

Optimizing Sensor Data Acquisition for Energy-Efficient Smartphone-based Continuous Event Processing

Lipyeow Lim

University of Hawai`i at Mānoa

Archan Misra

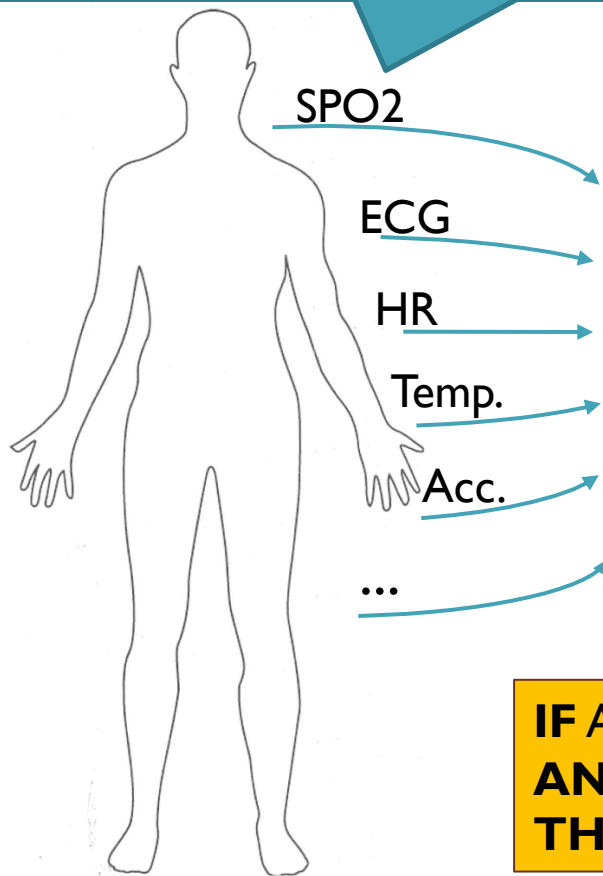
Singapore Management University

Telehealth Scenario

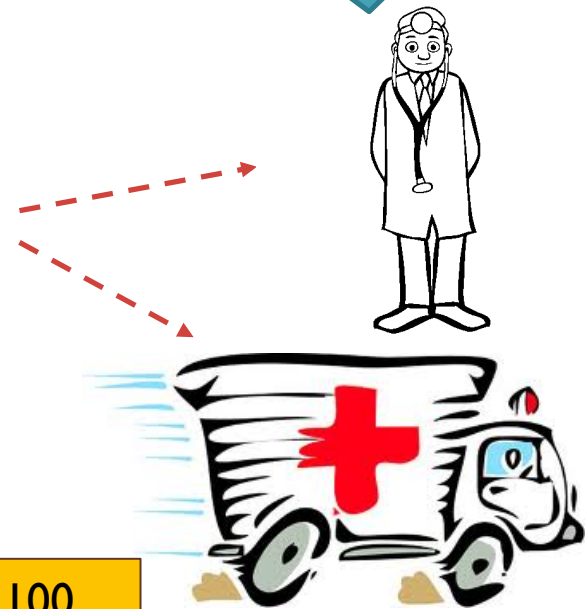
Wearable sensors transmit
vitals to cell phone via wireless
(eg. bluetooth)

