

Distributed Clustering Method for Energy-Efficient Data Gathering in Sensor Networks

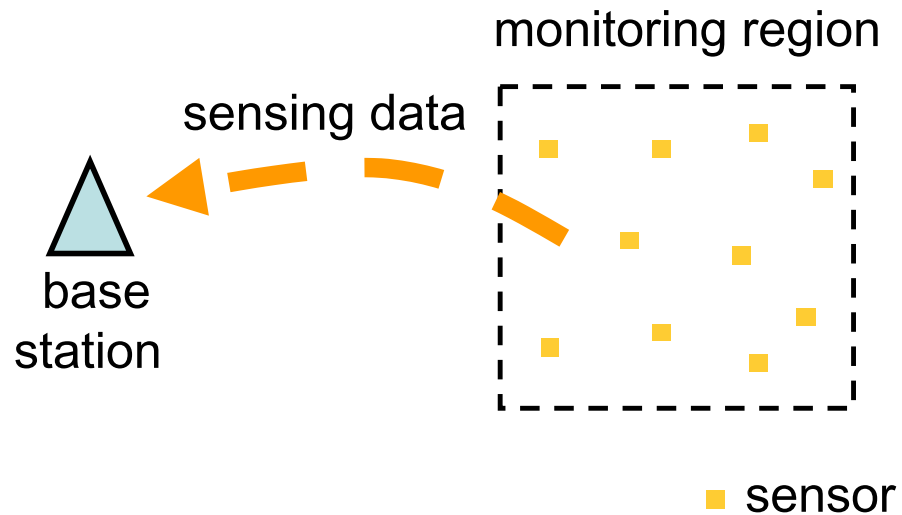
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1. Sensor Network

Network :

- Hundreds or thousands of sensor nodes are deployed to a region to monitor
- Sensor information is gathered to one base station



Sensors :

- Operate on a limited battery
- Difficult to recharge a battery
- Wireless communications
- Dynamically deployed, move, and halt
- Location aware (GPS, etc.)

Example Applications :

- Observation of severe environments (volcano, deep sea, etc.)
- Field surveillance for security
- Ecological research on wildlife

2. Objectives

Approach :

How to prolong lifetime of sensor networks ?

1. Limit communication range

Transmission cost is proportional to bit-size, n -th power of distance



Cluster-based data gathering

- Cluster-head gathers sensor data in its cluster
- Cluster-head sends the collected data to a base station

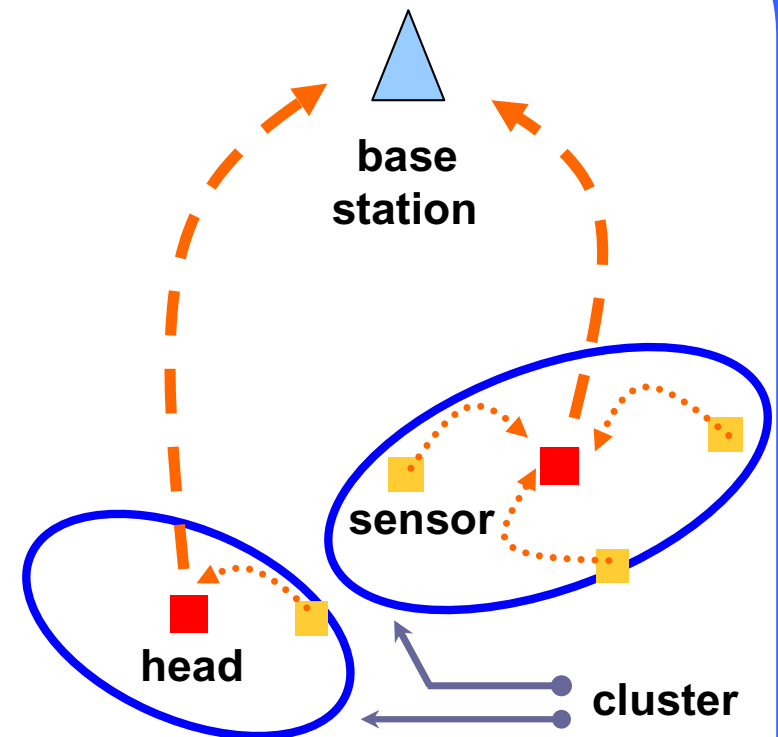
2. Equalize the residual energy of sensor nodes

