Smart Mobile Health Systems: Experience with Biological Rhythm Sensing and System Privacy

Guoliang Xing

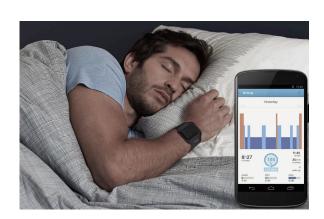
Associate Professor Department of Computer Science and Engineering Michigan State University

Outline

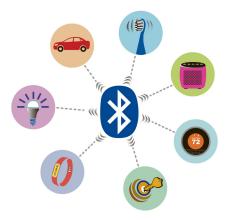
• Mobile health systems and privacy



Breath monitoring



Sleep quality monitoring

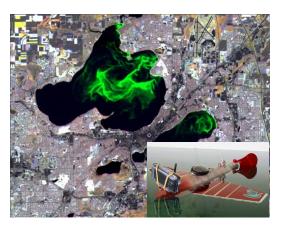


Privacy for smart devices

• Cyber Physical Systems



Real-time volcano tomography



Aquatic process profiling

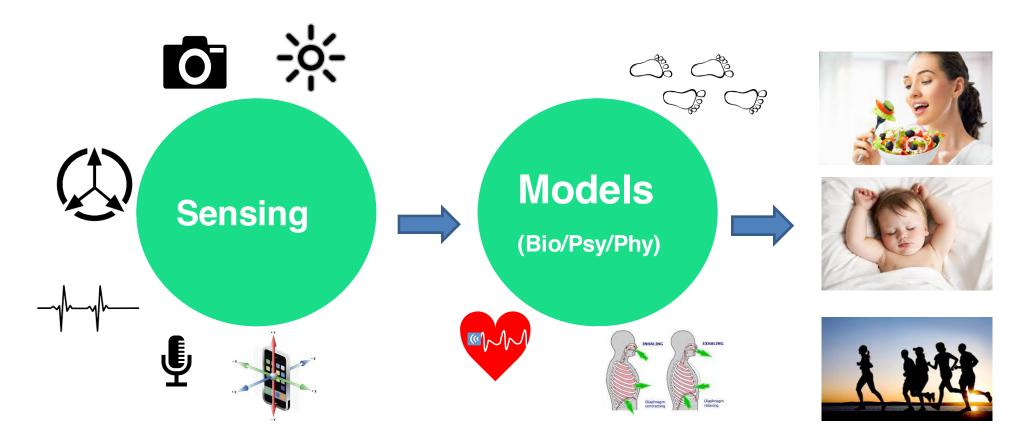


Data center thermal monitoring

Healthcare Crisis

- Total U.S. healthcare spending is 17.9% GDP
 - Avg. costs/stay \$10,000, 47% up since 1997
 - 45K death annually due to lack of healthcare access
- World-wide aging problem
 - US: 3M/yr baby boomers retire for next 20 yrs
 - China: 1/10 population aged 65+, 18% in 20 yrs
 - HK: 1/8 population aged 65+, 1/4 in 20 yrs
- China's national strategy Healthy China (健康中国) 2030

Mobile Health Approach



"transformation of healthcare from **reactive** and **hospitalcentered** to **preventive**, **proactive**,**person-centered** and focused on **well-being** rather than disease" -- National Science Foundation



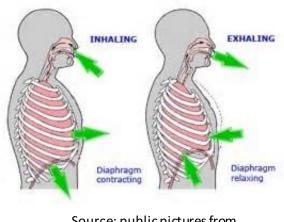
Source: public pictures from the Internet

Biological Rhythms

- Cyclic changes of bodily chemical/function
 - Sleep/wakefulness cycle, body temperature, blood pressure, running/walking, feeding cycles, heartbeat and respiration...







Source: public pictures from the Internet

Significance

 Bio rhythms is essential for daily productivity and well-being

- Disruptions can affect vital body functions
 - Hormonal balance, metabolism, digestion, sleep
 - Diseases such as diabetes, obesity, and depression

Challenges

- Long-term monitoring proves critical for detecting early signs of many diseases

 Have been largely limited to clinical settings
- Examples: sleep quality, breath pattern



Cardiopulmonary exercise testing (CPET) http://www.mettest.net/cardiopulmonary_exercise_test.html



Running Rhythm Monitoring



https://www.youtube.com/watch?v=hZMZqt4Pae4

Running is a Popular Exercise

65.5 million (~20.6%) jog/run in the U.S., 2014

~7% annual increase since 2008

ource: Nielsen Scarborough 2014)

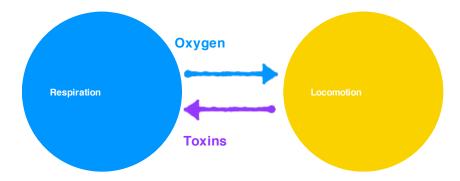
Image Credit: kris krüg @Flickr

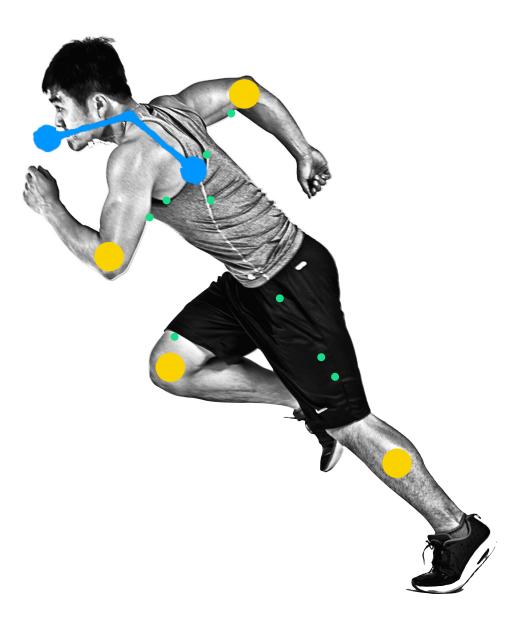
2132



More than Just Footwork

Running Rhythm the <u>Coordination</u> b/w **Breathing** and <u>Strides</u>





More than Just Footwork

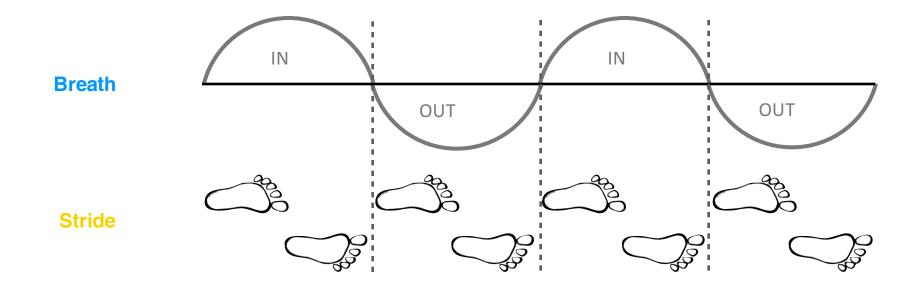
Running Rhythm the <u>Coordination</u> b/w **Breathing** and <u>Strides</u>



New Runner Gasping for air



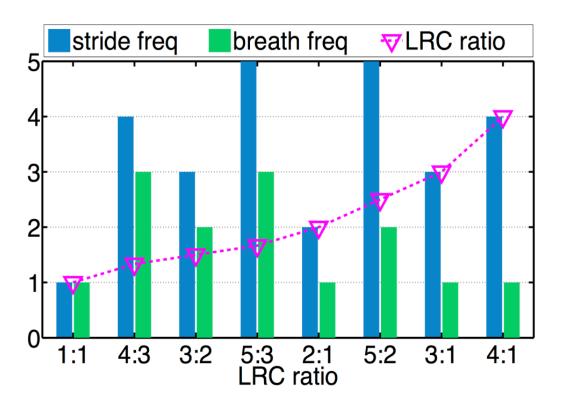
An Example of Running Rhythm



Breathe **IN** <--> 2 Steps Breathe **OUT** <--> 2 Steps

Locomotor Respiratory Coupling (LRC)

Rhythmic exercises (e.g., running, cycling)



a small number of integer ratios in humans during running

Tracking Progress

- Another aspect of the run
- Ability of maintaining stable LRC is a good indicator of fitness level

Helps push that extra little bit

Avid Runner,
 RunBuddy Experiment
 Participant



- Show promises to enhance running experience by music suggestion
- Play music with proper tempo to guide new runners to run in a comfortable rhythm