Phone runs a complex event
processing (CEP) engine
with rules for alerts

Alerts can
notify
emergency
services or
caregiver

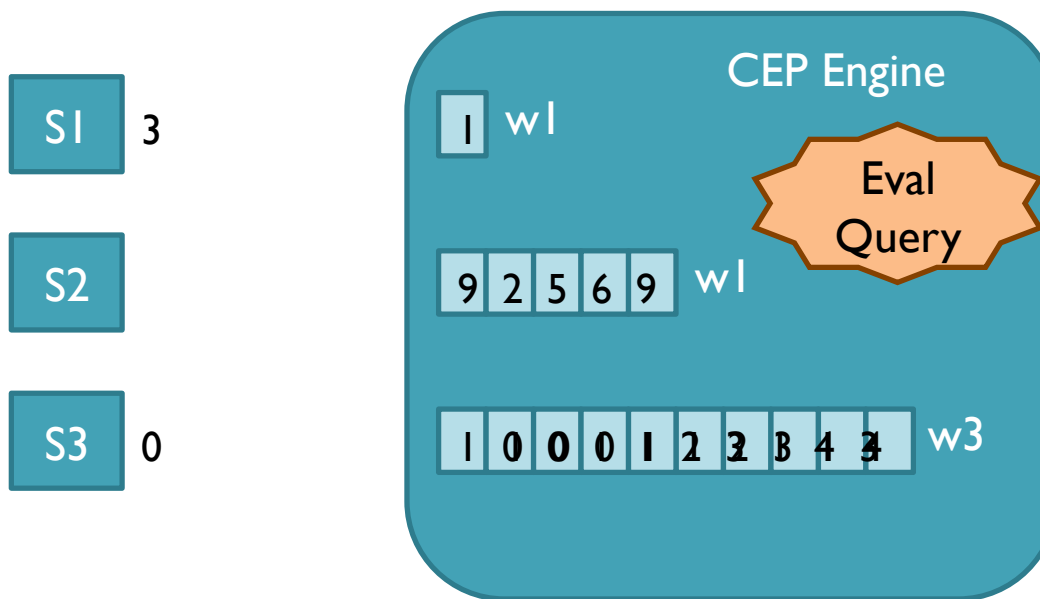


IF Avg(Window(HR)) > 100
AND Avg(Window(Acc)) < 2
THEN SMS(doctor)



Continuous/Streaming Evaluation

if $\text{Avg}(S2, 5) > 20$ AND $S1 < 10$ AND $\text{Max}(S3, 10) < 4$ then email(doctor).