Cluster-heads consume much energy



Rotation of a cluster-head

- Sensor nodes with large residual energy become cluster-heads



3. Cluster-based Data Gathering

Round :

1. Cluster-head candidacy

-- a part of sensor nodes broadcast a candidacy message



2. Cluster formation

-- Each sensor node chooses its cluster-head



3. Registration

-- Cluster members register itself to a cluster-head



4. Data gathering

-- Cluster members send sensor data to cluster-head
-- Cluster-head send aggregated data to a base station

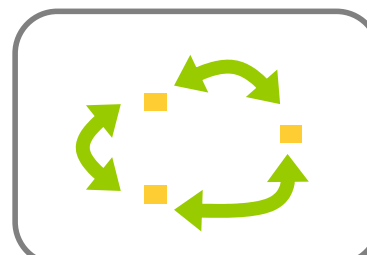
Clustering Method :

Demands :

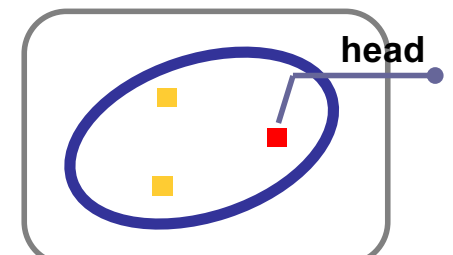
- Completely distributed
- Well-distributed clusters
- Energy-efficient
- Adapts to changes of sensor networks
 - Scalable
 - Robust

Our Approach :

- Clustering through local communications among neighboring sensor nodes



Local communication



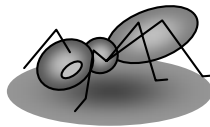
Cluster formation

4. Proposed Clustering Method

Reference :

- **Behavior of Ants (colonial closure)**

- Nestmate recognition by chemical substance
- Similar chemical substance in the same nest



- **ANTCLUST**

- Ant = Object
- Chemical = Information of cluster
- Nestmate = Cluster member
- Nest = Cluster

Randomly chosen two objects meet with each other
Clusters are created, merged, and discarded
through meetings

Proposed method :

Clustering method based on ANTCLUST
two phases of cluster formation

1. Cluster-head candidacy phase

- Sensor nodes with large residual energy announce their candidacy within range " R "
- Tentative clusters are formed



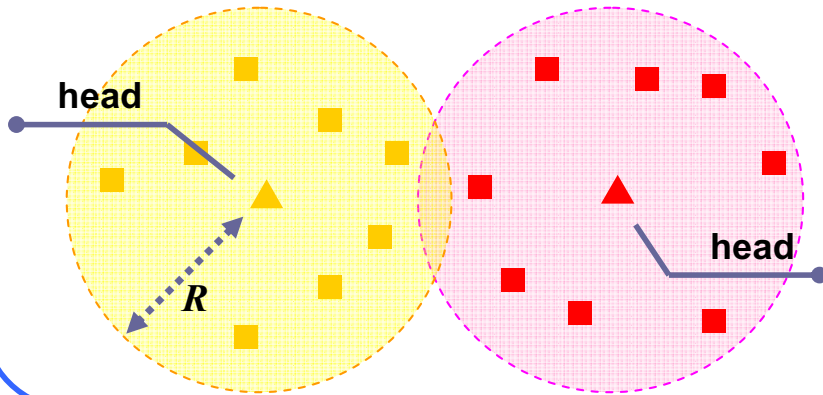
2. Cluster formation phase

- " P_{ex} " of sensor nodes broadcast their cluster information within range " r " to cause meetings
- Belong to a better cluster

