

Large-Scale Human Activity Recognition Using Ultra-Dense Sensing

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Intel Research

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Monitoring Human Physical Activity Is...

... an essential aspect of human life

e.g. care-giving, training, managing, surveillance, directing

... difficult and expensive

highly manual, tedious, intrusive

... beyond conventional computing

... at the limits of computing research

Application 1

Filling Activities of Daily Living (ADL) Forms

<u>Activity Class</u>	<u>Rating (1-5)</u>
Personal Appearance	
Oral Hygiene	
Toileting	
Washing up	
Appliance Use	
Use of Heating	
Care of clothes and linen	
Making a snack	
Making a drink	
Use of phone	
Leisure Activity	
Infant Care	
Medication Taking	
Housework	

shaving, brushing teeth, combing hair, flossing, gargling, applying make-up, bathing, using microwave, baking, blending, watching TV, doing laundry, mending, folding, putting away laundry, adjusting thermostat, making a sandwich, making a chocolate cake, making a martini, making a milkshake, getting a glass of water, phoning friends, phoning family, phoning caregivers, knitting, watching videos, going for a walk, walking the pet, putting grandson to bed, taking blood-pressure medication, taking vitamins, taking calcium, dusting, tidying, cleaning toilets, vacuuming, cleaning blinds, removing cobwebs, ...

Application 2

Managing Best Known Methods (BKMs)

- | Optimal methods confer strategic advantage
 - | machine rooms, operating rooms, burger franchises
- | Managing BKMs involves:
 - | Identifying them
 - | Training people to use them
 - | Monitoring their use
- | e.g. Intel maintenance, UW anesthesiology training



Application 3

Activity-Based Reminders

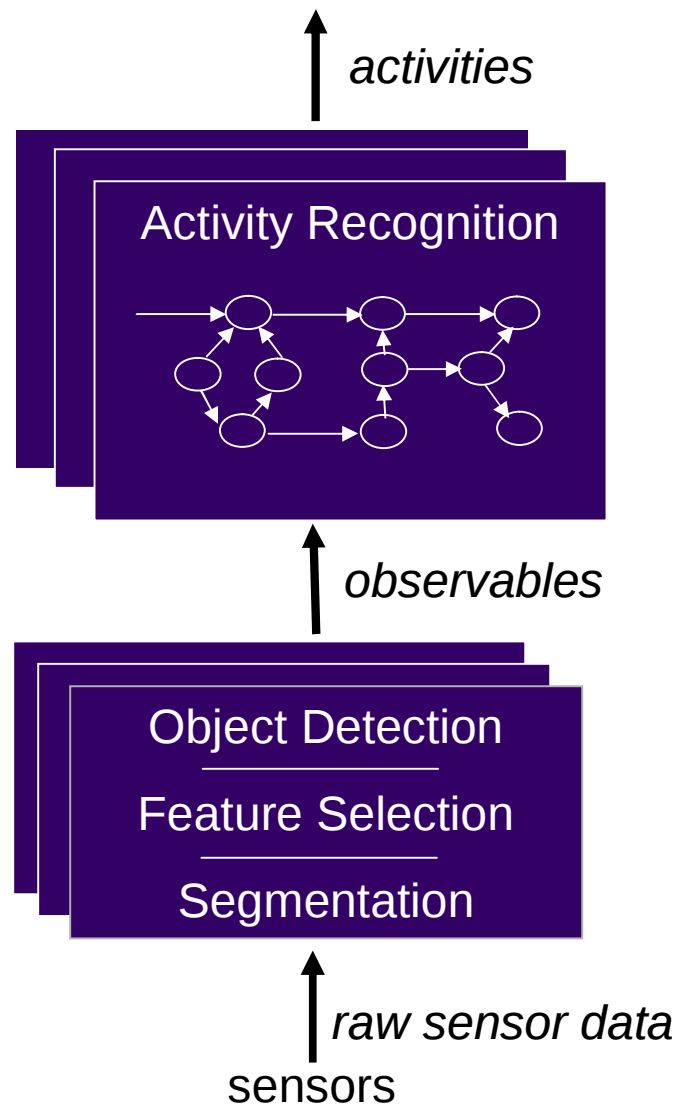
when I eat lunch remind me to take my pill

when I change baby's diapers remind me to apply cream

when I use the bathroom remind me to wash hands

when I use fab oven remind me to run debug sequence

The Traditional Architecture



Four scaling challenges:

1. Designing low-level sensing
2. Designing model schema
3. Creating models
4. Pinpointing anomalies

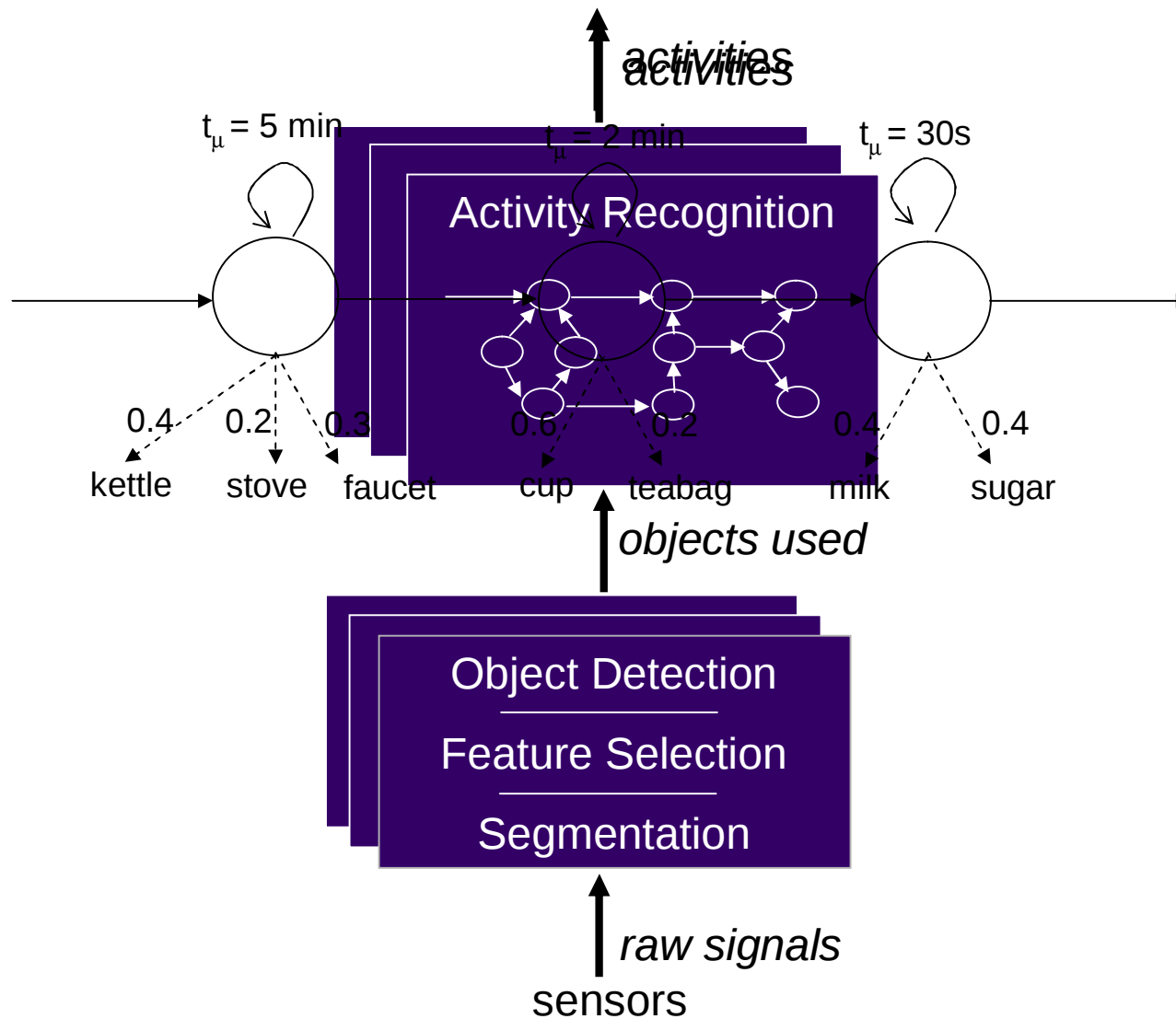
Basic Insight



What you use determines what you do

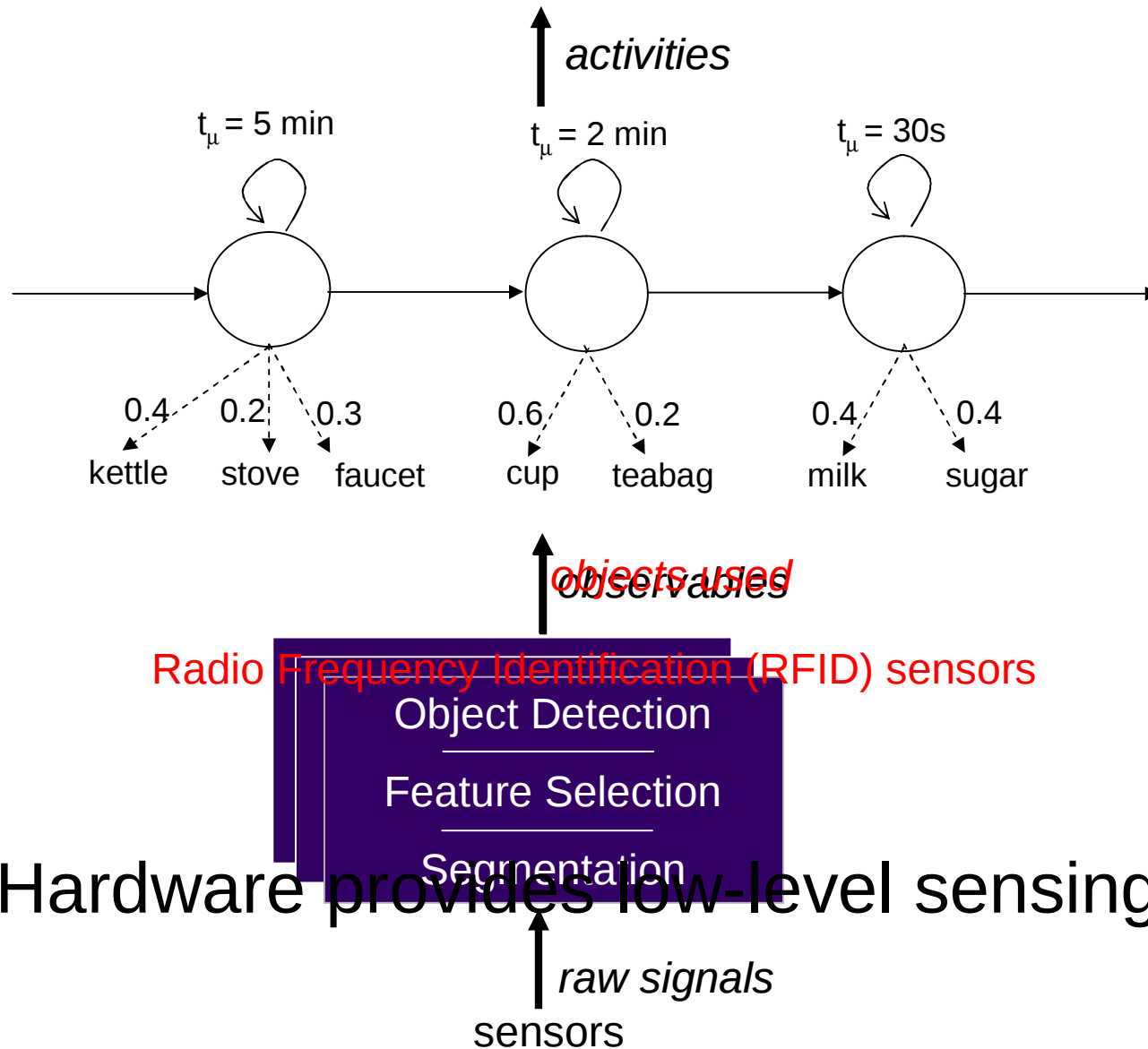