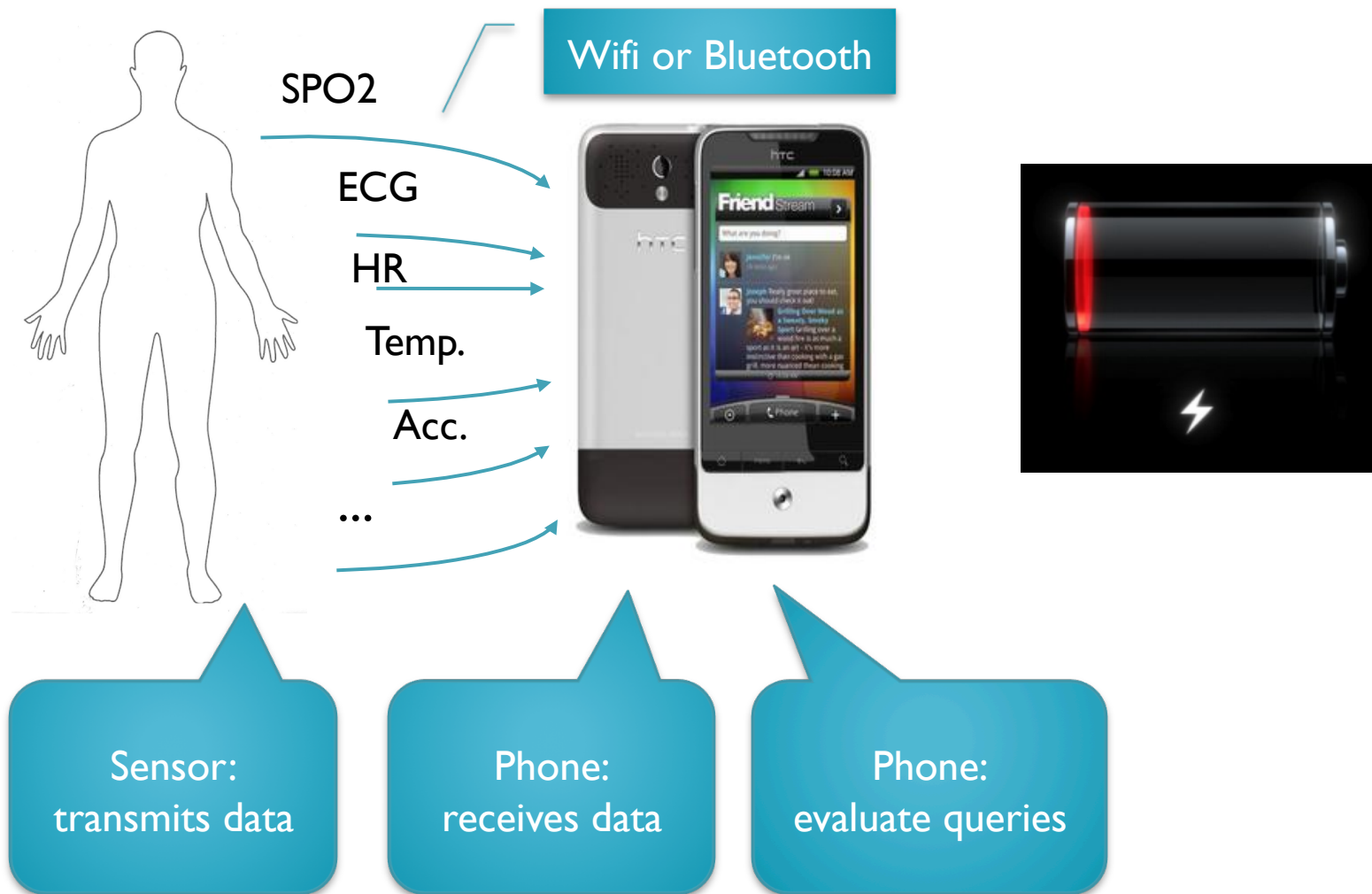


Algorithm

When t_i of S_i arrives
Enqueue t_i into W_i
If Q is true,
Then output alert

“Push”
model

Energy Consumption



Research Question



Is there a better way to perform such complex event processing that

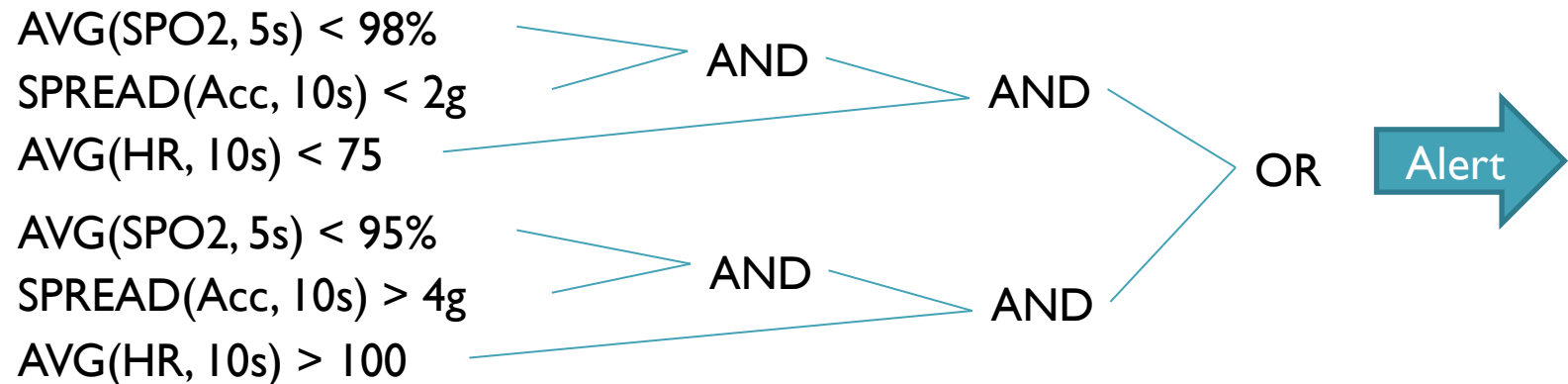
- Minimizes energy consumption at the phone, and/or
- Maximizes operational lifetime of the system.

