An example of acoustic signatures. The received attenuation of a flat frequency sweep is uneven over different frequencies. The result is an average of 100 trials over 1 minute.

Uneven Attenuation



Acoustic signature generation

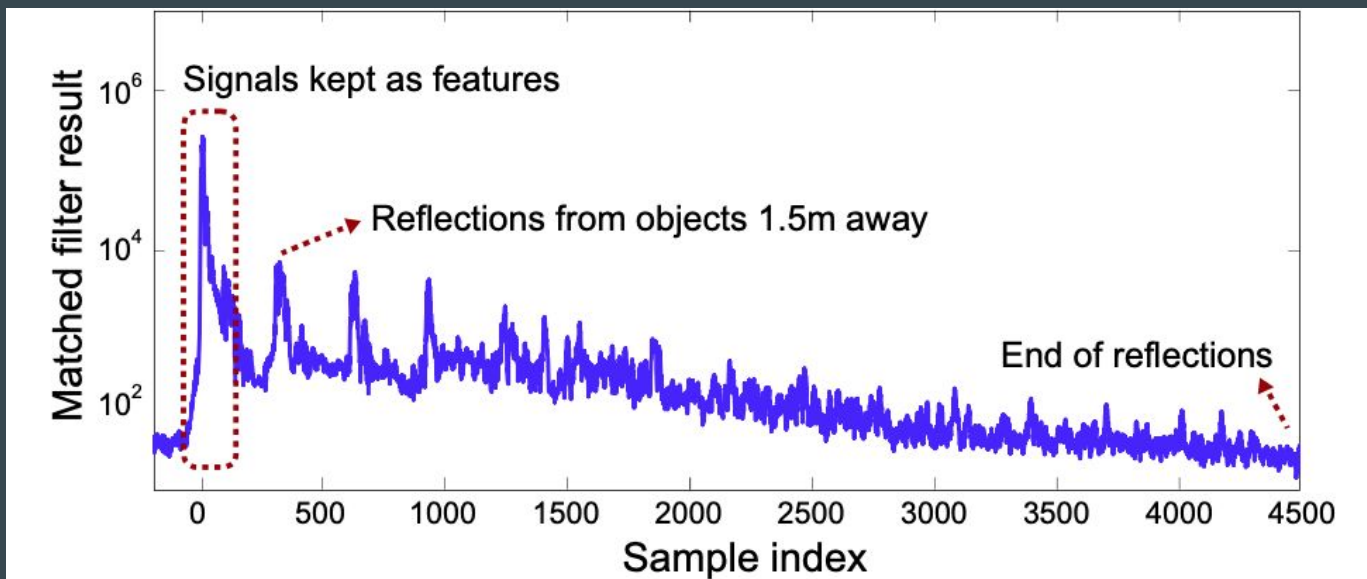


Figure 7—Characteristics of reflections. A matched filter is used to identify the reflections of a 100-sample chirp. Only first 200 samples after the largest peak are kept as a feature in EchoTag, excluding reflections from objects farther than 86cm away.

Classification

- Uses off-the-shelf classifiers
- Paper discusses K Nearest Neighbor (kNN) and Support Vector Machine (SVM)
- 200 features (samples after the first peak)

kNN

- 5 nearest neighbor using Euclidean distance
- 65% accurate over 1 cm resolution
 - Small training data, and non-linear acoustic signals

SVM

- One-against-all SVM
- N classifiers trained for N tags
- Location classified as tag #k if the kth classifier provides highest probability
 - Probability <50% means no tag is active
- Provides 98% accuracy over 1cm resolution

Optimization

- Continuous audio sensing is expensive in terms of power
- EchoTag generates audio, which is obtrusive

- EchoTag is activated only if inertial sensor detects no motion
- Locations coarsely classified using WiFi AP ids
 - EchoTag not activated if known APs are not detected

Tolerance Range

- Can decrease sensing frequency in order to increase tolerance
 - This allows for it to be less accurate, which could be useful
- Also, implementing a “No Tag SVM” instead of trying to locate a tag
 - In there test had only 1% false negatives

Optimization

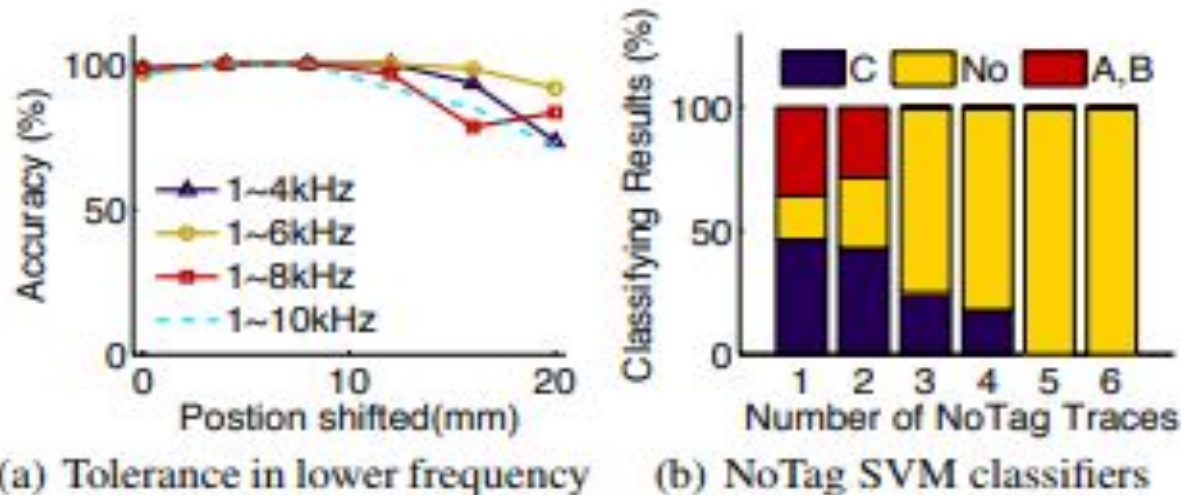


Figure 21—Extension of tolerance range. The tolerance range can be extended by sensing tags with lower-frequency signals. Building ‘NoTag’ classifiers can also prevent EchoTag from incorrect classification of misplacements.

Usability

Questions	Disagree	No option	Agree
Sensing accuracy is useful	1	0	31
Sensing noise is acceptable	0	3	29
Sensing delay is acceptable	1	6	25
Placing phones inside (echo)tags is easy	0	3	29
EchoTag can help me remember turning on silent mode when going to sleep	2	5	25
EchoTag can help me remember setting the timer for taking washed clothes	5	3	24
EchoTag can save my time in activating apps under specific scenarios	1	0	31

Summary

- EchoTag uses acoustic sensing for indoor localization
- Actions associated with locations

- Acoustic signature based on frequency absorption

- 98% accuracy, 1cm resolution using all-against-one SVM
- Not particularly robust to environment

- User study to gauge utility

Related Work

System	Resolution	Infrastructure	Signature
SurroundSense [3]	room-level	No	Fusion
Batphone [25]	room-level	No	Sound
RoomSense [23]	300cm	No	Sound
Radar [4]	400cm	Existing	WiFi
Horus [30]	200cm	Existing	WiFi
Geo [8]	100cm	No	Geomagnetism
FM [7]	30cm	Existing	FM
Luxapose [15]	10cm	Additional	Light
Cricket [22]	10cm	Additional	Sound/WiFi
Guoguo [17]	6–25cm	Additional	Sound
EchoTag	1cm	No	Sound

Table 1—Existing indoor location sensing systems.