Designing Model Schema

Activities as Object-Use Patterns



Low-Level Sensing

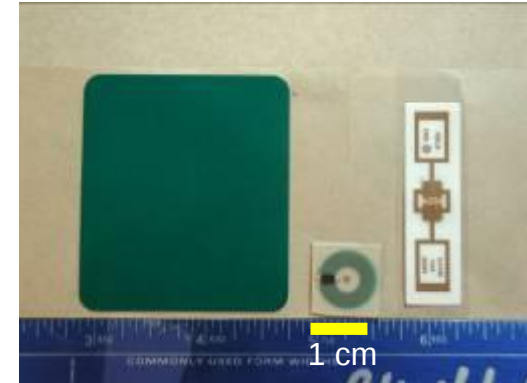
Detecting Object Use



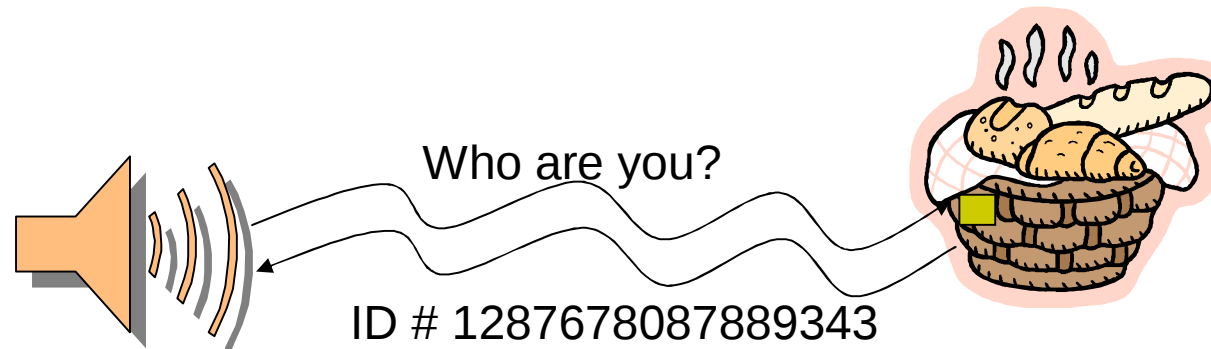
Low-Level Sensing

Radio Frequency Identification (RFID)