Key Ideas

- Pull model
 - Evaluate a query every ω seconds
 - Acquire only data that is needed
- Evaluation order of predicates matter!
 - Shortcircuiting can avoid data acquisition
- Batching

Assuming fairly smart sensors capable of buffering and supporting “pull”

Query Model



- A **query** is a boolean combination of predicates
- Predicates
 - **Aggregation functions** over a **time-based window** of sensor data

Sensor Data Acquisition

3D acc.
ECG,
EMG, GSR



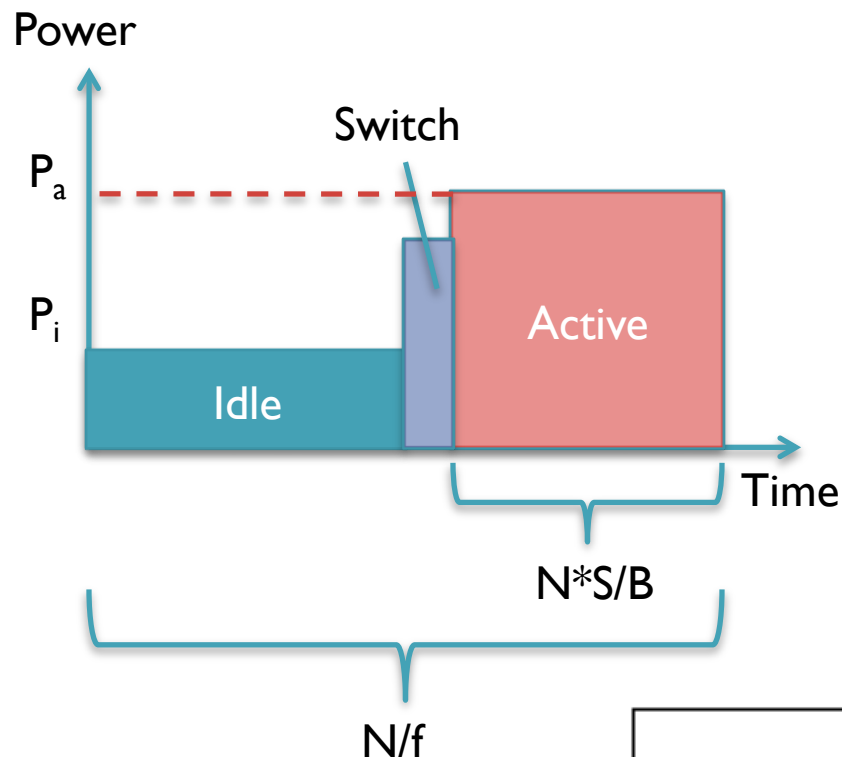
Bluetooth
Or 802.11
Or 802.15



- Constant sampling rate
- 802.11 (wifi) uses 2 power modes: active, idle
- Bluetooth has 3 modes: active, idle, sleep (not relevant).
- Time needed to switch modes
- Energy expended to switch

Sensor Type	Bits/sensor channel	Channels/device	Typical sampling frequency (Hz)
GPS	1408	1	1 Hz
SpO2	3000	1	3 Hz
ECG (cardiac)	12	6	256 Hz
Accelerometer	64	3	100 Hz
Temperature	20	1	256 Hz

