



School of Computing Science





Optimal Stopping of the Context Collection Process in Mobile Sensor Networks

Christos Anagnostopoulos¹, <u>Stathes Hadjiefthymiades</u>², Evangelos Zervas³

¹ University of Glasgow, School of Computing Science, UK ² National and Kapodistrian University of Athens, Dept. of Informatics & Telecommunications, Greece ³TEI Athens, Dept. Electronics, Greece

The considered setting

Consider a Mobile Sensors Network (MSN) with

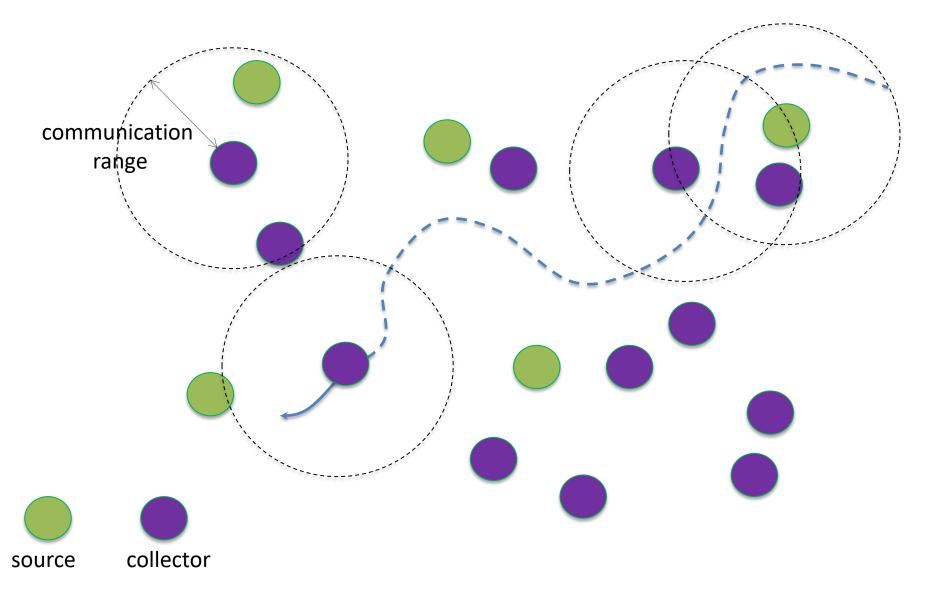
sources, i.e., sensors that produce *context* (e.g., humidity)

collectors, *i.e.*, mobile nodes that receive, store and forward context to their neighbors.

Context is **quality-stamped**, *e.g.*, freshness. The context *quality indicator* **decreases** with time.

The **aim** of a collector is to **gather** as many high-quality pieces of context as possible from sources and/or collectors.

The considered setting



The problem

The **collectors** in a MSN:

- *forage* for high quality context and, then, *deliver* it to mobile context-aware applications;
- Indergo a context collection process by exchanging data with neighboring collectors and/or sources in light of receiving context of <u>better</u> quality and/or <u>new</u> context;
- *cannot prolong* this process forever, since context quality decreases with time, thus, delivered context might be unusable for the application.

Some definitions

Context *c* is represented as:

$$c = \langle p, u, x_u \rangle$$

where:

p is contextual parameter/type (*e.g.*, temperature), *u* is contextual value (*e.g.*, 30 °C), x_u is quality indicator of value *u*

Context quality indicator

Indicator $x_u \in [0, 1]$ indicates *freshness* of *u*.

 $> x_{\mu} = 1$ indicates that μ is of maximum quality. $> x_{\mu} = 0$ indicates that μ is unusable.

Context at time *t* is called **fresh** if $x_u(t) > 0$; otherwise it is called **obsolete**.

Context quality indicator

Indicator $x_{\mu}(t)$ at time t > 0 is updated as follows:

$$x_{u}(t) = x_{u}(t-1) - 1/z$$
, $x_{u}(0) = 1, z \neq 0$

- z is the validity horizon for parameter p in which value
 u is considered usable.
- > z is application specific, e.g., z = 10min if p is temperature, z = 1min if p is wind-speed.

Notice: Alternative quality indicator functions can be, for instance, the inverse exponential function

Rationale

Consider a **collector** which has collected a set of N **fresh** pieces of context, $C = \{c_1, c_2, ..., c_N\}$, referred to as *local context*.