Tiny, 40-cent battery-free **tags**
+
ambient **readers**



Tags return 96-bit ID when queried by readers



<ID = e3f000e13431, desc = "bread basket", manufacturer = "...", ... >

Low-Level Sensing iGlove and iBracelet

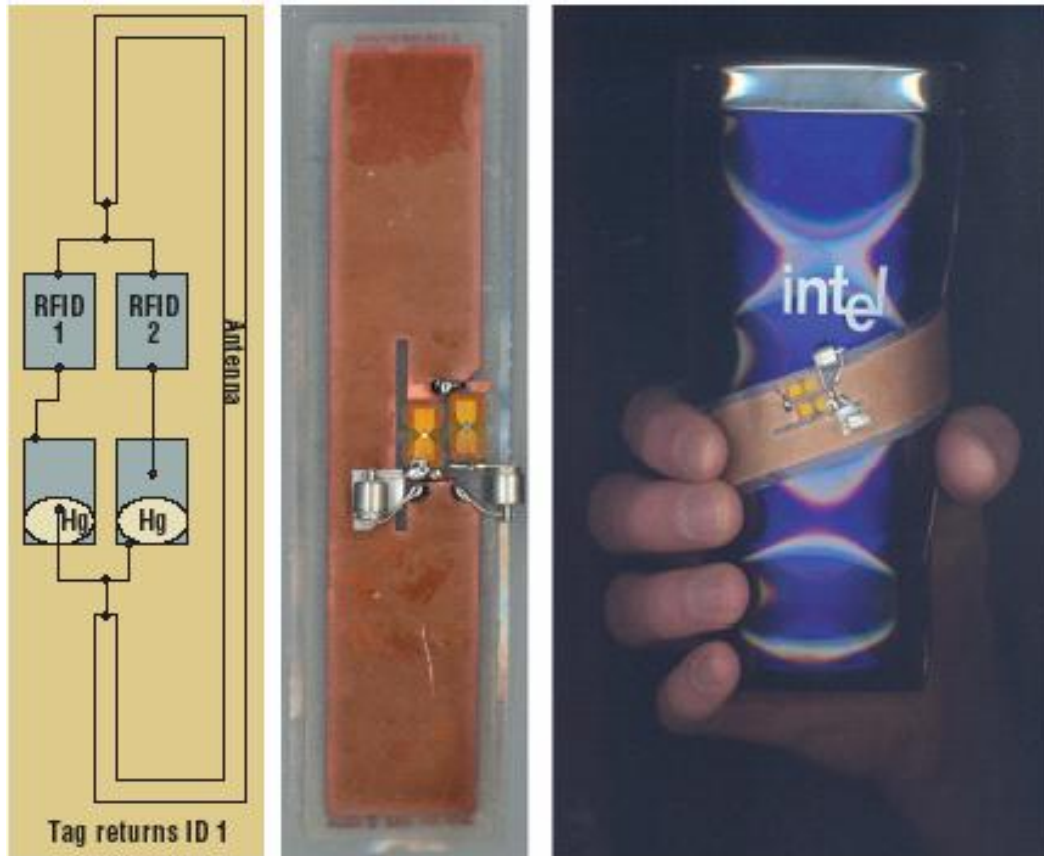


13.56MHz reader, mote, power supply, antenna

Objects tagged on grasping surfaces

Low-Level Sensing

WISP: Wireless ID & Sensing Platform



Passive RFID
+
Sensors

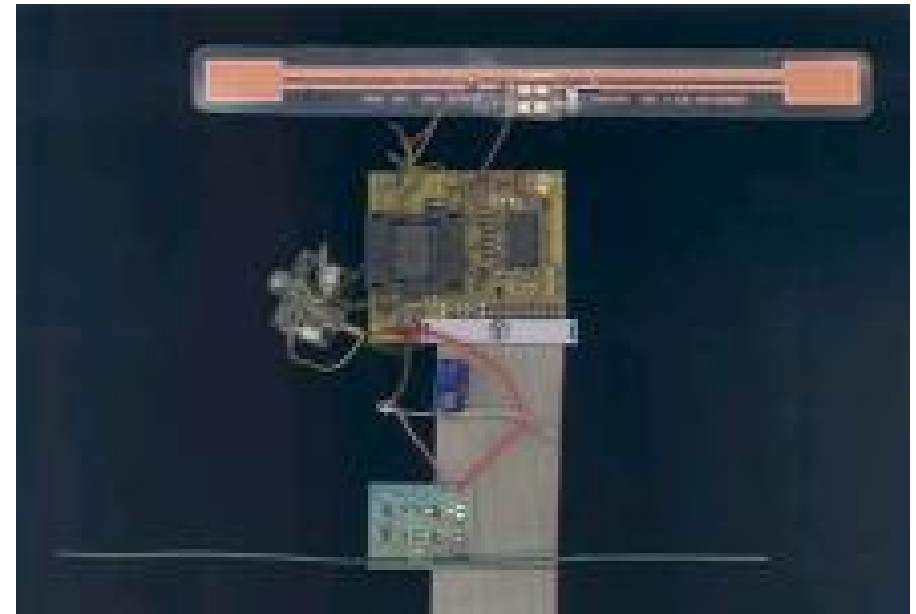
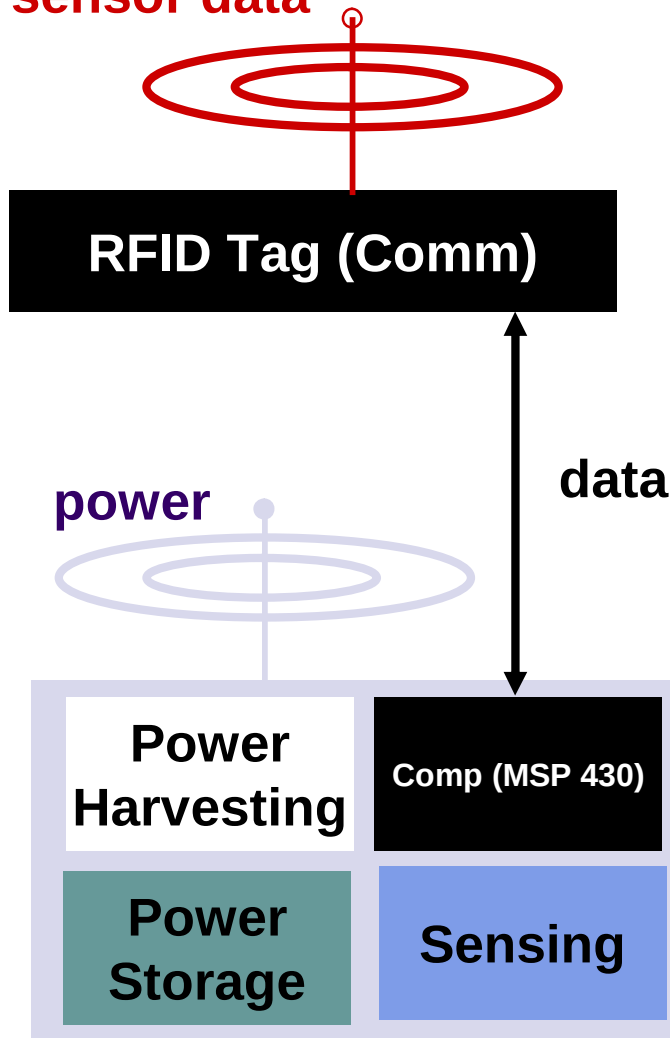
- ü small stickers
- ü battery-free
- ü room-range

α -wisp (motion sensor wisp)

Low-Level Sensing

The WISP Computer

ID, sensor data



- ü multiplexing, aggregation
- ü > 1 min power storage
- ü light sensor

End-to-End Evaluation: ADL Form Filling



- Tagged real home with 108 tags
- **14 subjects** each did **12 of 14 ADL (65 instances)** in arbitrary order
- Subjects wore gloves, recorded activities
- **Given objects touched, infer activities**

#2

Please take care of your oral hygiene as if you are about to go to bed.