Pulling N Tuples from Sensor

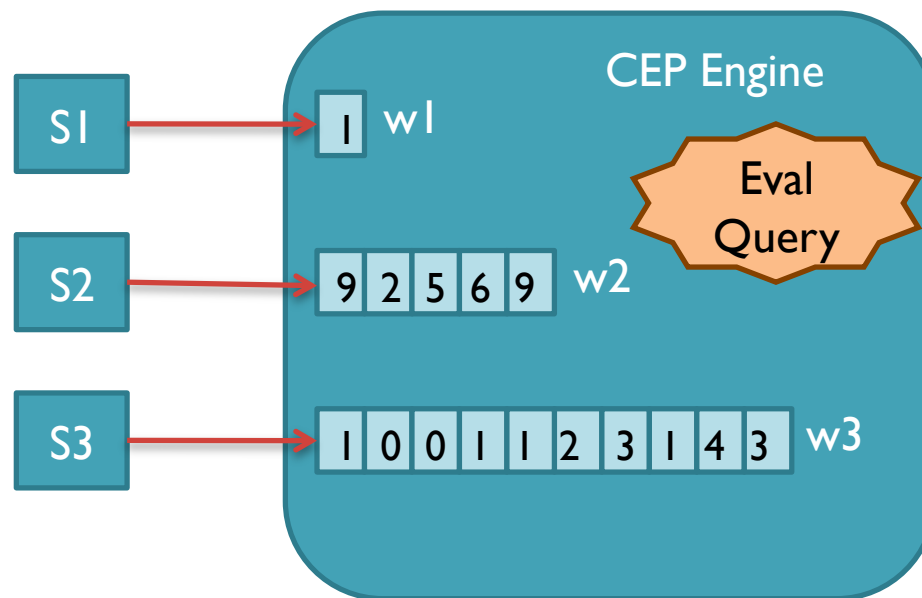


- Idle mode consumes P_i mW
- Active mode consumes P_a mW
- Sensor rate is f Hz
- A tuple is S bits
- Bandwidth is B Mbps

	IEEE 802.11	Bluetooth 2.0+EDR
P_a	947 mW	60mW
P_i	231 mW	5 mW
B	54 Mbps	1 Mbps
E_{switch}	14 μ Joule	–
T_{idle}	100 ms	–
T_{switch}	–	6 msec

Pull-based Evaluation

if $\text{Avg}(S2, 5) > 20$ AND $S1 < 10$ AND $\text{Max}(S3, 10) < 4$ then email(doctor).

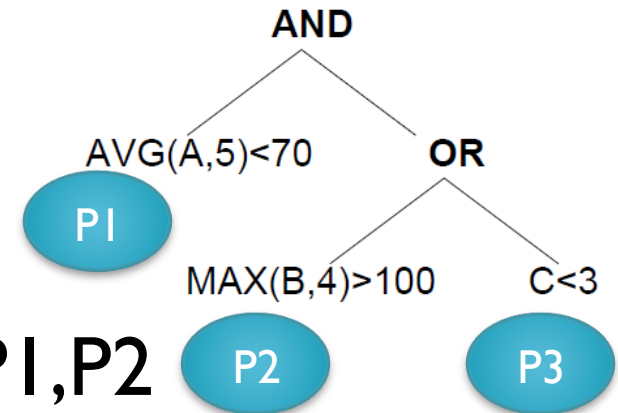


Pull

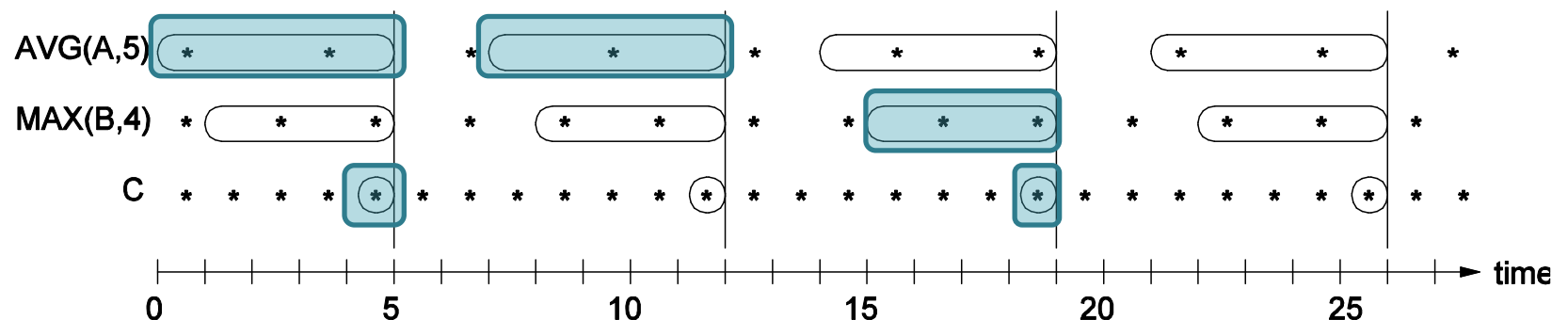
Loop every ω seconds
For each sensor S_i
 Acquire data for S_i
 Enqueue data into W_i
EndFor
If Q is true,
 Then output alert
End loop

- Complex interaction between ω , stream rates, and predicate windows
- If predicate $S1 < 10$ is false, why bother to acquire data for S2 and S3?