Proper rhythm helps postpone fatigue and increase endurance

CPET: Cardiopulmonary Exercise Testing



iBreath

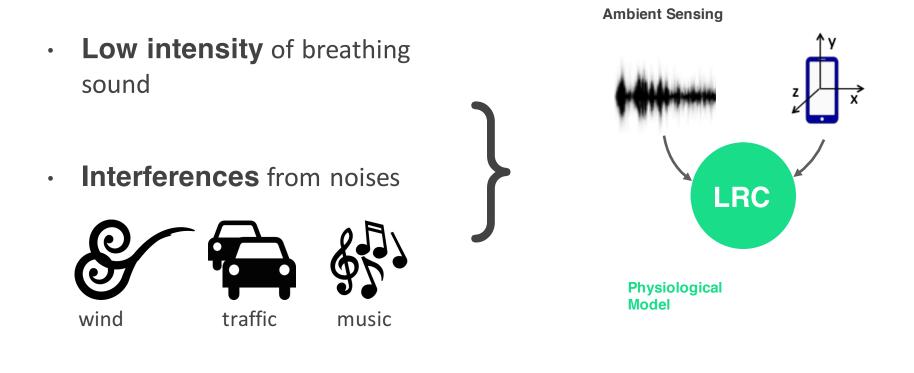
Continuous running rhythm (LRC) monitoring

- Convenient and unobtrusive
- Anytime, Everyone, Anywhere





Challenges



• **Privacy** requires realtime processing





Lightweight & Efficient Signal Processing



quiet environment

public road w/ traffic

traffic

3

low

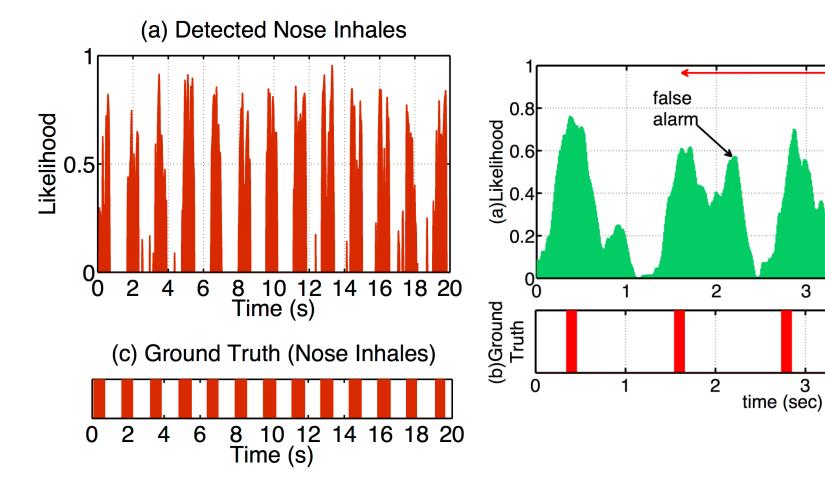
likelihood

4

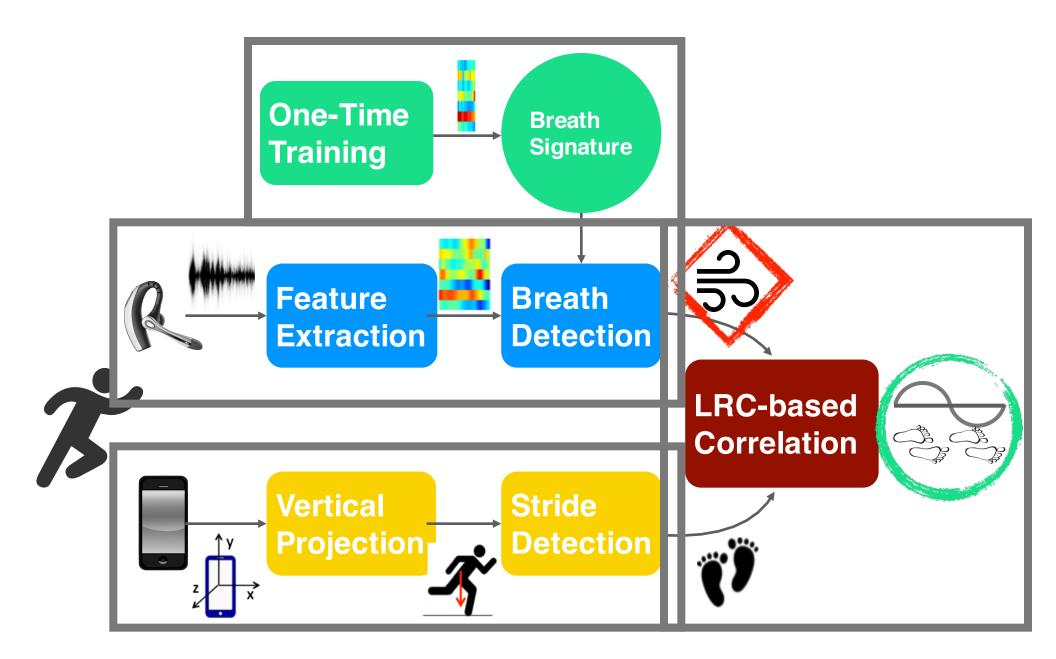
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5

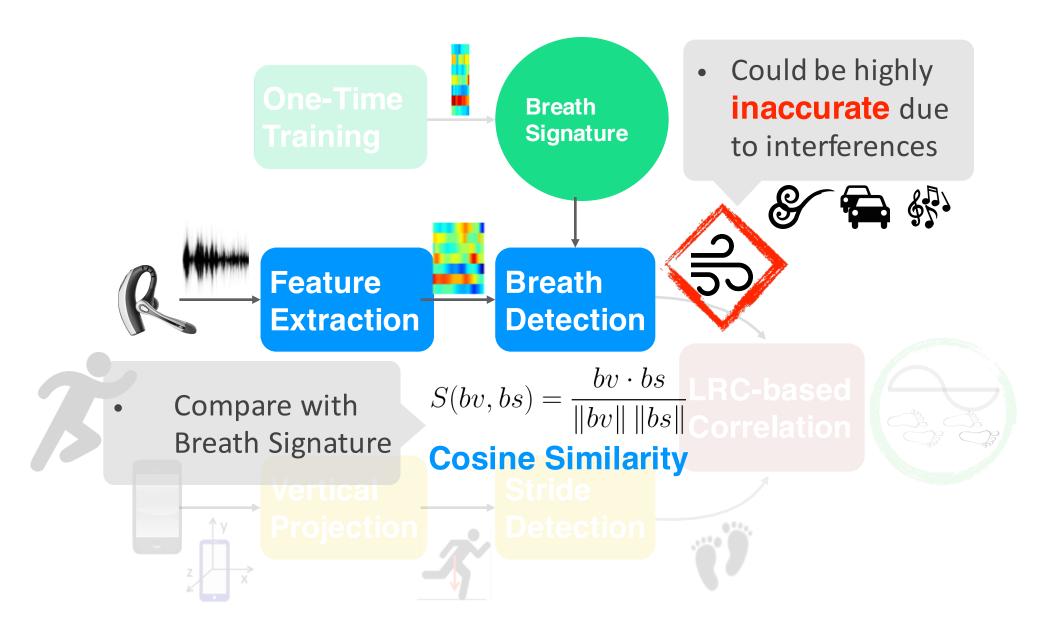
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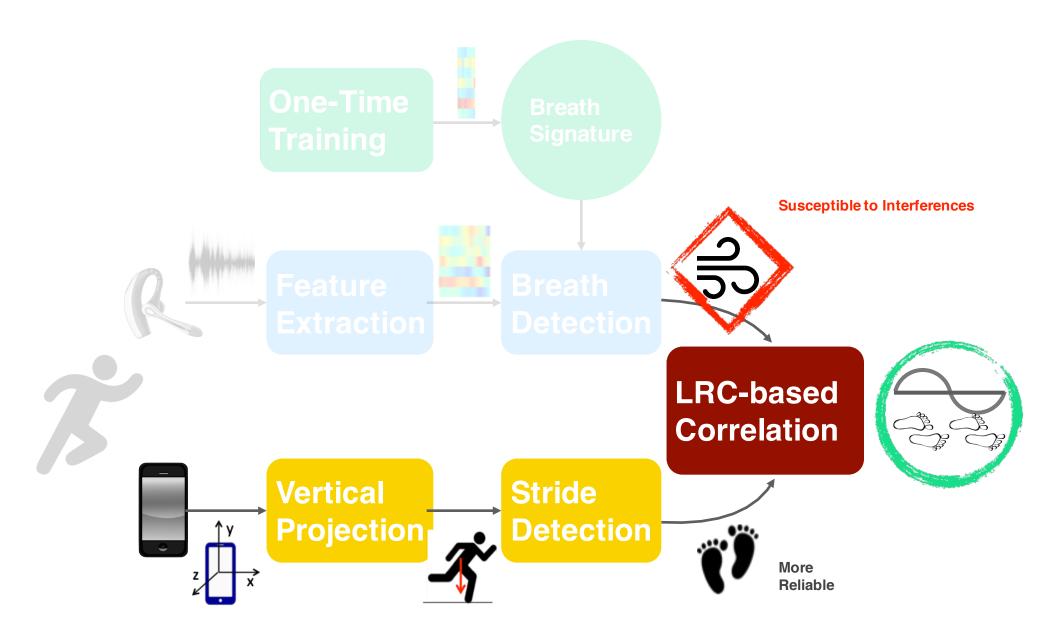
System Architecture