Let collector receive context q from a neighbor collector.Collector increases its local context in type and/or quality as follows:

- > If q is obsolete then collector <u>discards</u> q;
- For replaces definition I = 1 For q is fresh and there is some local context c with the same type of q but less fresh than q then collector <u>replaces</u> c with q;
- If q is fresh and there is no other local context of the same type, collector <u>inserts</u> q into C;

Degree of completeness

Local context *C* is quantified through **degree of completeness (DoC)**, *Y*, defined as the random variable [1]:

$$Y = N \cdot \sum_{k=1}^{N} X_k$$

- N is the *current* number (quantity) of collected pieces of context; N ∈ {0, 1, 2, ..., m}, m > 0.
- $\succ X_k$ is the *current* **quality** indicator of the *k*th contextual parameter in *C*.
- [1] C. Anagnostopoulos, S. Hadjiefthymiades, 'Multivariate context collection in mobile sensor networks', Computer Networks, Elsevier, 57(6):1394–1407, April 2013

Degree of completeness

- When the collector decides to **stop** the collection process at some time, it wants to achieve the highest expected value of Y.
- Hence, the collector has to find an optimal stopping time *t* of the collection process which maximizes:

$$E[Y_t] = E[N_t \cdot \sum_{k=1}^{N_t} X_k^t]$$

Optimal Stopping Theory (OST)

- Choose the **best** time to **take** a decision of performing a certain action.
- **Observe** the current state of a system and decide whether to:
 - continue the process or
 - **stop** the process, and incur a certain cost.

...the *discounted sum* problem, the *odds* algorithm, the *secretary* problem, the *parking* problem, the *asset-selling* problem, etc.

Application to context collection problem

Decision

- □ *When* to **stop** collecting pieces of context from neighboring collectors/sources and deliver them to the application.
- > Cost
 - Quality of local pieces of context decreases with time.
 - □ *Serving* obsolete context to the application.

> Approach

□ *Adoption* of the OST **discounted sum** problem

Discounted sum problem in context collection

The decision of the collector at time *t* is:

stop and deliver local context to the application, or
 continue the process and update local context

Let us define a *tolerance threshold* $\theta \in (0, m^2)$ such that:

If $Y > \theta$ Then local context is significantly adequate for the collector's requirements in terms of **quantity** and **quality.**

Discounted sum problem in context collection

Consider the indicator function:

 $I_{t} = \begin{cases} 1, \text{ if } Y_{t} > \theta \\ 0, \text{ otherwise} \end{cases}$

and the *cumulative* sum up to time t: $S_t = \sum_{n=1}^t I_n$

The problem is to **decide** how large S_t should get before the collector stops, *i.e.*, we have to determine a time t such that the *supremum*

 $\sup_{t} \mathbb{E}[\beta^{t} S_{t}]$

is attained, $0 < \beta < 1$.

Discounted sum problem in context collection

Optimal Stopping Rule: Observe *I_t* value at time *t* and stop at the *first* time for which it holds true that:

$$S_{t} \geq \frac{\beta}{1-\beta} \left(1 - F_{Y}(\theta) \right)$$

 $\succ F_Y(y)$ is the cumulative distribution function of Y.

- $\triangleright \beta$ is *discount factor* indicating the **necessity** of collector to take a decision;
 - \triangleright collector requires a rather extended time horizon for deciding on deliver context when β is high.

Simulation setup

- > MSN of 100 nodes; number of sources $\omega = \{5, 10, 20\}$
- Mobility model: Random Waypoint
- ▶ Validity horizon $z \sim U(2,10)$ min.
- \succ Tolerance threshold θ ∈ [0.2, 0.7]
- > Maximum quantity of contextual parameters $m = \{10, 20\}$

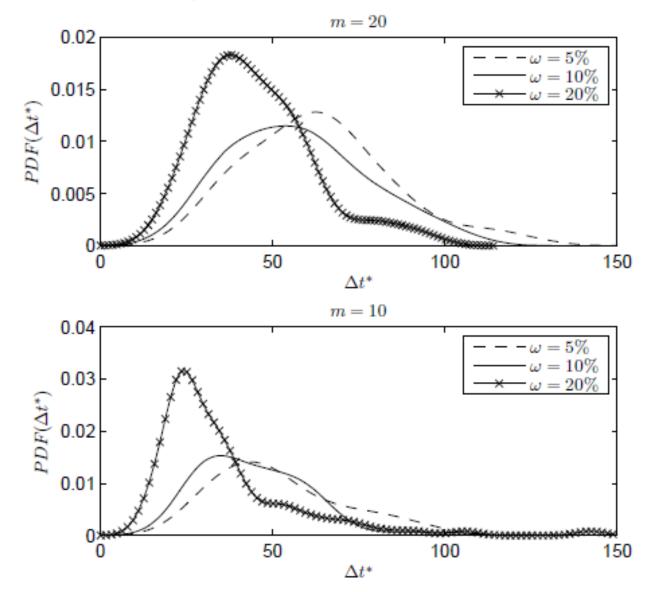
Comparison Schemes

- Scheme C: Randomized policy: collectors stop the process at a random time instance
- Scheme B: Finite-Horizon policy [1]: collectors stop the process based on a pre-defined deadline T; adoption of OST

