

# TagBreathe: Monitor Breathing with Commodity RFID Systems

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- Importance of Human Respiration Monitoring
- Background and Related Work
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- Implementation & Evaluation
- Conclusion



#### Importance of Human Respiration Monitoring

Medical evidence suggests that sick people are heavy breathers. Respiration rates are different for different diseases.



Data resource: <u>www.NormalBreathing.com</u> Year: 2016



# Human Respiration Monitoring in Hospital

**Respiration is a key index of human health** 





# **Current Industry Products**

#### • Industry

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#### **Related Work**

Research

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#### RF signal based respiration monitoring



1. H. Abdelnasser, et al, "UbiBreathe: A Ubiquitous non-Invasive WiFi-based Breathing Estimator", ACM MobiHoc, 2015

2. X.F. Liu, et al, "Contactless Respiration Monitoring Via Off-the-Shelf WiFi Devices", IEEE Transaction on Mobile Computing, 2016

3. H. Wang, et al, "Human Respiration Detection with Commodity WiFi Devices: Do User Location and Body Orientation Matter?", ACM UbiComp, 2016



TOF

sweep 1

Research

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#### >RF signal based respiration monitoring

**> FMCW (Frequency Modulated Carrier Waves)** [4]



sweep 2

Time

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# Limitation

- Limitation for existing RF signal based methods
  - Low resolution for WiFi system
  - FMCW systems require customized active radio, which are not readily available on the market.
  - >Both systems may fail in detecting multiple persons

A low cost, non-intrusive and convenient system that is able to detect breaths of multiple persons is needed !



# Advantages of RFID systems

- RFID based methods
  - >Widespread use
  - Provides opportunities for nonintrusive cost-effective healthcare monitoring
  - High resolution of 0.038mm movement detection
  - >Work for multiple persons



\$5-10 cents each







#### Low Level Data of the Signal

Low level data

 RSS (Received Signal Strength)
 DFS (Doppler Frequency Shift)
 SPV (Signal Phase Values)



Due to the distance between tag and reader changes periodically, corresponding low level data also changes.





#### Limitation of using RSS

Low resolutionUnavailable in more challenging working scenarios



#### Limitation of using DFS

*DFS* during the one packet transmission is not reliable and may be subject to noises in practice.
 More suitable for high speed movement of tags







# Signal Phase Values

# High resolution of 0.038mm movement

# **\*Suitable for both high and low speed movement**

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#### The TagBreathe System

- Challenge 1: Channel Frequency Hopping in standard EPC protocol
- ✤ 0.2 seconds for 1 channel

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- $\bigstar \lambda$  and the phase offset change
- Discontinuity of phase values for every 0.2s

$$\theta = \left(\frac{2\pi}{\lambda} \times 2d + \theta'\right) \mod 2\pi$$

phase offset is caused by the circuits of reader and tag, which is independent of the distance





#### Challenge 1: Channel Frequency Hopping in standard EPC protocal

(1)Group the phase values according to channel indexes.

(2)Then, calculate the displacement during two consecutive phase readings in the same channel to eliminate phase offsets.



The displacement values are not influenced by the frequency hopping and track the periodic body movement mainly due to breathing !



Challenge 2: Extract the breath pattern from the waveform

FFT (Fast Fourier Transform)

The peak of the FFT output corresponds to the breathing rate.

The Fourier transform for a window size of w seconds is that it has a resolution of 1/w.

In our initial experiment, *w*=25s, the frequency resolution is 0.04 Hz which corresponds to 2.4 breaths per minute.





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# The TagBreathe System

#### Challenge 2: Extract the breath pattern from the waveform

Apply an FFT-based low pass filter to filter out high frequency noises and then extract the breathing signals.

Average 12 - 20 breaths per minute for a healthy person.
 Lower than 40 breaths per minute.

the cutoff frequency of the low pass filter = 0.67 Hz.





Challenge 2: Extract the breath pattern from the signal phase values

- Record the time stamps of the zero crossing events and calculate the instant breathing rate as:
- 1. Count the number of 0-crossing points M during  $\Delta t$
- 2. Breath rate =  $(M-1)/2\Delta t$



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# The TagBreathe System

#### Challenge 3: Weak signals and NLOS

With **Sensor Fusion of Multiple Tags**, *TagBreathe* can effectively improve the performance by constructively add raw data, which substantially enhances signal extraction in case that **signals are weak or are blocked in LOS path.** 

- TagBreathe overwrites 96-bit tag ID with 64-bit user ID and 32-bit short tag ID.
- We fuse the raw data before extracting breath signals. That is because it can effectively improve signal strength.



The displacement value for tag  $T_i$  collected during the time period  $[t, t + \Delta t]$  can be described as:

$$T_{j}:Tag j, j \in [1,n]; \delta t \in [t,t+\Delta t]$$
$$\Delta \mathbf{d}(t) = \sum_{j=1}^{n} \sum_{\delta t=0}^{\Delta t} \Delta d_{i+1}^{T_{j}}(t+\delta t) \longrightarrow \Delta \mathbf{D}(t) = \sum_{i=0}^{N} \Delta \mathbf{d}(t+i\Delta t)$$



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#### The TagBreathe System

#### TagBreathe workflow



*TagBreathe* interrogates multiple RFID tags attached to users and translate the phase value into distance.  $\Delta d_{i+1} = d_{i+1} - d_i = \frac{\lambda}{4\pi}(\theta_{i+1} - \theta_i)$ 

*TagBreathe* groups the readings and carries out raw data fusion by synthesizing multiple data streams  $\int d(t) = \sum_{n=1}^{n} \sum_{k=1}^{n} \sum_{k=1}^{n}$ 

$$\Delta \mathbf{d}(t) = \sum_{j=1}^{n} \sum_{\delta t=0}^{\Delta t} \Delta d_{i+1}^{T_j} (t + \delta t)$$

*TagBreathe* analyzes the synthesized data stream and extracts breathing signals for each user  $\frac{1}{f_{res}(t_{r})} = \frac{1}{f_{res}(t_{r})}$ 

$$\overline{f_{BR}}(t_i) = \frac{M-1}{2(t_i - t_{i-M})}$$



#### Implementation

#### TagBreathe prototype system



We perform a real-time and online implementation of TagBreathe prototype by using COTS RFID. 21



#### Default Configuration

Table I

System parameters and default experiment settings

Parameter	Range	Default
Channel	channel 1 - channel 10	Hopping
Tx power	15 - 30 dBm	30 dBm
Distance	1m - 6m	4m
Orientation	$0^{\circ}$ (front) - 180° (back)	front
Number of users	1 - 4 users	1 user
Tags per user	1 - 3 tags	3 tags
Breathing rate	5 - 20 bpm	10 bpm
Posture	Sitting, Standing, Lying	Sitting
Propagation path	with/without LOS path	with LOS path



#### • Evaluation Results



- The breathing rates range from 5 to 20 breaths per minute (bpm).
- Repeat the experiments for 100 times
- Each experiment lasts for two minutes.



#### Evaluation Results



Reading rate and RSSI in different orientation and accuracy.



#### Evaluation Results



Breathing rate accuracy with different **# of tags.** 



#### Evaluation Results



Breathing rate accuracy with different **# of users**.



#### Conclusion

We propose a low cost, non-intrusive COTS RFID system to monitor human breath.

We carefully design sensor fusing algorithms to extract breathing signals from the low level data reported by commodity readers.

TagBreathe can monitor breathing for multiple users



# Thanks!