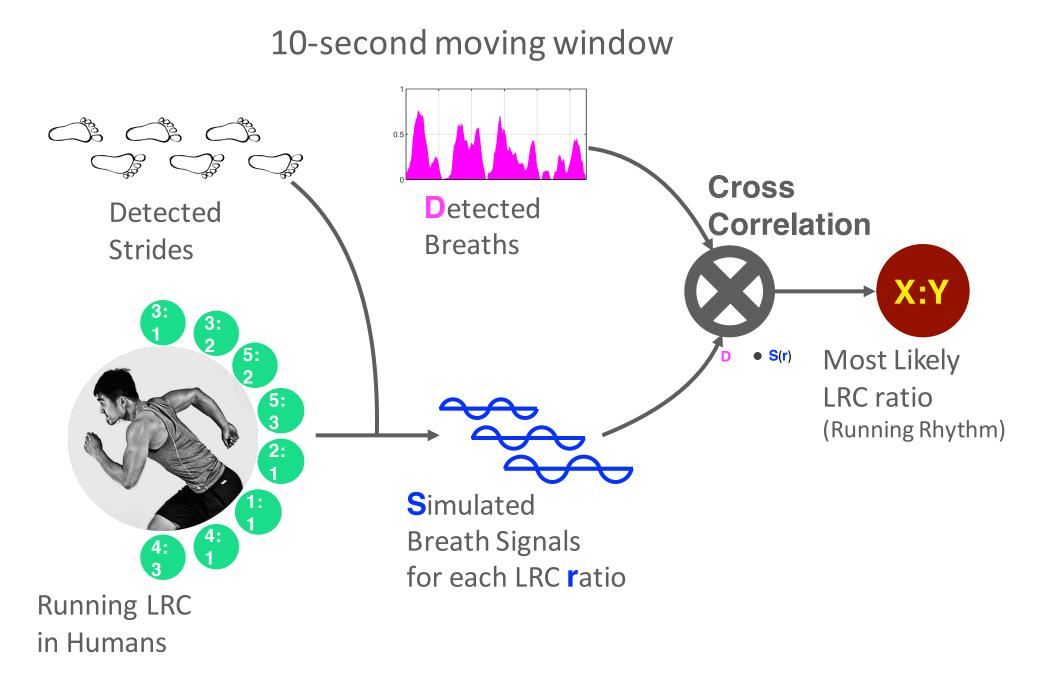
Preliminary Breath Detection

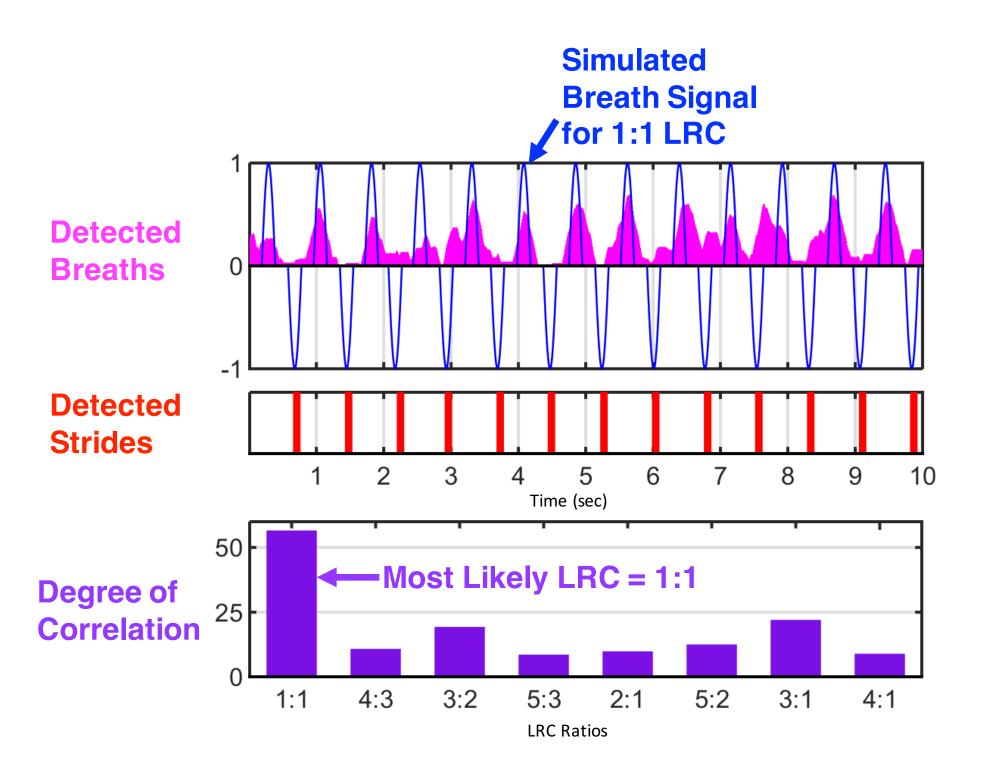


LRC-based Correlation



LRC-based Correlation

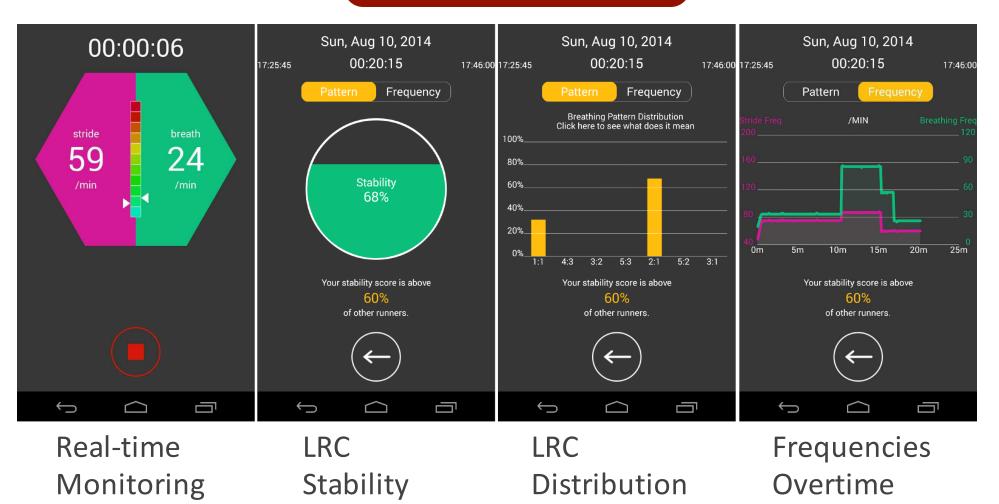






MobiCom 2014 Best Mobile App

Third Place



Evaluation

Measurement

every 5 sec, with 10-sec window

Metric

% of Correct Measurement

Accurately measure LRC Ratio

92.7% of the time

IRB-Approved

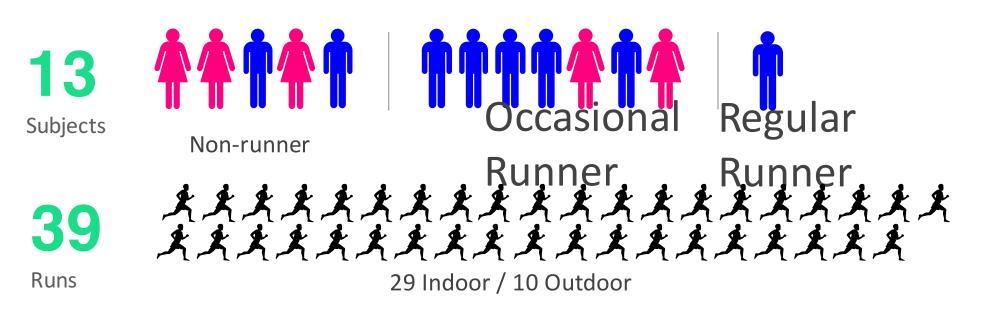
by the Institutional Review Board at Michigan State University