The
Toothpaste
Toothbrush
Dental floss



Are in the
bathroom



Simple Models Suffice

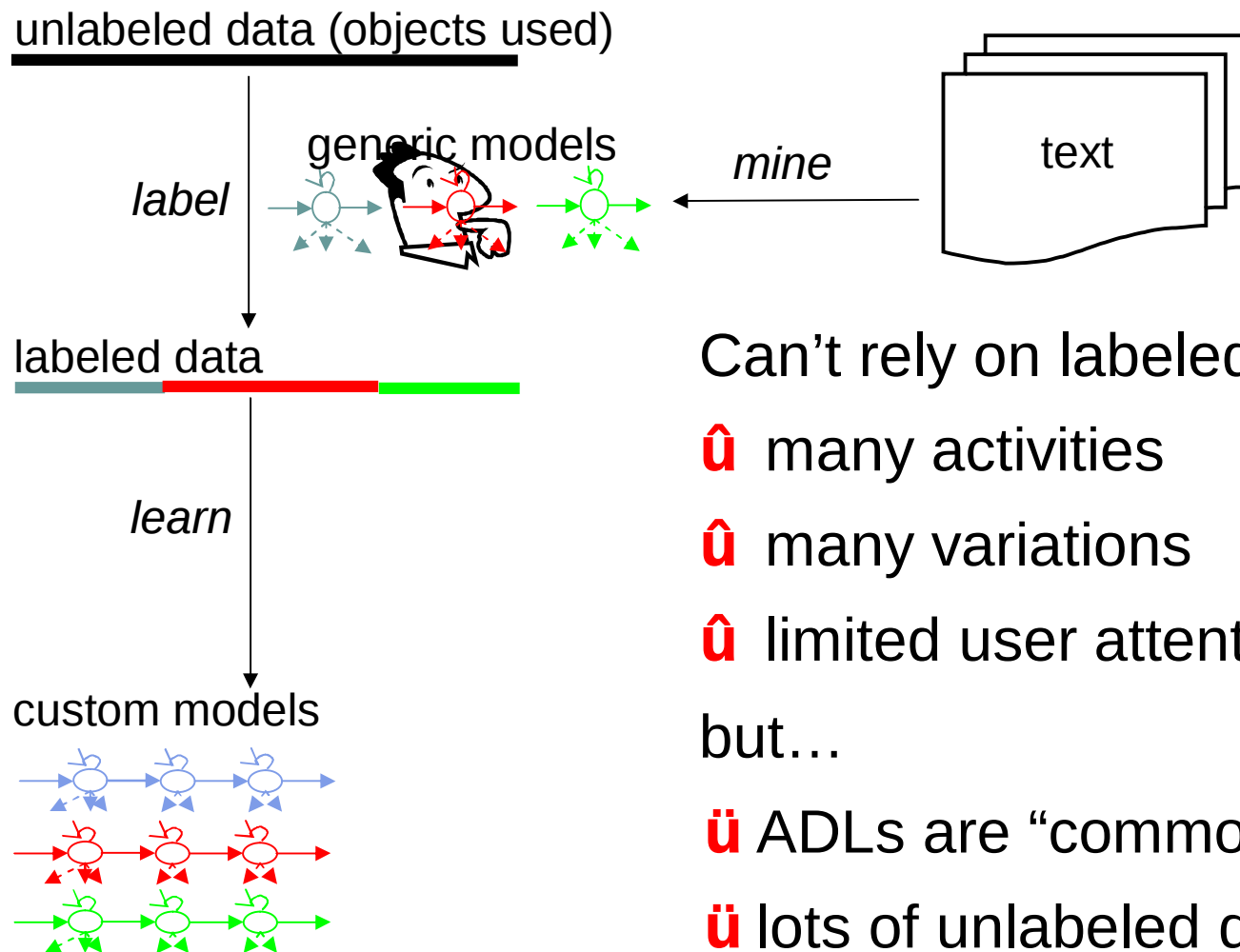
Activity	Prior Work	SHARP (precision/recall)
Personal Appearance		● 92/92
Oral Hygiene		● 70/78
Toileting	●	● 73/73
Washing up	● 95/84	● 100/33
Appliance Use		● 100/75
Use of Heating	●	● 84/78
Care of clothes and linen		● 100/73
Making a snack	● ● ●	● 100/78
Making a drink	●	● 75/60
Use of phone		● 64/64
Leisure Activity		● 100/79
Infant Care		● 100/58
Medication Taking	●	● 100/93
Housework		● 100/82