Example: $\omega=7$



- **Time 5:** eval order is P3,P1,P2
- **Time 12:** eval order is P1,P2,P3
- **Time 19:** eval order is P2,P3,P1



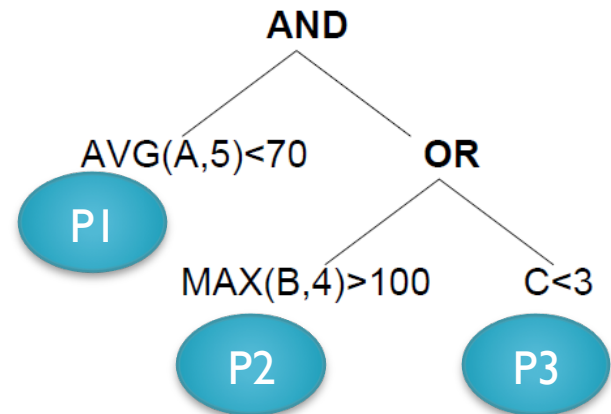
Evaluation Order

if $\text{Avg}(S2, 5) > 20$ AND $SI < 10$ AND $\text{Max}(S3, 10) < 4$ then email(doctor).

Predicate	$\text{Avg}(S2, 5) > 20$	$SI < 10$	$\text{Max}(S3, 10) < 4$
Acquisition	$5 * .02 = 0.1 \text{ nJ}$	0.2 nJ	$10 * .01 = 0.1 \text{ nJ}$
Pr(false)	0.95	0.5	0.8
Acq./Pr(f)	0.1/0.95	0.2/0.5	0.1/0.8

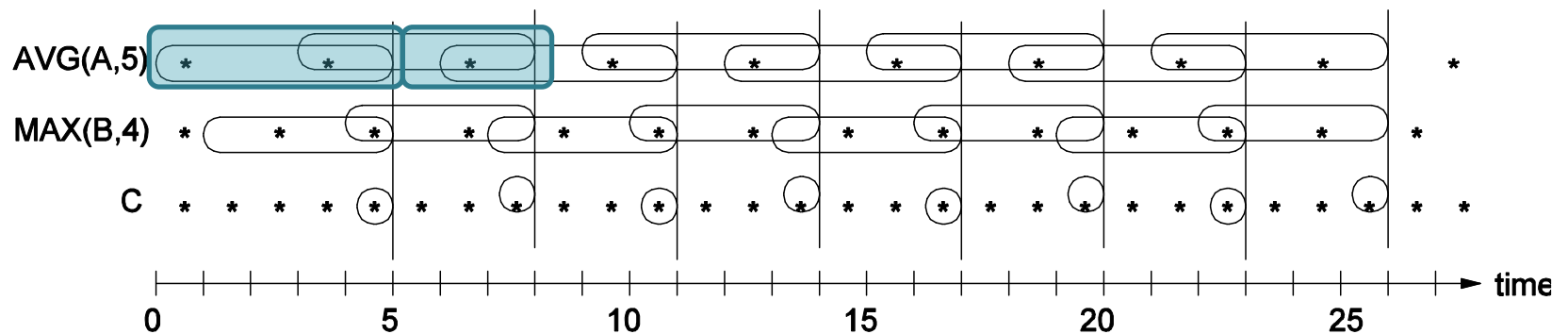
- Evaluate predicates with lowest energy consumption first
- Evaluate predicates with highest false probability first
- Evaluate predicate with lowest normalized acquisition cost first.