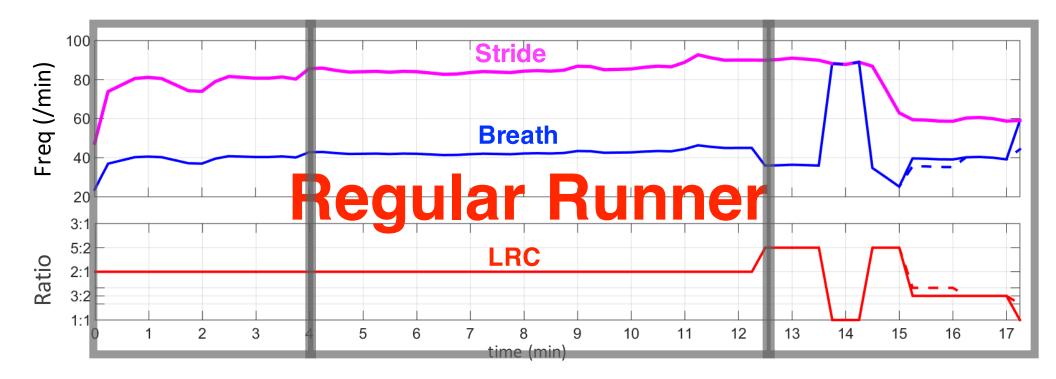
Procedure

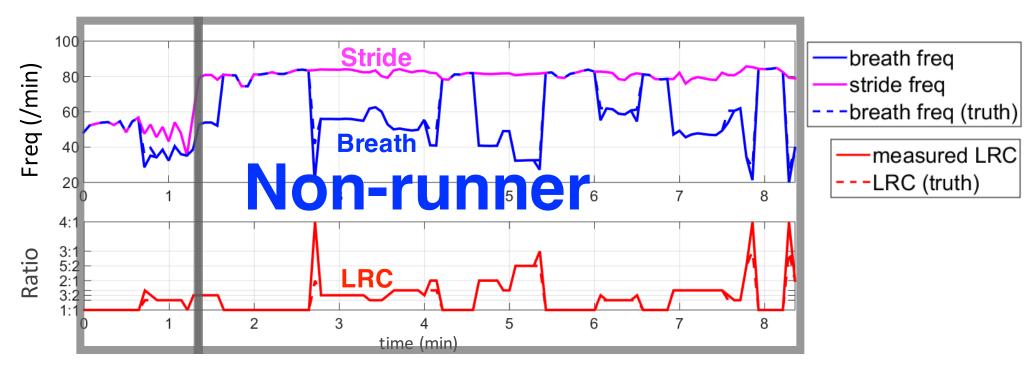
- Run as usual
- No limit on speed, distance and phone placement

Ground Truth

- Inline mic under nose (34/39)
- Manual counting (5/39)







Unobtrusive Sleep Monitoring



Poor Sleep is Common

1 out of 5 Americans are affected by various sleep disorders

[National Institutes of Health]

RLS	OSA	DSPD	RBD
Restless	Obstructive	Delayed	Rapid Eye
Legs	Sleep	Sleep Phase	Movement
Syndrome	Apnea	Disorder	Behavior Disorder



image (top left) source: http://www.venusbuzz.com/archives/28076/study-lack-of-sleep-slows-down-productivity/sleepy-at-work/ image (top right) source: http://www.dchtsdf.org/news_and_press/driver_fatigue_a_factor_in_20_percent_of_crashes.aspx image (bottom left) source: http://micah.sparacio.org/05/16/2011/10-ways-to-fight-depression/ image (bottom right) source: http://thinkprogress.org/health/2012/06/18/501286/17-million-tors-overweight/



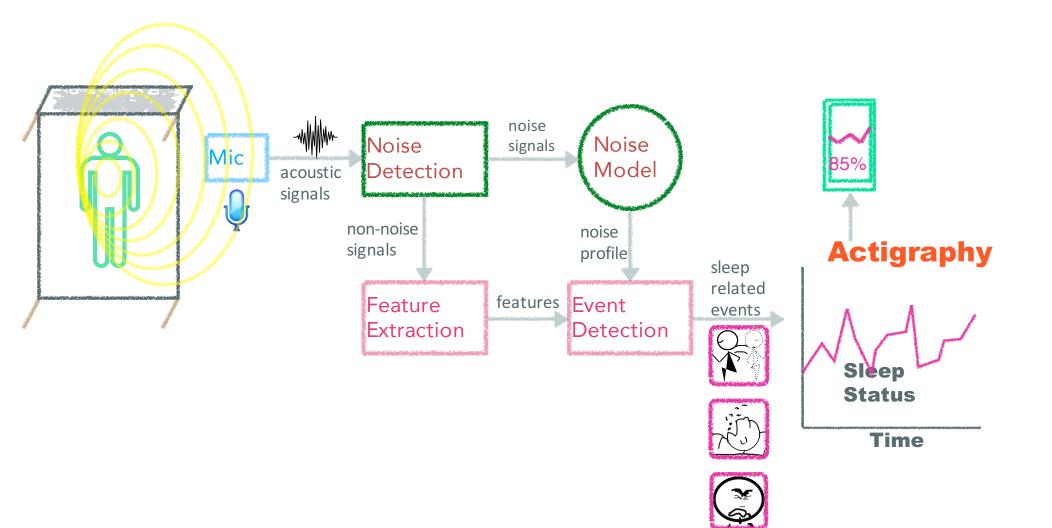
iSleep

- The first sound-based sleep monitor
- Helps users keep track of sleep quality and better understand sleep

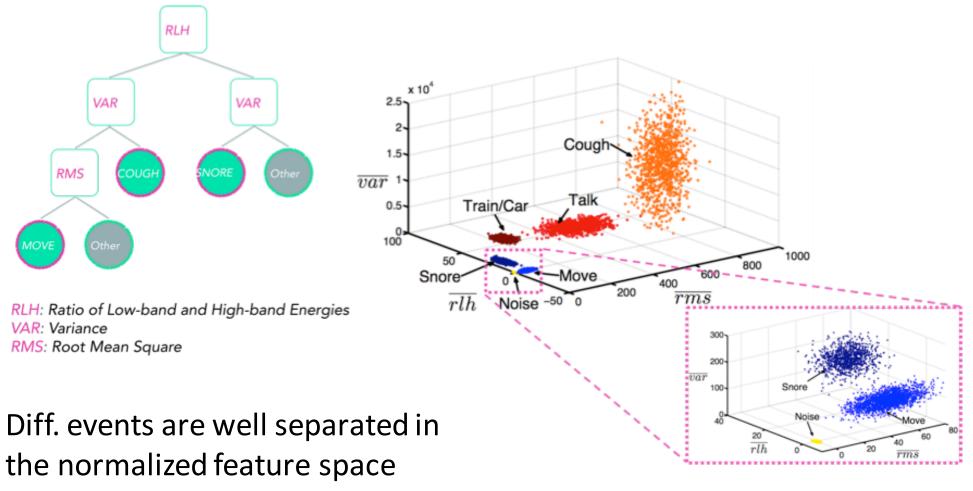
()

- Contact-free and gadget-free
- Snoring and coughing detection

System Overview



Feature Space & Decision Tree

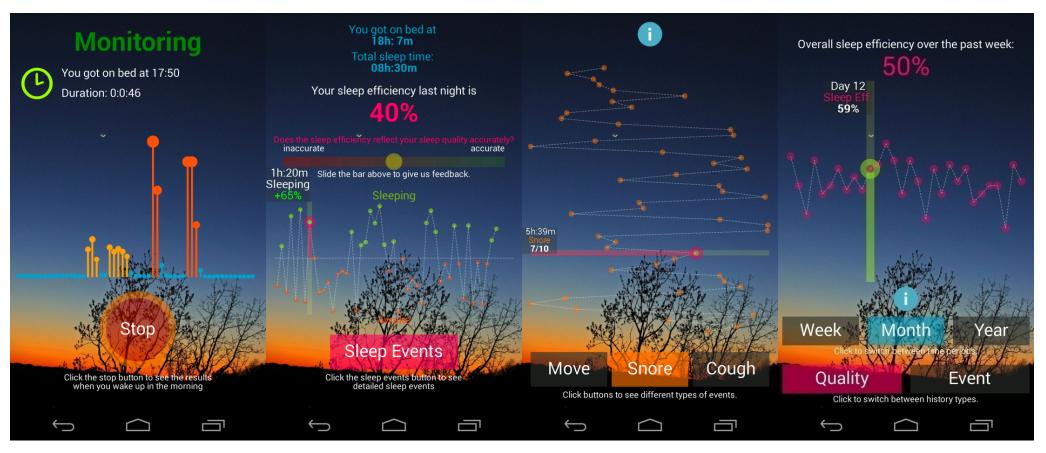


(plotted using data from 7 subjects)