Legend

● General solution

● Point solution

Creating Models Using Web-Mined Common Sense



Can't rely on labeled data

$\hat{U}$ many activities

$\hat{U}$ many variations

$\hat{U}$ limited user attention

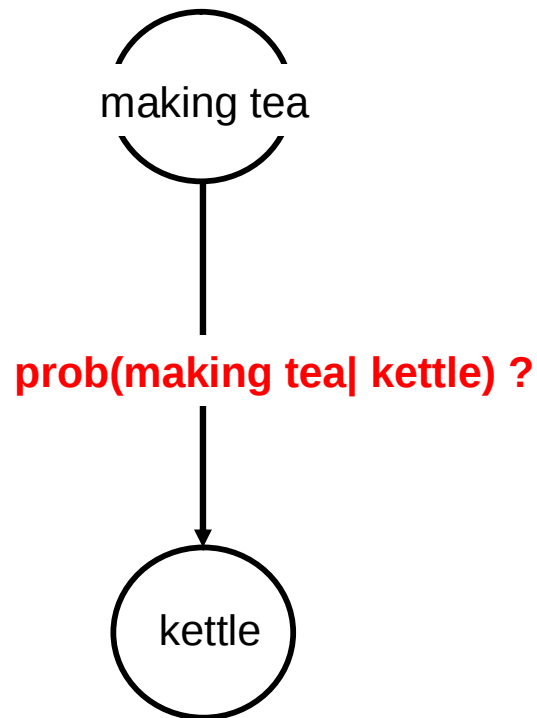
but...

$\bar{U}$ ADLs are “commonsense”

$\bar{U}$ lots of unlabeled data

Creating Models

Shallow, Large-Scale Web Mining



Google("making tea") = 194,000

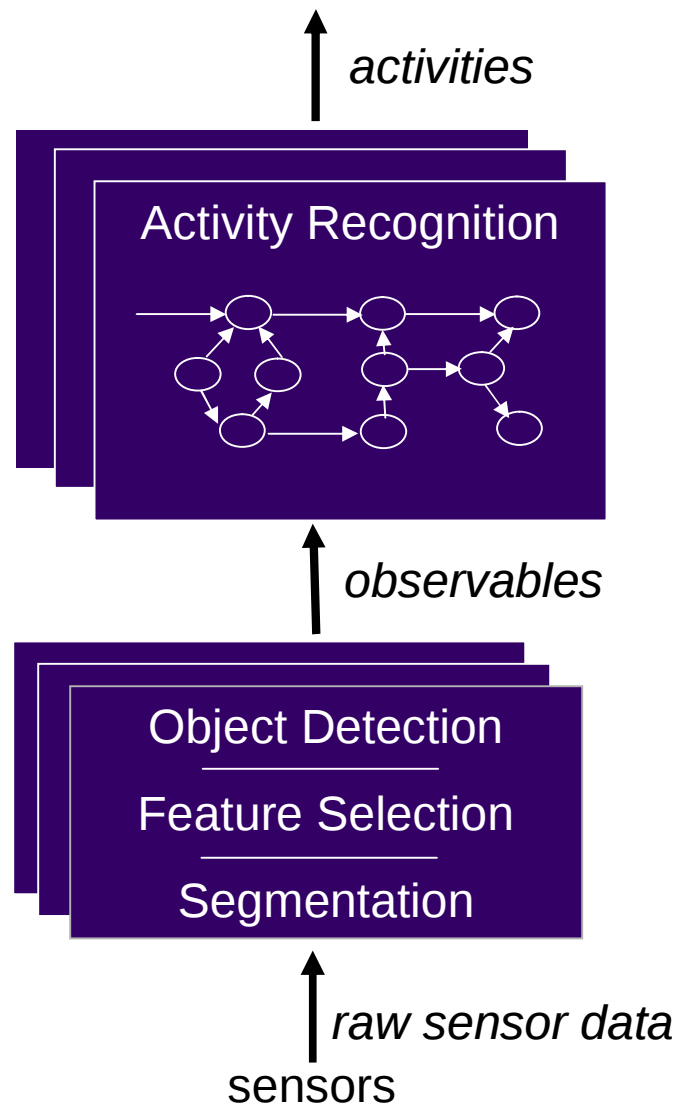
Google(kettle "making tea") = 30,100

prob = $30,100 / 194,000 = 15.5\%$

Performance quite good in practice

40,000 activities modeled in 5 weeks

The Challenges Revisited



Four scaling challenges:

- ü Designing low-level sensing
- ü Designing model schema
- ü Creating models
- ü Pinpointing anomalies

