Deterministic Communication in the Weak Sensor Model

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OPODIS 2007



Capabilities

- processing
- sensing
- communication



University of California, Berkeley and Intel Berkeley Research Lab.

- range
- memory
- life cycle



PicoBeacon
Berkeley Wireless Research Center

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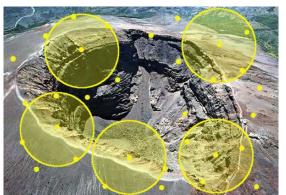
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- Hostile or remote environment
 - \Rightarrow deterministic deployment not feasible
 - \Rightarrow controlled random deployment.
 - \Rightarrow unknown topology, except for n and max degree k-1.
- Uniform Density: Random Geometric Graph, Unit Disk Graph, etc.
- Arbitrary Density: Geometric Graph $\mathcal{G}_{n,r,k}$.

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Topology Models

Node Deployment in Sensor Networks

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Node Constraints Models

Sensor Networks

THE WEAK SENSOR MODEL [BGI 92, FCFM 05]

- Constant memory size.
- Limited life cycle.
- SHORT TRANSMISSION RANGE.
- Low-info Channel Contention:
 - Radio TX on a shared Channel.
 - No collision detection.
 - Non-simultaneous RX and TX.

- Discrete TX Power range.
- Local synchronism.
- One channel of communication.
- NO POSITION INFORMATION.
- Unreliability.
- Adversarial wake-up schedule.

tx = transmission.rx = reception.



- Only one channel of communication
 - ⇒ must deal with collision of transmissions!

Popular solution \rightarrow random protocols.

- BUT scarcest resource is energy and
 - random protocols \Rightarrow redundant transmissions!
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 - Clear Reception at node x time slot t:

 Any one node in N(x) transmit and x does not transmit
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• Recurrent generalizations:

- Recurring Selection (single-hop RN, k out of n active)
 Every active node clearly transmits
 infinitely many times.
- Recurring Reception (multi-hop SN, max degree k-1) Every active node clearly receive from all neighbors infinitely many times.
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Related Work

Message passing:

[ABLP'92] Each node receives from all neighbors in $O(k^2 \log^2 n/\log(k \log n))$. \rightarrow synchronous start. $\omega(1)$ -degree bipartite-graphs requiring $\Omega(k \log k)$. \rightarrow not embeddable in GG.

• Broadcast & gossiping:

 $[\mathrm{CGR'00},\,\mathrm{CGOR'00},\,\mathrm{CR'03},\,\mathrm{CGGPR'02}] \to \mathrm{synchronous}$ start, global clock, etc.

• Selection

[Kowalski'05] Static, $\exists O(k \log(n/k)), +[I'02]: O(k \text{ polylog } n). \rightarrow$ synchronous start. Dynamic $O(k^2 \log n). \rightarrow$ nodes turn off upon succ. transmission.

• Selective families:

[I'02] $\exists (k,n)$ -selective families of size O(k polylog n). [DR'83] (m,k,n)-selectors must be $\Omega(\min\{n,k^2\log_k n\})$ when m=k. [DBGV'03] (k,k,n)-selectors must be $\geq (k-1)^2\log n/(4\log(k-1)+O(1))$ and $\exists (k,k,n)$ -selectors of size $O(k^2\ln(n/k))$. All \rightarrow synchronous start.

- Message complexity oblivious deterministic $\geq k$.
- $O(kn \log n)$ -delay message-complexity-optimal: Primed Selection.
- Recurring Selection-lower-bound
 - ⇒ Recurring Reception and Recurring Transmission lower bounds
 - $\Omega(k^2 \log n / \log k)$ -delay for Recurring Selection
 - (mapping (m, k, n)-selectors \leftrightarrow Recurring Selection)
 - \(\Omega(kn)\) for Recurring Selection with equiperiodic protocols.
 (memory limitations motivate)
- Choosing appropriately node periods, for $k \leq n^{1/6\log\log n}$, Primed Selection also delay-optimal for equiperiodic protocols .
- $O(k^2 \log k)$ -delay adaptive-protocol, using Primed Selection.
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Study deterministic oblivious (no history) and adaptive protocols for Recurring Selection, Recurring Reception and Recurring Transmission.

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1 Introduction

2 Oblivious Protocols

3 Adaptive Protocols

Oblivious Protocols

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Theorem

Any oblivious deterministic algorithm for Recurring Selection, single-hop RN, asynch. start, k active nodes \Rightarrow message complexity is > k.

PRIMED SELECTION:

For each node i with assigned prime number p(i), node i transmits with period p(i).

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Memory limitations \Rightarrow periodicity.

Definition

Equiperiodic protocol: set of schedules s.t. in each schedule, every two consecutive transmissions are separated by the same number of time slots.

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Using log log n-factors composite numbers instead of primes...

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Adaptive Protocols

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Primed Selection using O(k) coprime periods yields $O(k^2 \log k)$ delay. BUT, how do we guarantee every pair of neighbors use different period?

- Further assumptions:
 - Relax memory limitation to $O(k + \log n)$ bits
 - Double density to be able to half radii of transmission.
- Sketch of protocol:
 - Leave first k primes available.
 - Assign next k primes as before
 - Nodes use big primes to compete for small primes using Primed Selection with r/2.

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