5K Users/2 mo without marketing

Best Mobile App Third Place, MobiCom 2013



Monitoring

Sleep Quality

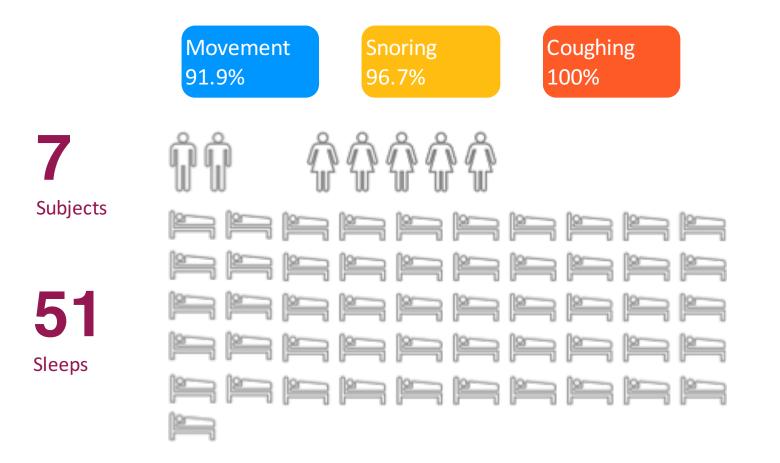
Sleep Events

History

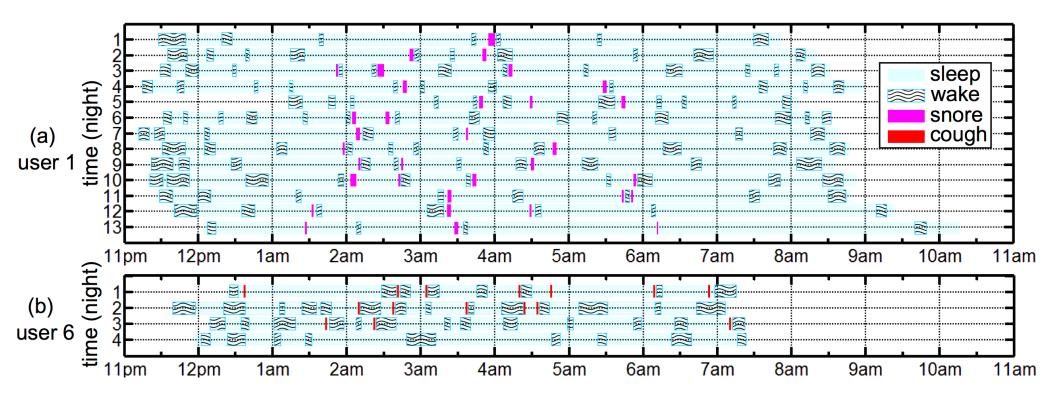
Evaluation

MSU IRB#12-1178

Approved by the Institutional Review Board at Michigan State University

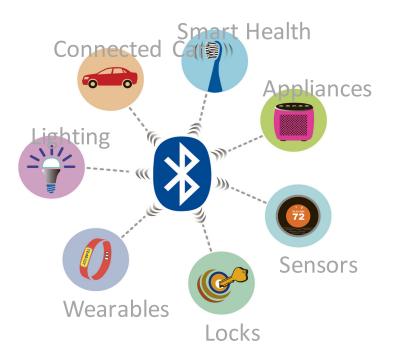


Long-term Result



Wireless Privacy for Smart Devices

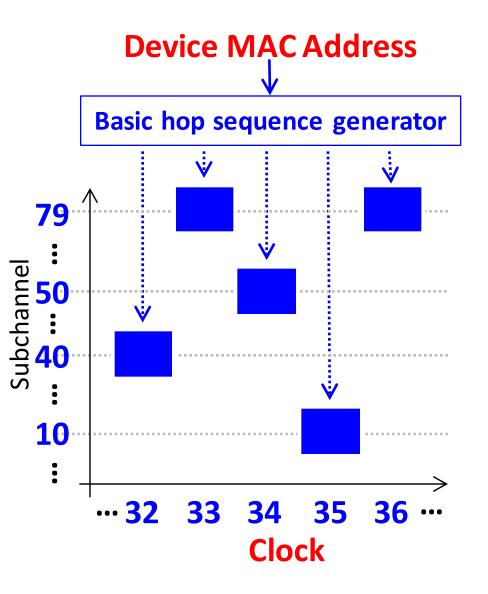
- Bluetooth
 - 3 billions Bluetooth devices were shipped in 2015
- Near-field communication (NFC)
 - 200 M NFC equipped smartphones shipped in 2013
 - 50% smartphones support NFC in 2015





Basic Frequency Hopping

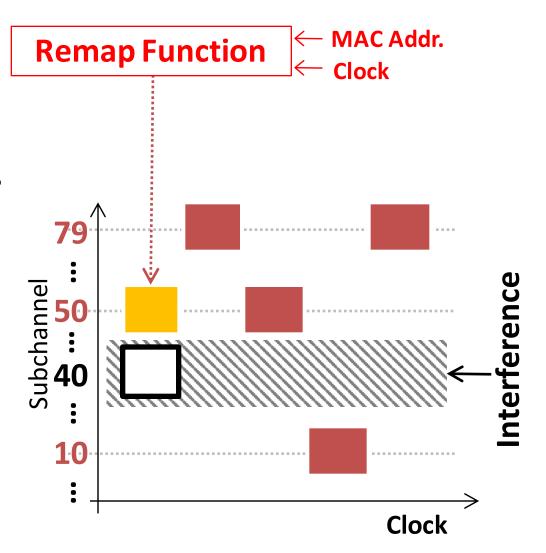
- Switch b/w 79 subchannels randomly
 - 1600 hops/s
- Current hop is defined by
 - Device MAC address (public)
 - A 27-bit logic clock (secret)
 - Index of current hop
 - A basic hop seq. of 227 long



Adaptive Frequency Hopping

Remaps `bad' subchannels

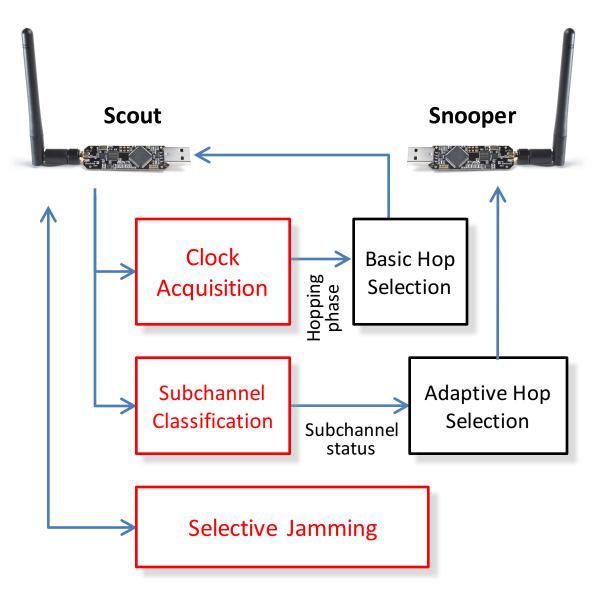
- MAC address (public)
- Clock (secret)
- Vendor-dependent behavior
 - No standard definition of 'good' and 'bad' subchannel conditions



BlueEar: A Dual-Radio Sniffer

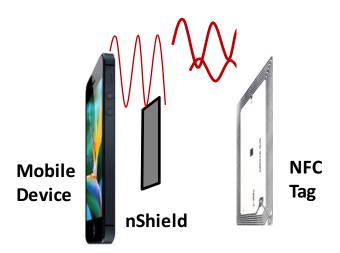
• The Scout radio

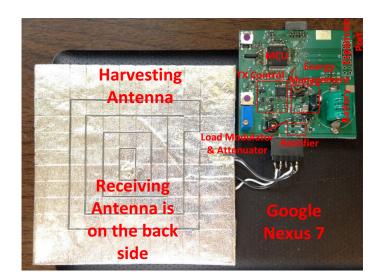
- Sniffs a single subchannel to acquire clock value
- Perform basic hoping to learn subchannel status
- Selectively jam to avoid interference
- The Snooper radio
 - Follow adaptive hopping to capture packets



nShield: A Noninvasive Security Device

- Dynamic NFC signal strength attenuator
 - Absorbs excess NFC RF energy
 - Variable attenuation w/o sacrificing communication reliability
- Noninvasive
 - No modification, no physical connection to host
- Harvest and store absorbed RF energy
 - Enable perpetual operation