> Metric

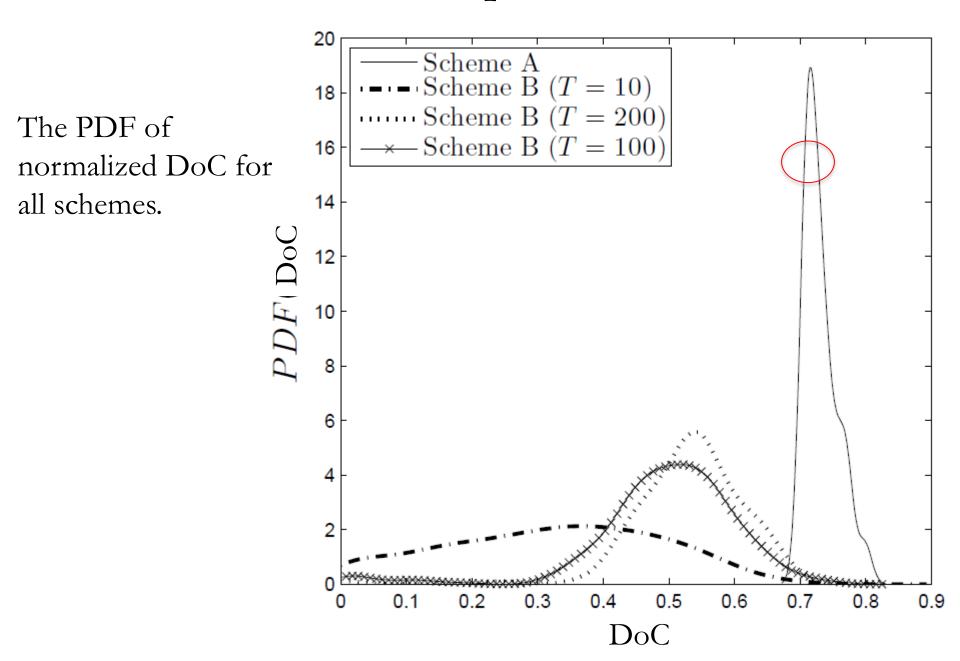
- ▶ Normalized average value of DoC delivered to the application;
 - the higher DoC is, the higher context quality and quantity is delivered to the application

The Probability Density Function (PDF) of the **decision delay** Δt^* , *i.e.*, interval between following collection processes, for diverse number of contextual parameters *m*.

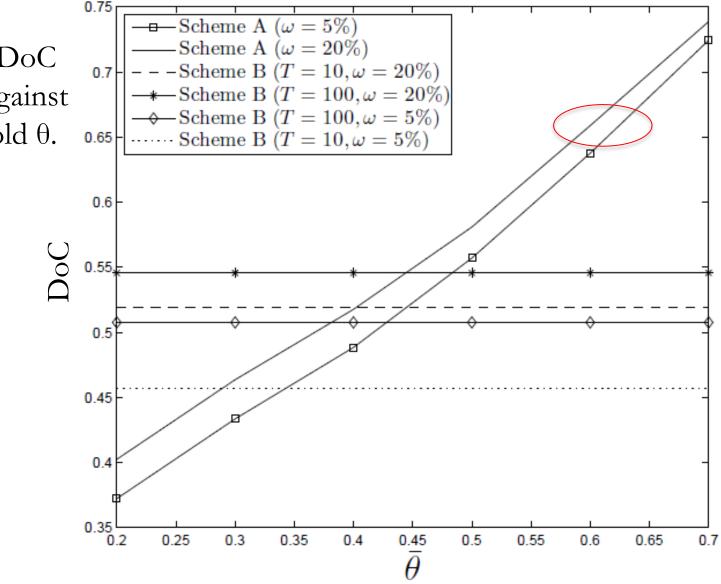


THE DOC FOR SCHEMES A, B, & C

Scheme	B			Α			C
m = 10							
$\omega(\%)$	T = 10	100	200	$\beta = .5$.9	.98	
5	.33	.52	.44	.64	.72	.71	.175
10	.33	.55	.57	.65	.74	.74	.175
20	.35	.55	.56	.66	.75	.74	(.175)
m = 20							
$\omega(\%)$	T = 10	100	200	$\beta = .5$.9	.98	
5	.30	.42	.45	.62	.63	.63	.17
10	.31	.48	.48	.64	.63	.62	.17
20	.32	.50	.50	.64	.64	.63	.17



The normalized DoC for all schemes against tolerance threshold θ .



Conclusions

- A solution to the *context collection problem* based on Optimal Stopping Theory;
- Collectors *autonomously* take time-optimized context delivery decisions **without** a deadline;
- Collectors deliver context of high quality and quantity within **short** delays;
- Our scheme performs **well** when dealing with applications which require context of high quality and quantity (*i.e.*, high tolerance threshold)

Thank you!