5. Data Gathering with Proposal

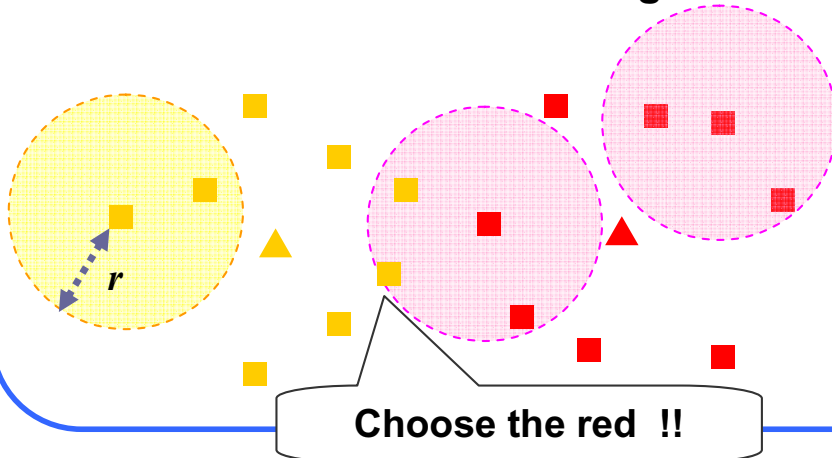
1. Cluster-head Candidacy Phase :

- Candidacy for a cluster-head



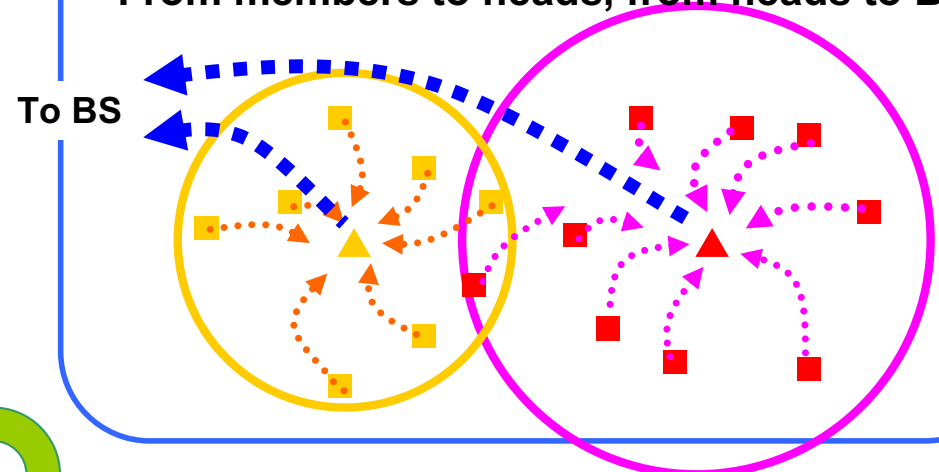
2. Cluster Formation Phase :

- Choose a better cluster through "Meetings"



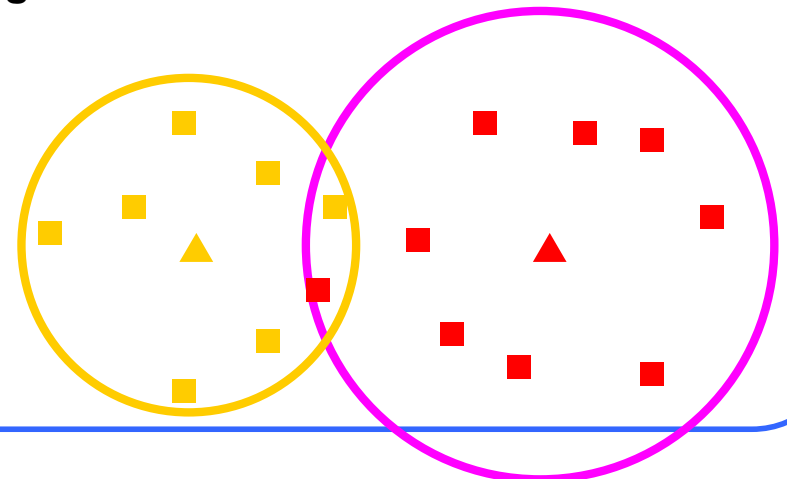
4. Data Gathering Phase :