Cyber-Physical Systems

- "Cyber-physical systems are engineered systems that are built from and depend upon the synergy of computational and physical components"¹
- Many critical sustainability application domains

 Environment, smart grid, medical, auto, transportation...
- No. 1 national priority for Networking and IT Research and Development (NITRD)
 - NITRD Review report by President's Council of Advisors on Science and Technology (PCAST) titled "Leadership Under Challenge: Information Technology R&D in a Competitive World", 2007

Our CPS Projects



Data Center Monitoring, HPCC, MSU



Tungurahua Volcano, Ecuador



Volcano Monitoring Sensors

- Data center thermal monitoring
- Real-time volcano monitoring
- Aquatic process profiling



Harmful Algae Bloom in Lake Mendota in Wisconsin, 1999

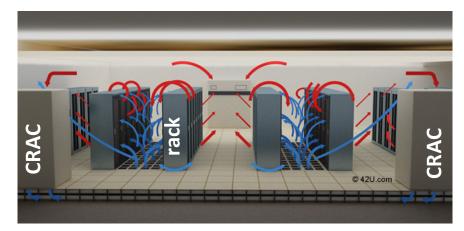


Robotic fish, Smart Microsystems Lab, MSU

Motivation



EMC's new data center in Durham, NC



Raised-floor cooling [www.42u.com]

- Data centers eat massive energy
 - An industry data center = a mid-size town
- 60% non-computing energy ratio [Uptime 2012]
 - 50% for cooling

24°C in 90% data centers vs. recommended 27°C

– 10% for circulation

High fan speeds and simple control

Our Solutions

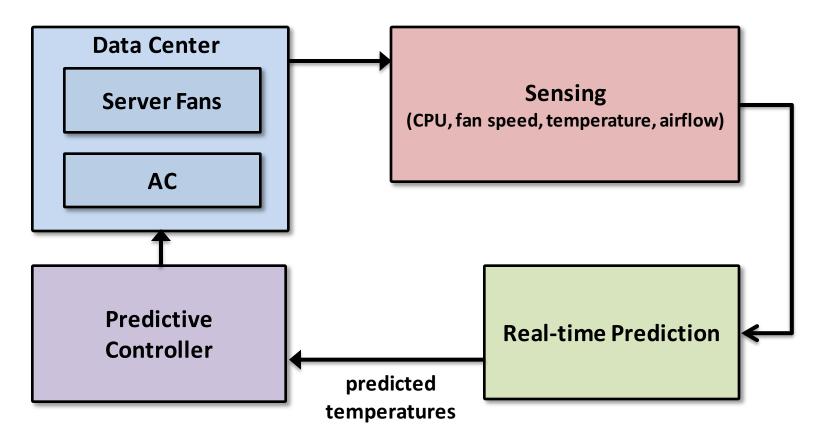
- Data center sensor system
 - Built-in sensors, external sensor network

Temperature & energy prediction [RTSS'12, TOSN'15]

Sensor data + energy models + candidate control action

• Predictive control [RTSS'14]

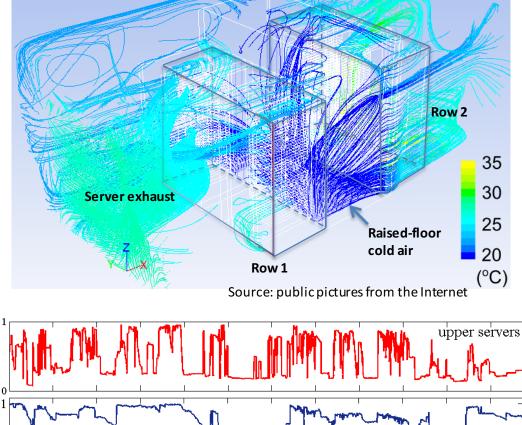
Constrained optimization



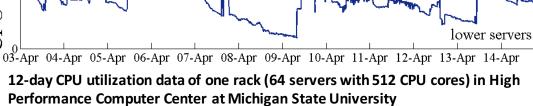
Challenges

CPU utilization

• Complex air and thermal dynamics

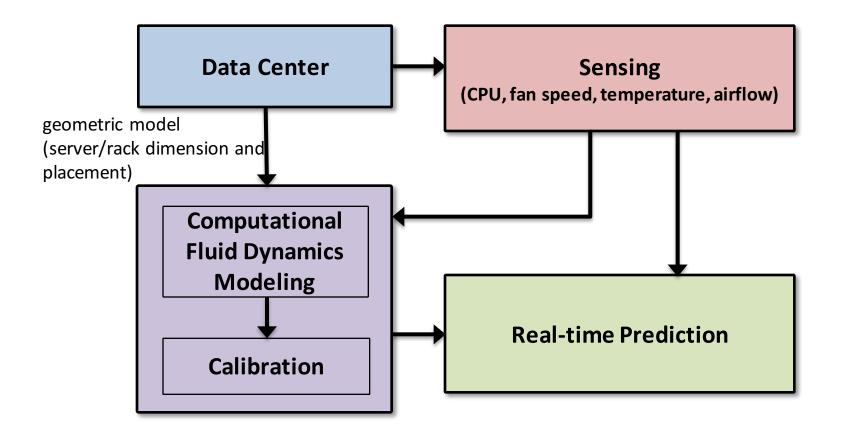


 Highly dynamic workloads

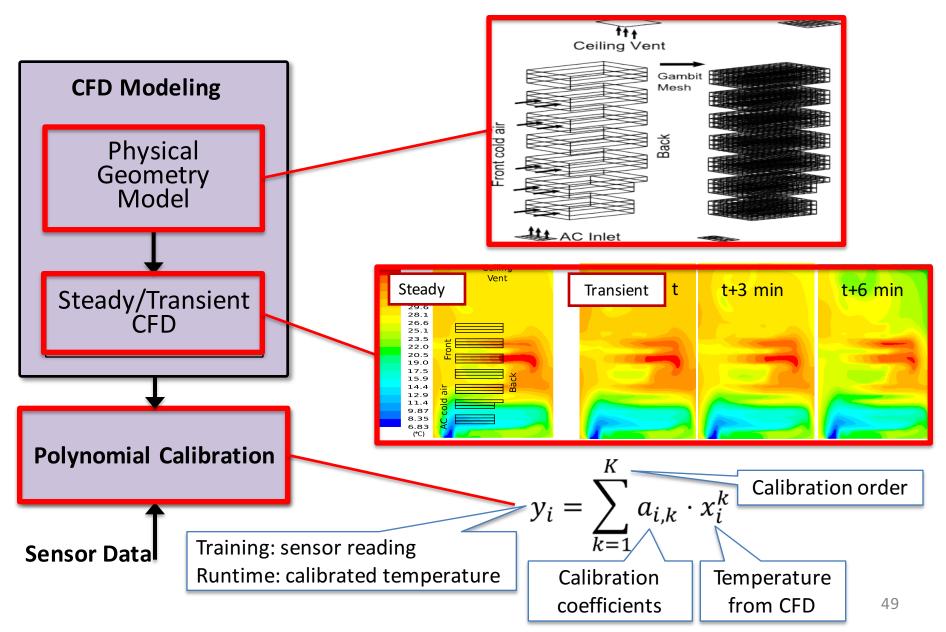


System Architecture

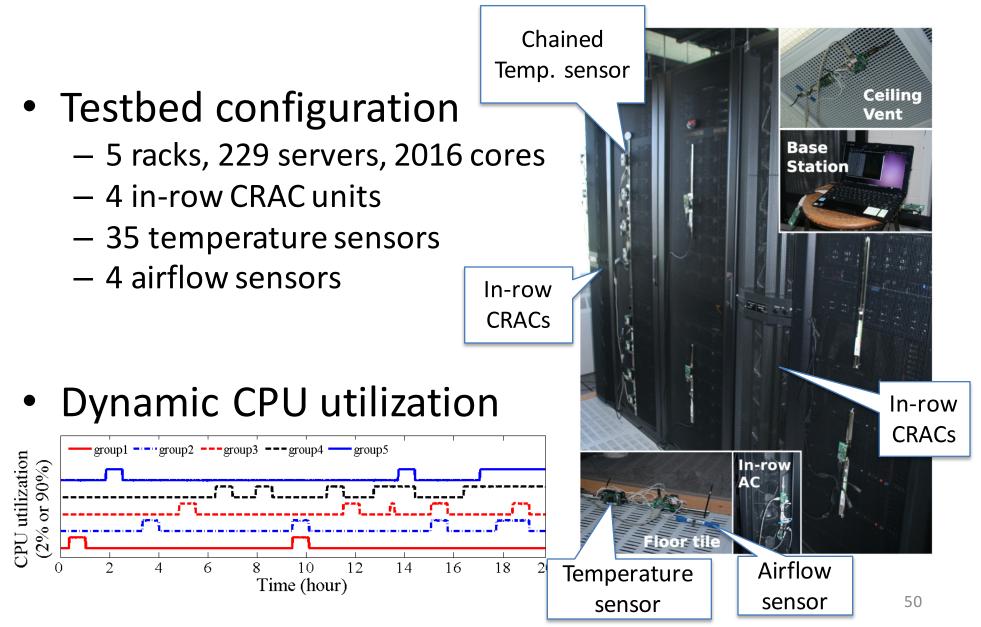
- CFD Simulation + Wireless Sensing + Data-driven Prediction
 - Preserve realistic physical characteristics in training data
 - Capture dynamics by in situ sensing and real-time prediction



