

Actionable Sensing

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Sensors & sensing

- Sensors generate data representative of physical phenomena
 - examples: imagery, acoustic, seismic, acceleration
- Applications often require networks of sensors
 - examples: area, environmental, or building monitoring
- Communication has traditionally been the key problem
 - examples: routing, power, mobile ad hoc networks
- Sensors are increasingly viewed as (low-level) services



An emerging challenge

- Sensor systems are typically stovepipes
 - created for a specific purpose (task)
 - often controlled by a single vendor (stack)
- Example: a home with separate networks for security, entertainment, healthcare, etc
- Challenge: **open sensor/task architectures**
 - Why can't you ask your Kinect box to help keep your home secure, and monitor your health?

The “knowledge management” perspective

- Knowledge management:
 - “Getting the right knowledge to the right people at the right time, in the right place”
- Knowledge is seen as “actionable information”
- **Sensor networks research has been overly-focussed on input (data) at the expense of output (informing human action)**

Actionable sensing

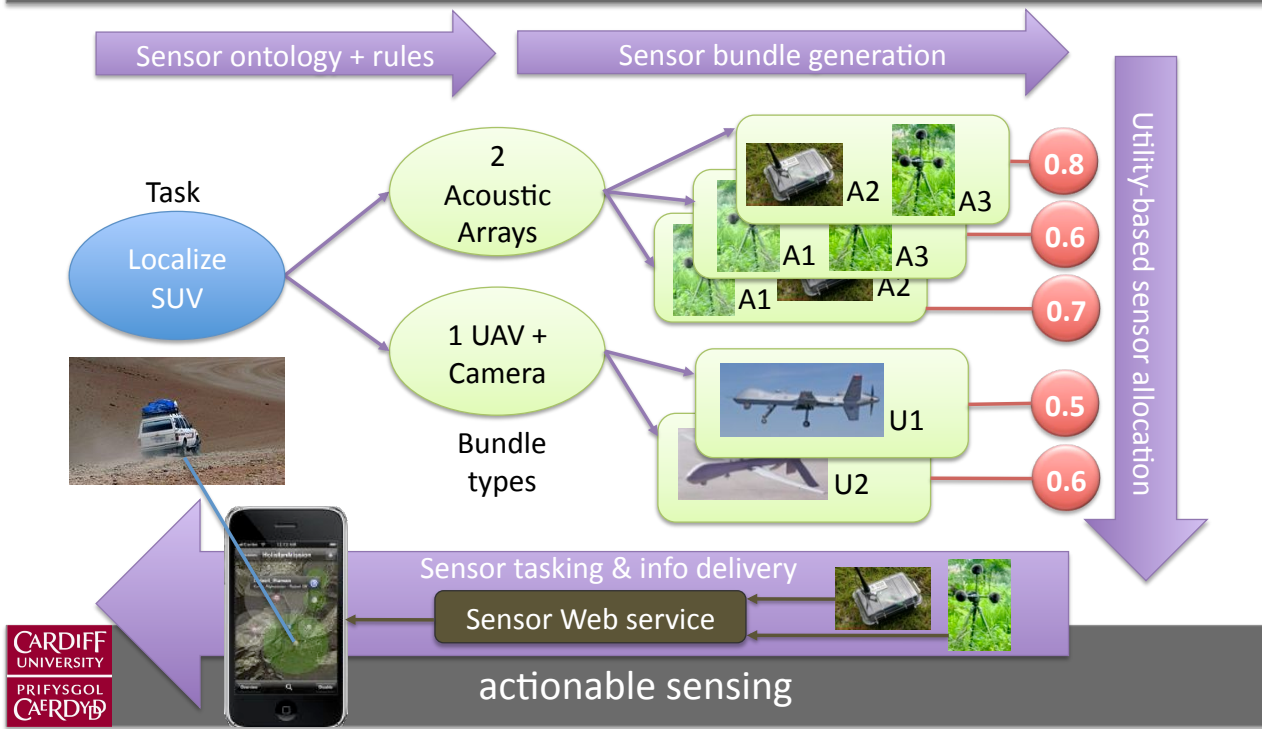
- Organising and optimising the deployment and configuration of a sensor network to the provision of a set of users' tasks
- Key issues:
 - **representation of tasks** at an appropriate level
 - accounting for **value of information**
 - **transparency**: hiding features of the sensor network from users