- From members to heads, from heads to BS



3. Registration Phase :

- Register oneself as a cluster member



6. Proposed Method ~detail~

Parameters of sensor node i :

Parameters

- R : range of candidacy
 r : range of meetings
 P_{ex} : percentage of sensor nodes cause meetings

Information of sensor node

- i : identifier
 c_i : coordinates
 e_i : residual energy
 P_i : probability of candidacy
 $Template_i$: threshold variable

Information of a cluster

- $head_i$: identifier of cluster-head
 E_i : residual energy of cluster-head
 C_i : coordinates of cluster-head
 M_i : estimator of cluster member

A message contains $i, e_i, head_i, E_i, C_i,$ and M_i .

At cluster-head candidacy phase and cluster formation phase, when sensor node i received a message from sensor node j , sensor node i does:

-- Update probability of candidacy P_i

$$P_i = \begin{cases} P_i + p, & \text{if } e_i > e_j \\ P_i, & \text{if } e_i = e_j \\ P_i - p, & \text{if } e_i < e_j \end{cases}$$

-- Update threshold $Template_i$

$$Template_i = \frac{\overline{d(i, \cdot)} + \text{Max}(d(i, \cdot))}{2}$$

$d(i, \cdot)$ is the distance to a newly known cluster-head

7. Cluster-head Candidacy Phase ~detail~

Candidacy for cluster-head :

Sensor nodes with larger residual energy become a cluster-head

- Cluster-head candidacy phase has T time unit duration
- Sensor node i announce its candidacy at $T(1-P_i)$
- Cluster-head decreases P_i

Neighboring node i :

Compare distance d to a new cluster-head with threshold $Template_i$

- $d \leq Template_i$
 - *If* (sensor node i is not a cluster-head and belongs to no cluster)
 - ▶ belongs to the new cluster
 - *Else*
 - ▶ conducts the same procedure as in the next cluster formation phase
- $d > Template_i$
 - does nothing

8. Cluster Formation Phase ~detail~

Meetings :

P_{ex} of sensor nodes which are not a cluster-head broadcast cluster information

Neighboring node i :