CFD Modeling & Calibration

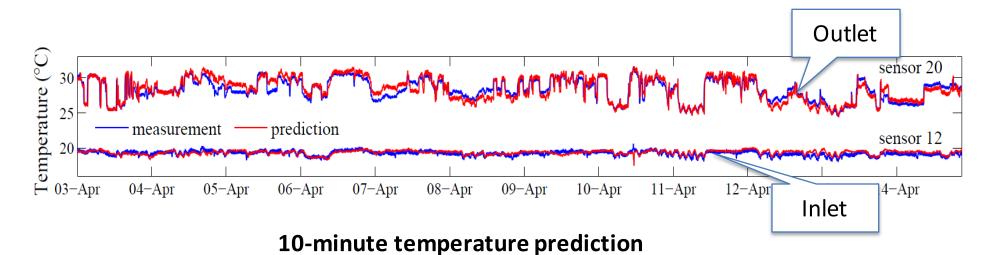


Data Center Experiment



Experiment Results

12-day experiment



Volcano Hazards



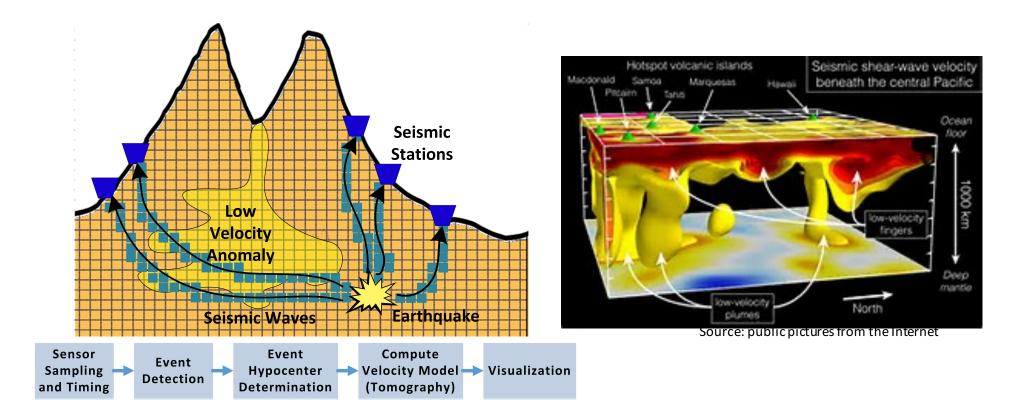
Eruption in Chile, 6/4, 2011 \$68 M instant damage, \$2.4 B future relief. www.boston.com/bigpicture/2011/06/volcano_erupts_in_chile.html



Eruptions in Iceland 2010 A week-long airspace closure [Wikipedia]

- 7% world population live near active volcanoes
- 20 30 explosive eruptions/year

Volcano Tomography



State of the Art

- Seismic activity monitoring
 - Earthquake localization, tomography, early warning etc.
- Traditional seismometer
 - Expensive (~\$10K/unit), difficult to install & retrieve
 - Only ~10 nodes installed for most threatening volcanoes!

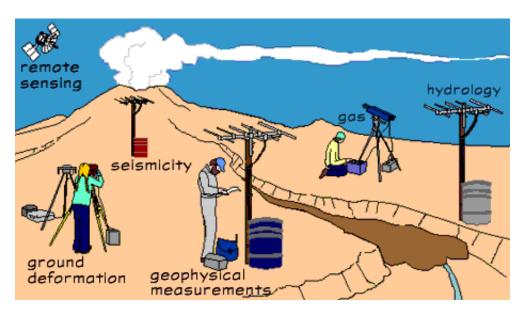
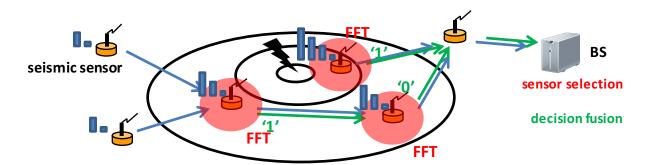




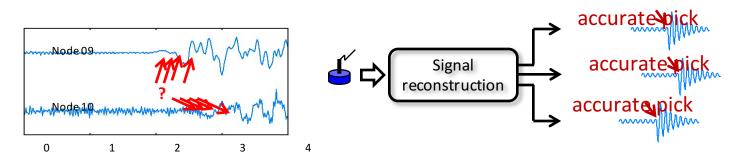
Photo credit: USGS, <u>http://volcanoes.usgs.gov/activity/methods/</u>

Our Solutions

• In-network sensor selection & fusion [RTSS'10, TOSN'13]



• Seismic event timing [IPSN'13]



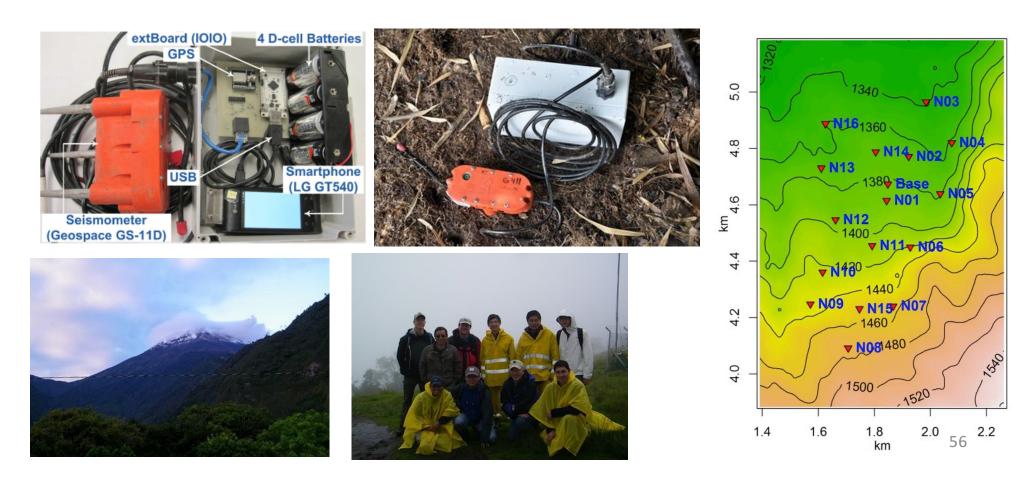
Source localization Seismic tomography

...

• In-network tomography [SECON'13]

Systems and Deployments

- Ecuador June 2013
 - Detected event 20Km from Tungurahua Volcano
- Chile January-March 2015
 - 16 nodes plus base station



Aquatic Environment Monitoring

• Monitoring aquatic ecosystems is critical for urban planning, public safety etc.