Ultra-Dense Sensing: The Road Ahead

Applications

Physiological monitoring

Structural integrity monitoring

Resource usage monitoring

Quantitative Sociology

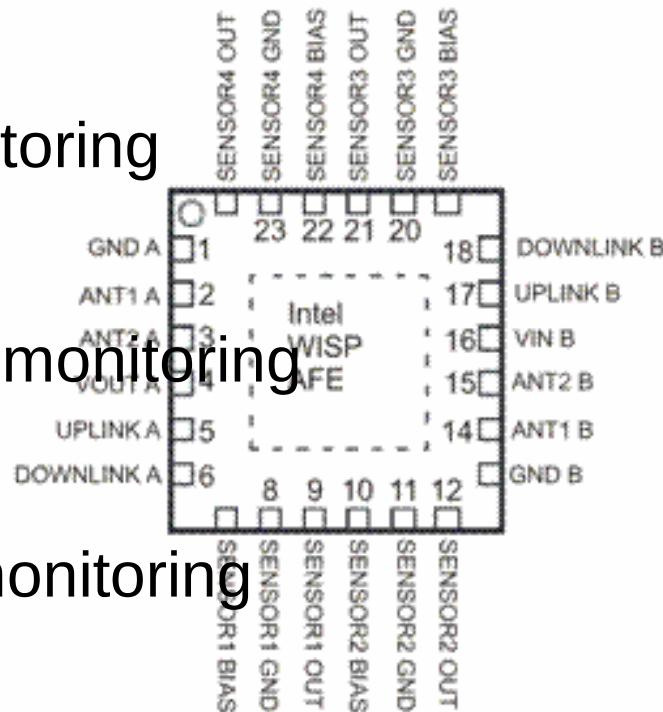
Challenges

range, power

aid manufacturing

ing model

privacy, public acceptance



Ultra-Dense Sensing and Privacy

Variants of privacy

- Share no data

- Avoid sharing data with non-authorized parties

- Share data but control inferences

Practical approach: public policy + education

- Strong legal recourse

- Maximize perceived cost vs. benefit

Conclusions

Monitoring human activity a “disruptive” capability

Mix of enabling technologies within grasp

- Very high density sensing is key

Ultra-dense sensing has implications beyond activity monitoring

Application 2

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Requirements

functionality logging, rating, prompting

scope thousands of activities

timescale seconds to hours

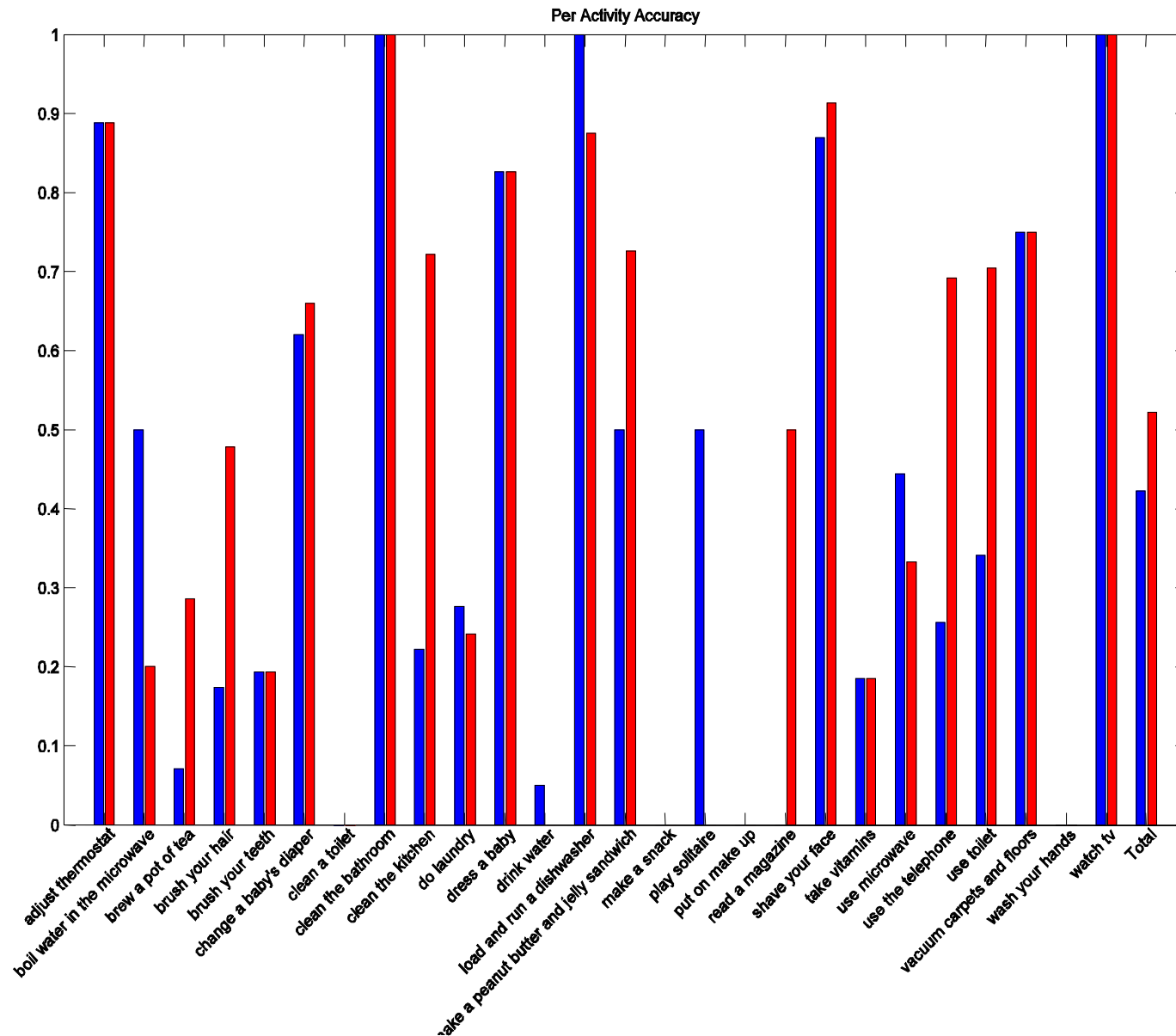
speed interactive time

accuracy moderate to high (>>50%)

resolution detect individual “steps”