Example: $\omega=3$



- **Time 5:** P1,P2,P3
- **Time 8:** acquisition cost for A becomes cheaper, because some tuples are already in buffer

Acquisition cost depends on state of the buffer at time t



Algorithm Sketch

At each ω

1. Calculate normalized acquisition cost (**NAC**) based on **buffer state** and **$P(\text{pred}=\text{true})$**
2. Find evaluation order using **NAC**
3. Acquire sensor data and eval pred using eval order with shortcircuiting.

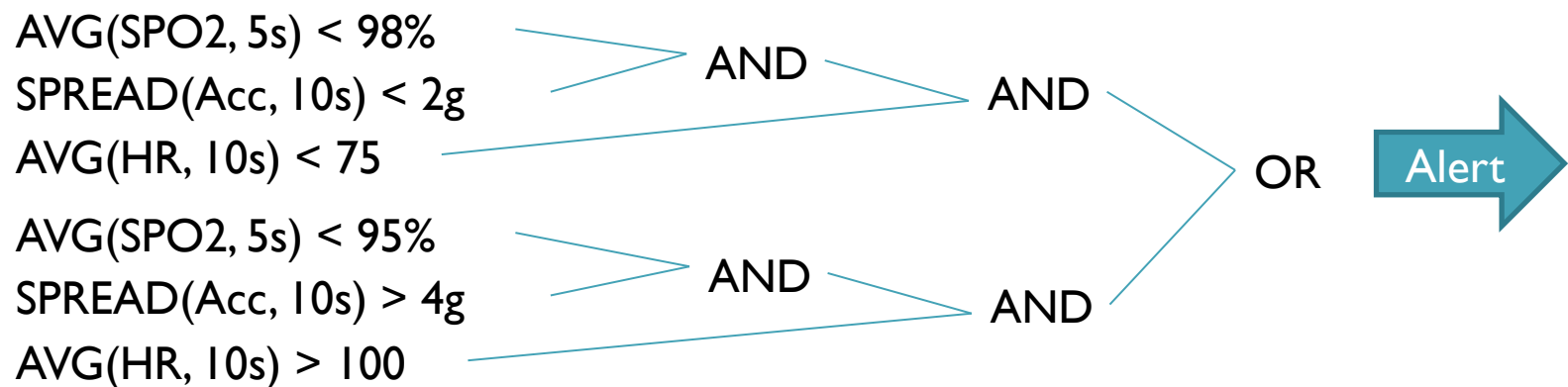
What happens if >2 predicates operate on the same sensor data stream?

Simulation Setup

- **Naive**
 - data from all sensors acquired in batches
- **ASRS-static**
 - Evaluation order determined once at initialization and never changes
- **ASRS-dynamic**
 - Evaluation order determined at each ω time period.
- Simulation results averages 5 1-hour traces with 95% confidence intervals.
- **P(pred=true)** distributions obtained from half the data streams themselves

Simulation Data & Query

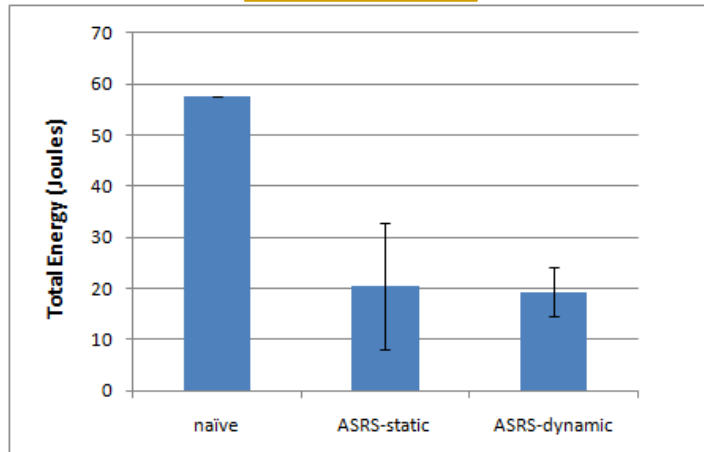
- Data streams generated using independent Gaussian distribution
 - SPO2 $\sim N(96,4)$, 3 Hz, 3000 bits
 - HR $\sim N(80,40)$, 0.5 Hz, 32 bits
 - Accel $\sim N(0,10)$, 256 Hz, 196 bits