Forward & backward chains

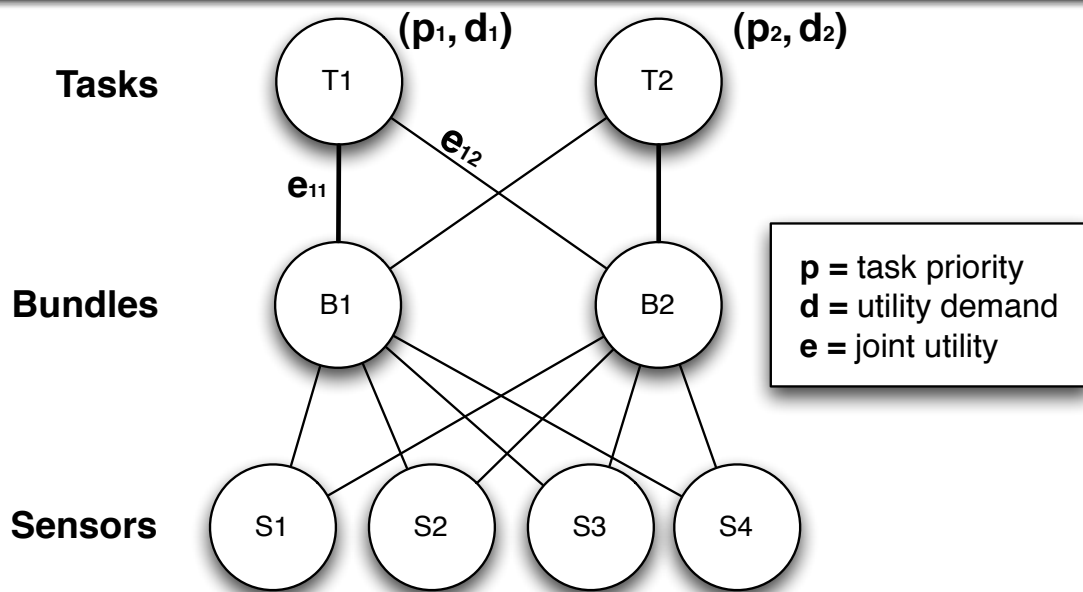


- **Data-to-decision**: a person needs to make a decision based on actionable information from sensors & sources
- **Intent-to-assets**: a person needs to determine what kinds of sensors & sources will help them achieve their intent, and thereby identify suitable assets

Walkthrough



Task-bundle-asset framework



Tasks

- Tasks are triples consisting of:
 - **operation** (defined in an appropriate task ontology)
 - **area-of-interest** (point or region)
 - **time** (instant or period)
- For example, using our NIIRS-based* task ontology, an **operation** is a pair:
 - **operator** (one of: detect, identify, distinguish)
 - **operand** (one or more entity classes)