• Compare distance d to a new cluster-head with threshold $Template_i$

If ($d \leq Template_i$) ► choose a better cluster

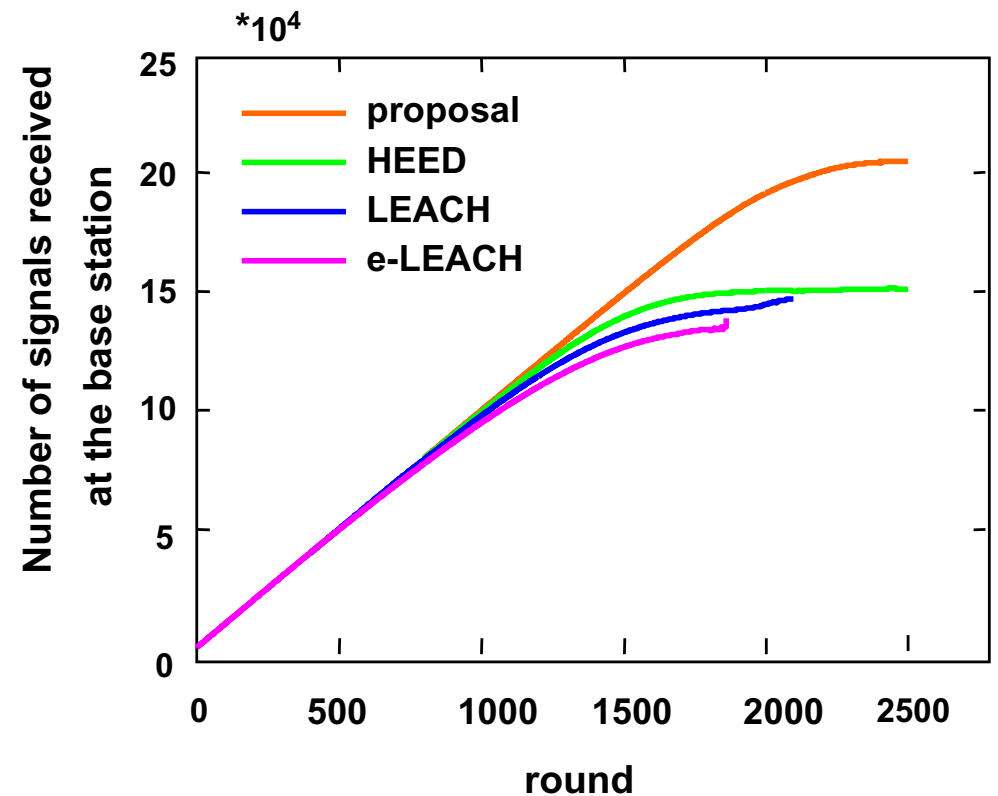
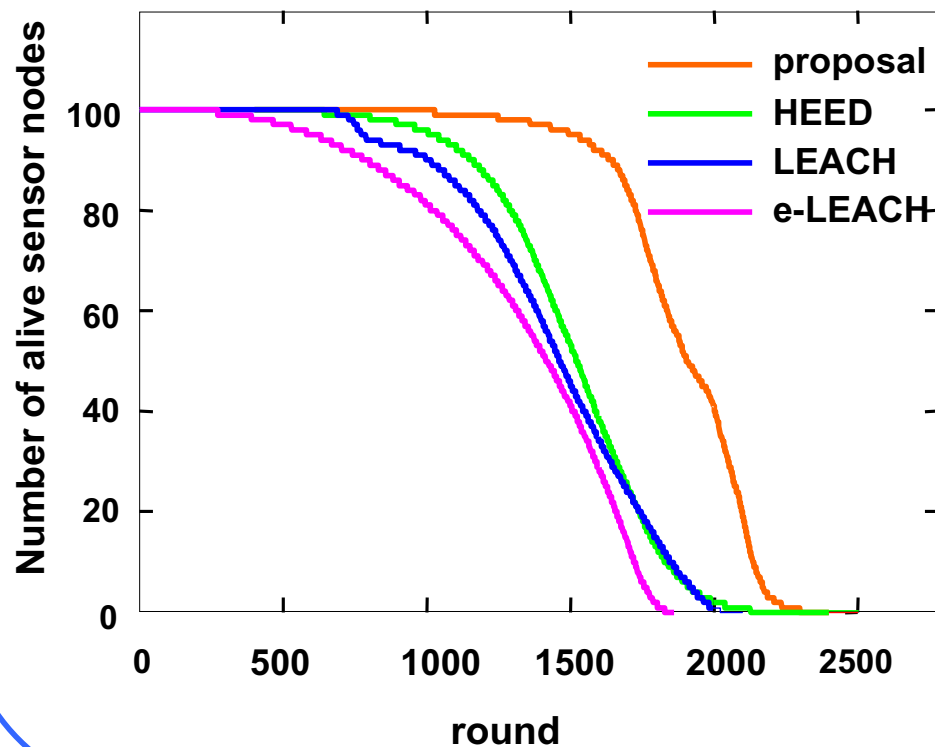
- closer cluster-head (smaller d)
- cluster-head with larger residual energy (larger E)
- cluster-head with less cluster members (smaller M)

Compare the value $\frac{E}{M \cdot d^2}$ of current cluster with newly known cluster

Then, belong to the cluster with large value

9. Simulation Experiments

- Comparison with HEED, LEACH, and e-LEACH
 - For each of methods, parameters were set to derive the best performance
- Setup: 100x100 area, 100 nodes, and initial energy 0.5J



10. Future Work

- Consider the case where data aggregation is impossible
- Introduce multi-hop transmission among cluster-heads
 - To adapt to large scale sensor networks

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