- HABs, oil spills, invasive species...





harmful algal blooms (HABs)

oil spill

Existing Approaches



manual sampling

patrol boat/ship

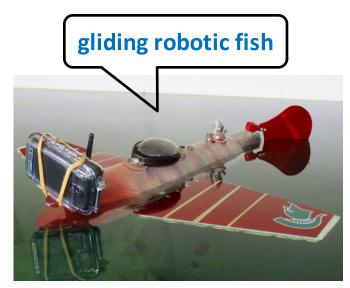
remote sensing

AUV/sea-glider

- Manual sampling
 - Small-scale, labor-intensive
- Patrol boat/ship
 - Costly, short-term
- **Remote sensing** (e.g., balloon, satellite)
 - Highly costly, low-resolution
- AUV/sea-glider
 - Costly (~\$50k), bulky, heavy

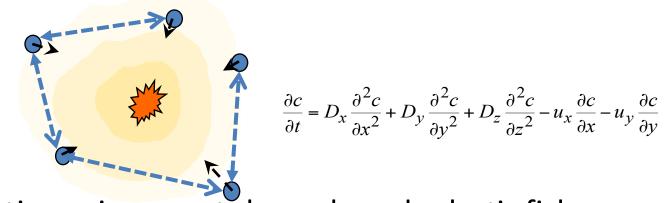
Our Prototypes

- Multi-modality sensing
 - Camera(s)
 - Inertial sensors
- Rich storage & computation power
 - Online learning algorithms
 - Computer vision algorithms
- Controlled mobility
 - Adaptation to environment dynamics
- **Low-cost** (~\$3k)

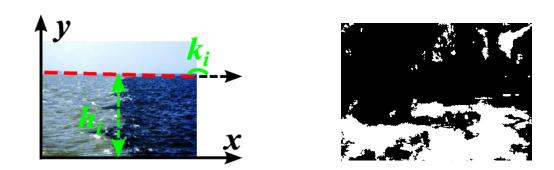


Overview of Our Work

• Diffusion process profiling & reconstruction [IPSN'12, RTSS'12]



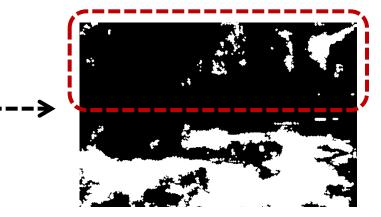
• Debris detection using smartphone-based robotic fish [IPSN'14, IPSN'15]



Detection in Complex Environment



captured image





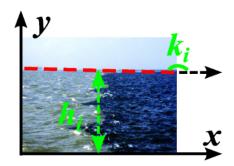
detection result

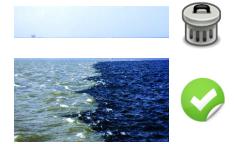
Excessive false alarms

- Water area: target algal patches in green
- Non-water area: trees in similar color
- Partition water and non-water areas
 - Focus on the water area only

Vision-based Segmentation

- Segmentation reference: shoreline
- Hough transform
 - Line extraction
 - Visual features: h_i and k_i
- Drawbacks
 - Compute-intensive (> 2 seconds/frame)
 - Low image quality
 - Blocked line-of-sight





Inertia-assisted Segmentation

Motivation

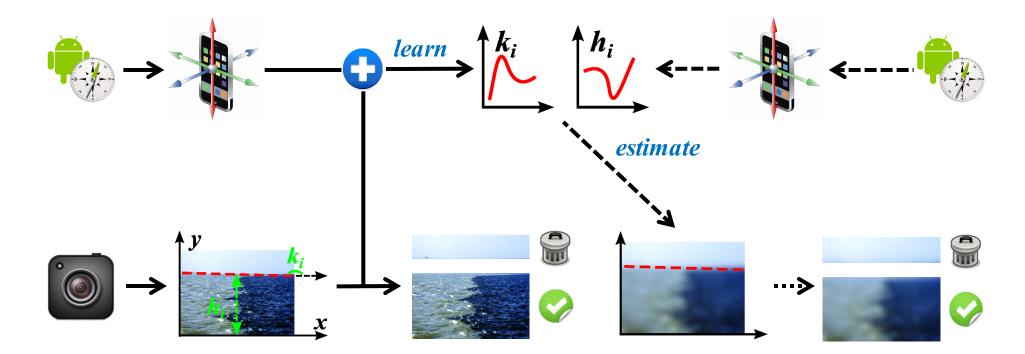
- Camera orientation characterizes projection
- Various on-board inertial sensors

Mapping models

- Relate inertial with visual sensing

 $k_{i} = \omega_{1} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{3} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{3} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{3} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{3} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{3} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{3} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{3} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{3} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{4} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{4} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{4} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{4} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{4} \times \underbrace{f_{1} - \omega_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{4} \times \underbrace{f_{1} - \omega_{2} \times f_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{4} \times \underbrace{f_{1} - \omega_{2} \times f_{2} \times f_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{4} \times \underbrace{f_{1} - \omega_{2} \times f_{2} \times f_{2} \times f_{2} \times f_{2}}_{\text{Euler, angles trigonometric functions of}} f_{4} \times \underbrace{f_{1} - \omega_{2} \times f_{2} \times f_{$

Energy-efficient Image Segmentation



Learning Phase \rightarrow

- Use detected visual features
- Learn mapping models

Estimation Phase -→

Use estimated visual features

Experiment Settings

• Samba prototype

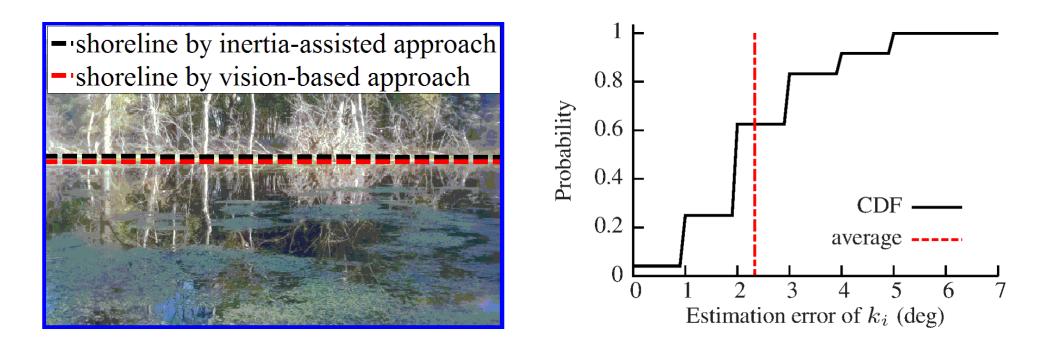
- Samsung Galaxy Nexus
- Gliding robotic fish
- 6.24 MB storage, 10.8 MB RAM
- Frame: 720 × 480 @ 0.5 fps



• Experiment environments

- Field experiments in an inland lake: ~ 200,000 square feet
 Captured and processed 5,211 frames
- Lab experiments in a water tank: 15 feet × 10 feet
 Evaluated under more dynamic environment

Image Segmentation Performance



Inertia-assisted approach accurately estimates the shoreline

Acknowledgement

- Current & past group members
 - 15 Ph.D students, 4 postdocs
- Collaborators
 - 20+ top experts in algorithms, signal processing, robotics, image processing, mechanical engineering, volcanology, communications, nursing
- National Science Foundation
 - Total 10 grants (8 as PI, 2 as Co-PI), over 8M USD since 2009

Representative Publications

- Practical Bluetooth Traffic Sniffing: Systems and Privacy Implications, The 14th International Conference on Mobile Systems, Applications, and Services (**MobiSys**), 2016, acceptance ratio: 31/197 = 15.7%.
- RunBuddy: A Smartphone System for Running Rhythm Monitoring, The ACM International Joint Conference on Pervasive and Ubiquitous Computing (**UbiComp**) 2015, acceptance ratio: 93/394 = 23.6%.
- Unobtrusive Sleep Quality Monitoring using Smartphones, The 11th ACM Conference on Embedded Networked Sensor Systems (SenSys), 2013, acceptance ratio: 21/123 = 17%.
- Nemo: A High-fidelity Noninvasive Power Meter System for Wireless Sensor Networks, ACM/IEEE Conference on Information Processing in Sensor Networks (**IPSN**), acceptance ratio: 24/115=21%, **SPOTS Best Paper Award**.
- Beyond Co-existence: Exploiting WiFi White Space for ZigBee Performance Assurance, The 18th IEEE International Conference on Network Protocols (ICNP), 2010, acceptance ratio: 31/170 = 18.2%, Best Paper Award.
- Supero: A Sensor System for Unsupervised Residential Power Usage Monitoring, 11th IEEE International Conference on Pervasive Computing and Communications (PerCom), 2013, acceptance ratio: 18/170 = 10.6%, Best Paper Award Runnerup.
- Passive Interference Measurement in Wireless Sensor Networks, The 18th IEEE International Conference on Network Protocols (ICNP), 2010, acceptance ratio: 31/170 = 18.2%, Best Paper Candidate (6 out of 170 submissions).
- Negotiate Power and Performance in the Reality of RFID Systems, The 8th Annual IEEE International Conference on Pervasive Computing and Communications (PerCom), 2010, acceptance ratio: 27/227=12%, Best Paper Candidate (3 out of 227 submissions).
- ZiFi: Wireless LAN Discovery via ZigBee Interference Signatures, The 16th Annual International Conference on Mobile Computing and Networking (**MobiCom**), Chicago, USA, September 2010, acceptance ratio: 33/233=14.2%.
- LEAD: Leveraging Protocol Signatures for Improving Wireless Link Performance, The 10th International Conference on Mobile Systems, Applications, and Services (**MobiSys**), acceptance ratio: 33/210 = 15.7%
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