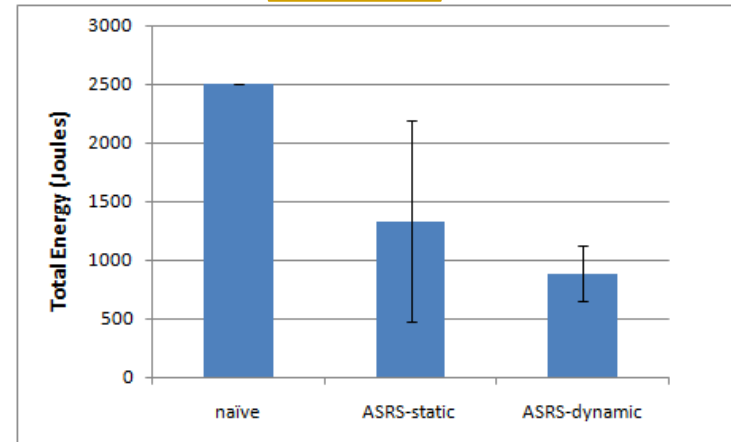
Simulation Results

Energy

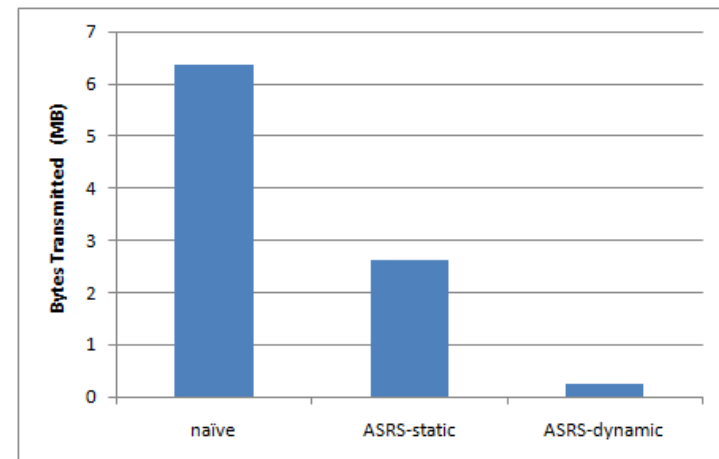
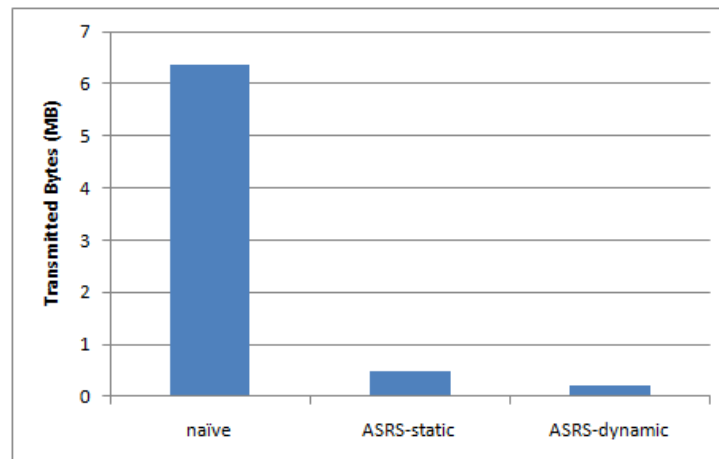
Bluetooth



802.11



Bytes





Conclusion

- Pull-based processing paradigm can have a significant impact on data acquisition energy consumption
- Ordered evaluation of predicates can help shortcircuit the evaluation and avoid costly data acquisition
- We proposed evaluation algorithms based on these two observations to minimize data acquisition cost at CEP engine
- Results on synthetic traces show that savings up to 70% are possible.

Future Work

- Improve simulator
 - Disjunctive normal form query representation
 - More realistic data generators
- Estimation algorithms for $P(\text{pred}=\text{true})$
- **Batching**: wait say 3ω before query evaluation
- End-to-end evaluation on **Android** phone
 - Maximize operational lifetime of phone+sensors