*National Image Interpretability Rating Scale

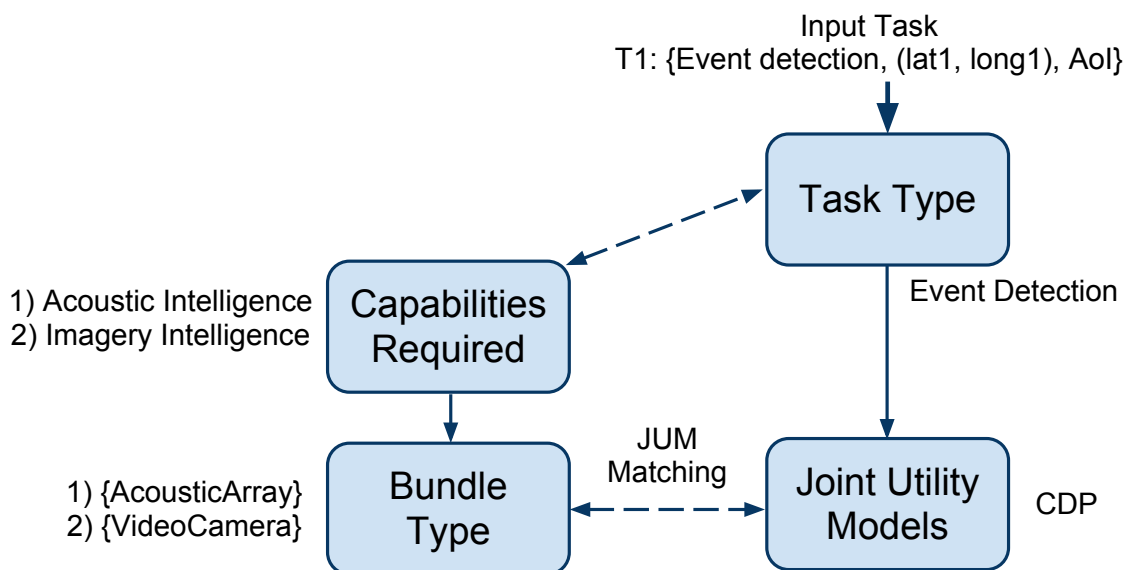
Task examples

- Area monitoring:
 - detect SUV
 - localize vehicle
- Home monitoring:
 - identify person
 - localize Alun
- Environmental monitoring:
 - detect/identify animal

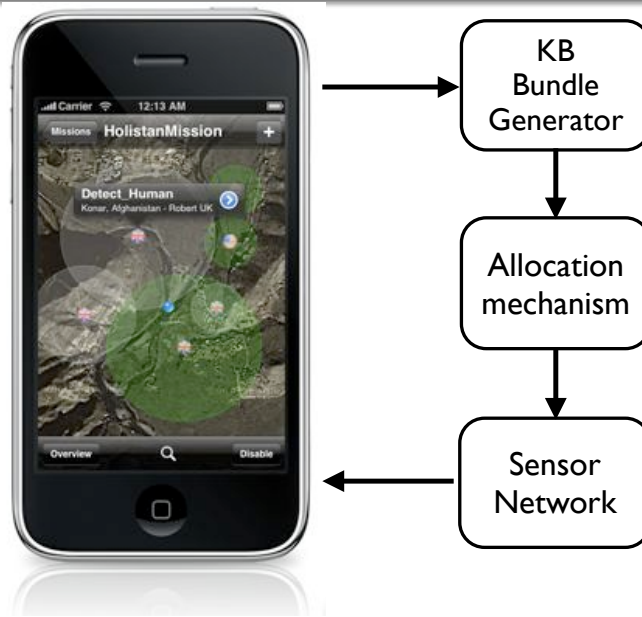
A knowledge-based approach

- **Qualitative knowledge**
 - Web ontology language (OWL) descriptions, rules defines what types of sensor (bundle) are appropriate for which task types
 - examples: vehicle identification can be done visually (“grade 4”) or acoustically (“grade 2”)
- **Quantitative knowledge**
 - joint utility models (functions) determines the value of a set of sensors
 - examples: cumulative detection probability; 2D-localization