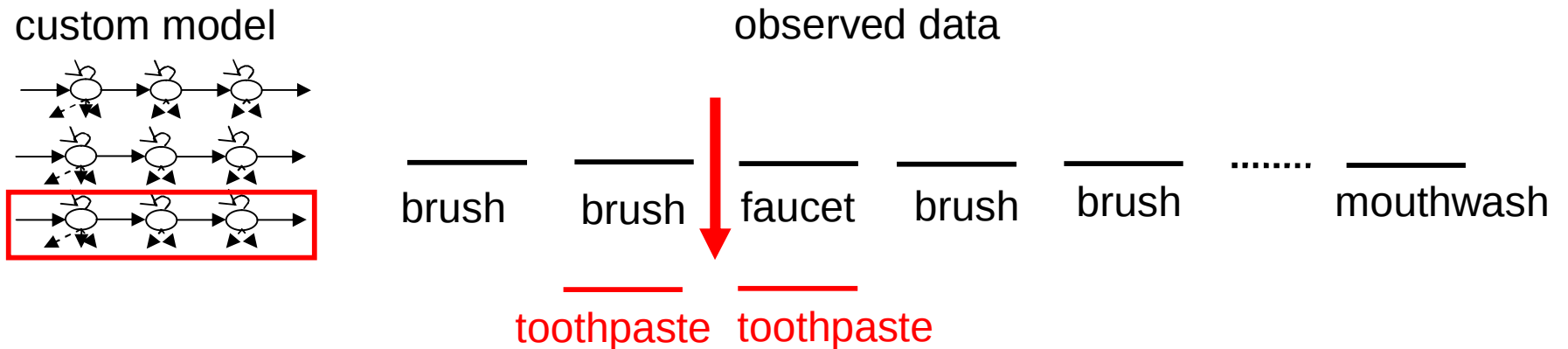
Creating Models

Mined Models Can Bootstrap Learning



Anomaly Detection

Anomaly Localization as Error Correction

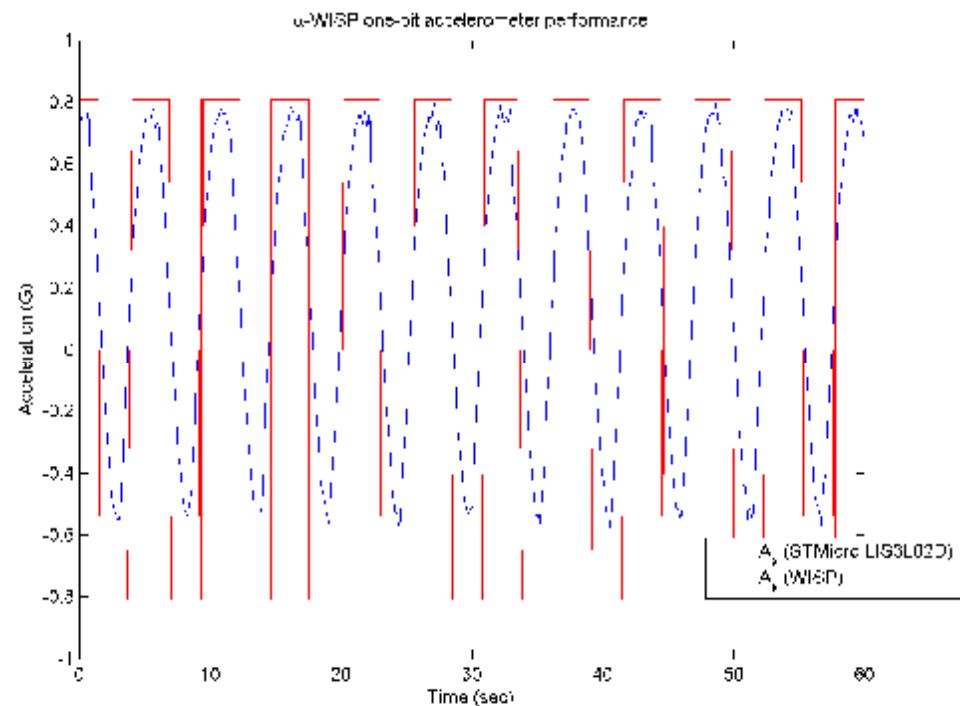
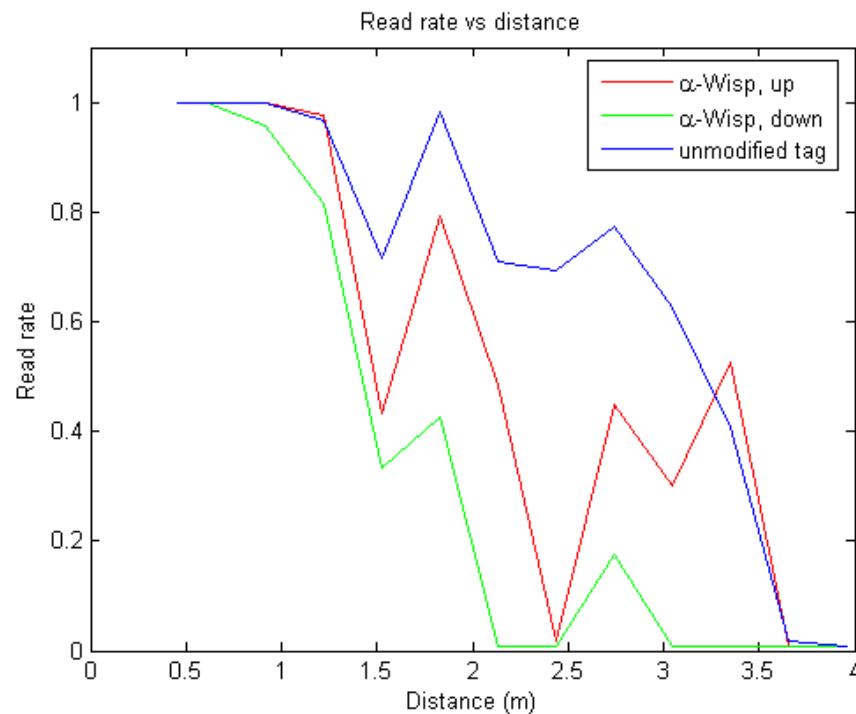


“Error Correcting Viterbi Parsing”

New algorithms fast even without explicit error models

Low-Level Sensing

α -wisp: Read-Range and Response



1-bit accelerometer accurate @ 1-10 feet
new tags @ 10-20 feet