A Long-Term Sensory Logging Device for Subject Monitoring



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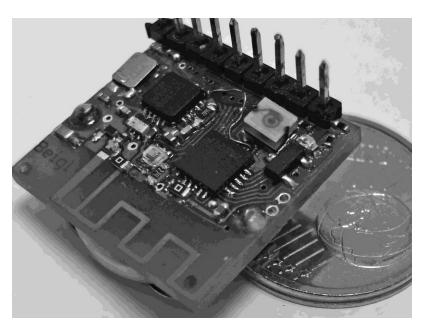
Motivation

Group focus: Pervasive Computing

- sensor networks: ultra-low power protocols/hardware (e.g. WoR, superimposing signals)
- context/activity recognition: non-personalized with high number of classes

Logging Device Requirements

- Creation of context recognition data sets
- Local data storage (as opposed to transmission)
- Easily modifiable
 - Data preprocessing
 - File formatting
- Easy access to data



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Requirements and Hardware

Lightweight for 24/7 wearable use

- < 25g w/ CR2477 >7 days
- < 50g w/ AAA >26 days
- Sensor board (Akiba)
 - PIC18
 - Light, Temperature, Vibration
 - 3 pin A/D external

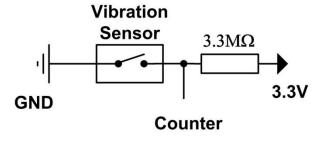
Memory Board

- PIC32
- microSD
- 'Eject' button



85mm



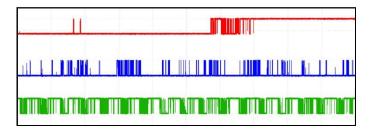


Requirements and Software

- Software consists of two entities
- Sensor board software
 - Samples sensors
 - Preprocesses data (unit conversion etc.)
 - Minimal application
- Memory board software
 - FAT32 system on microSD
 - Communication with Sensor board (UART)
 - Card ejection and insertion

Demonstative Application

- Demonstrate sensing / preprocessing / saving
- Inspired by medical activity monitoring
- Classify activity level based on vibrational intensity
- 3 activities selected
- 2 subjects used to create thresholds
- 1Hz storage rate:
 - Sample (>10 kHz)
 - Calculate activity level
 - Save to memory board



Typing	Walking	Jump-Rope
0	648	1966
0	228	1266
0	594	2040
0	1000	2734
0	1188	1628
0	1444	1898
34	2172	1284
14	1506	1236
0	1484	1972
0	1262	1986

ID	Level	Repr. Activity	Levels
1	Low	Typing	≤ 550
2	Medium	Walking	551 - 1508
3	High	Jump-Rope	≥ 1509

Evaluation

- 3 subjects (2 + 1) used for evaluation
- 30s per activity per subject
- Mixed results:
 - Inter-personal variance
- Overall: 74% recognition
- Possible causes
 - Less than optimal thresholds (more training data)
 - Poor classifier (see further work)
 - Inter-personal differences (personalization)

• Rate of consumption: 4.478mW, >26 days 2 x AAA

	Subject 1	Subject 2	Subject 3
Low (Typing)	81	97	100
Medium (Walking)	72	69	86
High (Jump-Rope)	0	66	97

CLASSIFICATION RATES IN PERCENT BY USER AND ACTIVITY

CONFUSION MATRIX FOR CLASSIFIER OUTPUT IN PERCENT

	Low	Medium	High
Low (Typing)	95	3	2
Medium (Walking)	6	75	18
High (Jump-Rope)	2	43	55

A Lesson Learned

- Energy Consumption strongly dependent on microSD card
- Large variance between brands
- Energy consumption correlates with transfer speeds

	Nokia	Kingston 4GB	Kingston 2GB	SanDisk
Write	0.5µJ/B	3.05µJ/B	3.15µJ/B	0.27µJ/B
Read	0.1µJ/B	0.5µJ/B	0.53µJ/B	0.09µJ/B
Write speed	83.3KB/s	43.1KB/s	49.9KB/s	115.5KB/s
Read Speed	216.9KB/s	179.8KB/s	170.6KB/s	228.4KB/s

Further Work

Time synchronization of nodes

- Using our Beacon-based approach (~5us max drift)
- Base for communication present

Annotation

- HC interface for user input
- Keypad (chord)

Embedded classifiers

- C4.5 Decision tree
- kNN
- Fuzzy inference system

Thank you!

Questions?