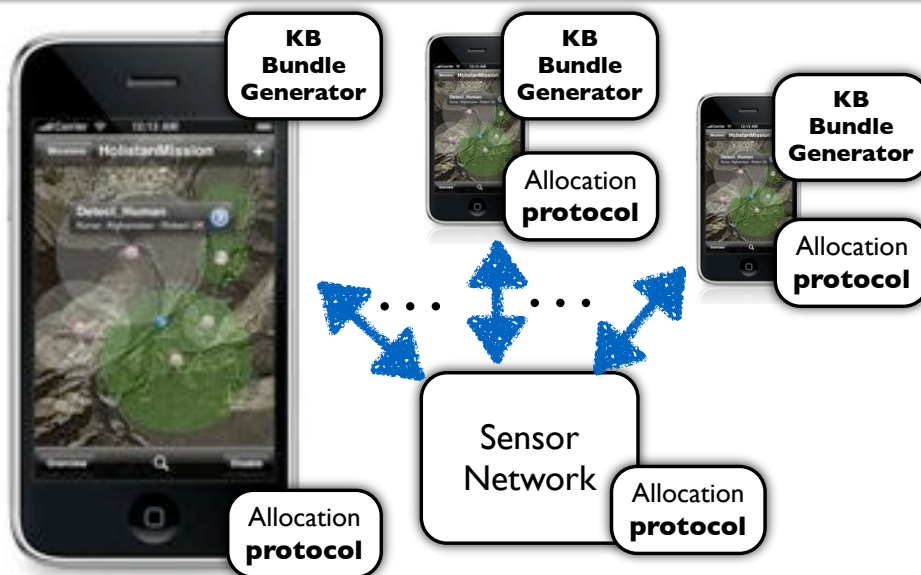
Reasoning procedure



Conceptual architecture



Distributed system architecture



Current work: sensor-task assignment

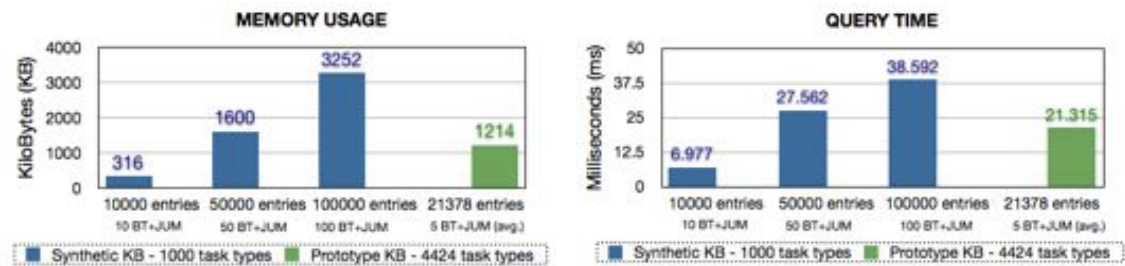
- Performance evaluation of the implemented distributed architecture, to establish
 - feasibility of implementing reasoner (knowledge-based bundle generator) on mobile device
 - relative effectiveness of a variety of task preemption strategies
 - costs of distributed approach in terms of messages exchanged
- (Comparison with a centralised architecture will follow...)

Lightweight reasoner: lookup table

Task Type	Recommendation
1	$(BT_1 + JUM_1)$
1	$(BT_2 + JUM_2)$
2	$(BT_3 + JUM_1)$
2	$(BT_2 + JUM_1)$
2	$(BT_2 + JUM_2)$
...	...
N	

A complete set of NIIRS tasks requires approximately 4500 task types; in practice there appear to be few BTs & JUMs.

Feasibility of mobile implementation

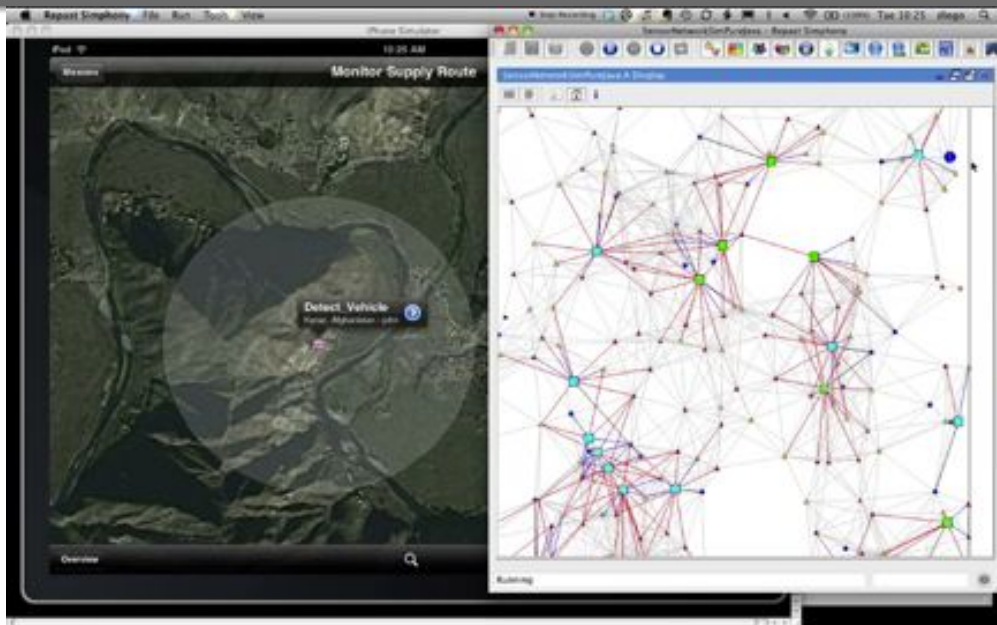


Memory usage and query times are manageable and grow linearly as lookup table size increases (tested on iPod Touch).

Experimental testbed

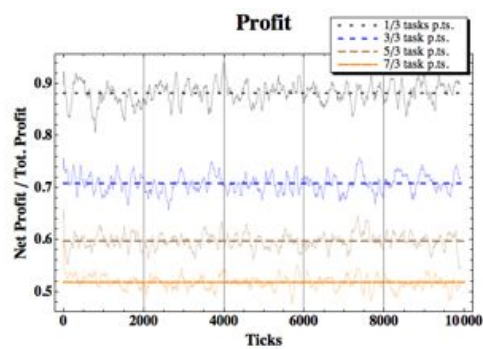
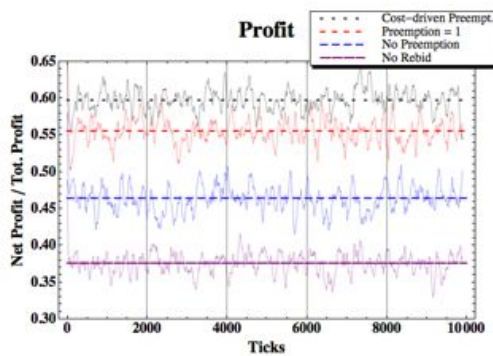
- Implemented distributed coalition formation protocol in Java using Repast Symphony 2.0
 - Extended [Shehory & Kraus, 1998] with preemption, rebidding, & deadlines
- Deployed
 - 250 sensor nodes of 2 types
 - 50 user nodes
- Varied task creation rate from 1 to 7 tasks every 3 timesteps
- Task priority & utility demand generated with uniform random distribution

Demo



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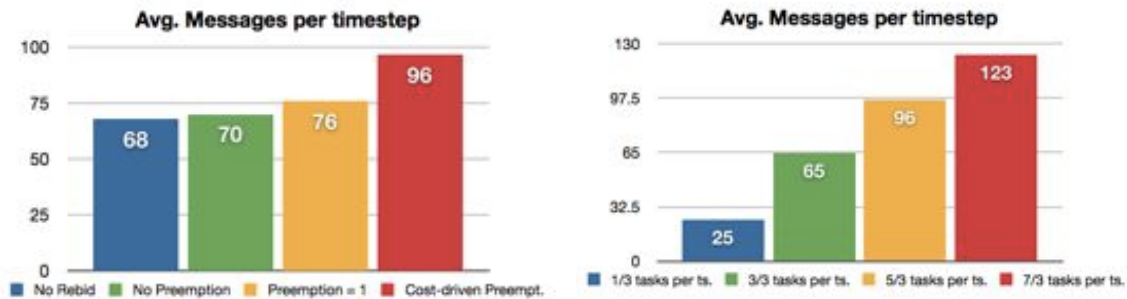
Results: task satisfaction



Profit significantly better with full preemption & rebidding;
performance degrades with higher task arrival rates.

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Results: messages exchanged



Message rates not excessive, though full preemption more costly; message rates increase linearly with task arrival rates.

Current work: information delivery

- Shifting the focus to “forward chaining”:
 - **data to information to decision**
- Collaboration with SOHCS (physiotherapy)
 - use RFID “factory grade” technology to monitor patients’ movements (wrist & ankle tagging)
 - create an open sensing platform based on Semantic Web ontologies & rules
 - proof-of-concept: generate equivalent quality-of-information as current clipboard-based monitoring



The need for “embedding”

- Sensor networks are tools; they need to be usable by end-users
- Fitting sensor network technology into the context-of-use means understanding that context
- We need to train “**embedded sensorsticians**”*
- Example: rainforest “local knowledge” project with BIOSI



image from *Jungle Times* Dec/Jan 2010/11

*“sensorstician” © M Williams

Conclusion & future work

- Fashionable to talk about “internet of things”
... as distinct from “internet of people” ...
Challenge: creating an **internet of people & things**
- Our work is aiming to connect sensors & people through semantic models
 - make it easier for users to find and exploit sensors
 - make sensors & services (re)usable for a range of tasks

Social sensor informatics?

- People are often not deciding/acting/intending in isolation
 - sensors & sources need to be shared
 - locally or globally
 - cooperatively or competitively
- People are sensors too
 - they can provide data (“HUMINT”)
 - and metadata (local knowledge)

Questions?

Thanks for listening!

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