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A New Approach for Efficient Clustering Using **CECM** Algorithm in WSN

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Abstract: In this paper the proposed work is to compare the properties of the nodes in the cluster on the basis of probability of their being retrieved by other cluster or network. All this nodes and clusters contains battery level, memory, reverse signal strength indication, multicast routing information cost, time to live new node and time to live cluster on the basis of expectation clustering and maximization clustering in wireless sensor networks.

Keywords: Expectation, maximization, cluster head, energy efficiency, k-means, data aggregation, WSN.

I. INTRODUCTION

A wireless sensor network consists spaciously distributed of data, energy, memory, efficient techniques and data self-sufficient devices using sensors to monitor substantial or environmental conditions such as temperature, sound, vibration, pressure, motion or pollutants to collaboratively pass their data from the network to a destination or required to suffice the application quest. In addition, plunge where the data can be analysed and observed. A sensor node is a basic element in a sensor network, with on-board sensors, processor, memory, wireless modem and power supply. Sensor nodes contain reasonable battery power with few celestial spaces and processing power.

In a computer system, a cluster is a group of servers and other assets that act like a single system and enable high convenience and in some cases, load balancing and parallel processing. It is currently canvassing the Earth's magnetic environment and its interaction with the solar wind in three dimensions. Expectation and Maximization algorithm a habitual method to find maximum likelihood or maximum a posteriori (MAP) estimates of parameters in statistical models, where the model depends on unperceived latent variables.

K-means algorithm aims to disseverance n observations into k clusters in which each observation belongs to the cluster with the nearest contemptible, serving as an archetype of the cluster. This results in a partitioning of aggregation quandary, we discuss related work, and we the data space into glial cells. K-means clustering tends to suggest a number of algorithms. For respective of the find clusters of comparable ample extent, while the methods we provide unintelligible guarantees on the expectation-maximization mechanism allows clusters to endowment of unravel. have various shapes.

II. LITERATURE REVIEW

In data mining field nowadays there is a vast range of Step 1: cluster creation various fields like science research, statistical application, disaster area and War Zone. All this are related to mining

aggregation and data collection in WSN with calculation of cluster nodes, cluster head and other efficient techniques using data cube. A serviceable network is energy devastation of nodes is a great demur in order to maximize network endurance. Unlike other networks, it can be venturesome, very swank or even impossible to charge or replace kaput batteries due to the rancorous nature of encompassment.

The main focus of this gizmo is predominantly on duty cycling contrivance which represent the most reconcilable technique for energy saving and the data-driven passage that can be used to skyrocket the energy efficiency. We will make a review on some intercommunication protocols proposed for sensor networks. The work is how to poll the sensor nodes concretion to guarantee a full indemnity and to reduce energy devastation in a sensor network. During this paper there's a shot to allow a large analogize of the routing protocols in WSNs.

The Problem nascency does not require a prairie information about the number of clusters, and it gives a spontaneous way for superintendence missing values. We give a ceremonialistic statement of the clustering-

III. MECHANISM

(1) Phase 1: calculation of efficient expectation clustering

Cluster-Creation ()

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{	}
For (ss_id=0; ss_id <n; ss_id++)<="" td=""><td>,</td></n;>	,
{	(2)phase 2: calculation of efficient maximization
If $(Bcon \ge Apr \&\& Pm \ge Tm)$	clustering
// Bcon = Battery consumption	step 1: new node selection for maximization clustering
// Apr =Present Cluster battery level // Pm= Present memory	NewlyArrivingNodeInWSN ()
// Tm= Total cluster memory	
{	Calculate the following factor
Node will not be Included in Cluster	Bcon[i], Apr[i], Pm[i], Bandnew[i], Bandclst[i],
}	TTLnew[i],Tm[i], TTLclst[i], MRICnew[i],
else if (TTLnew>TTLclst)	RSSInew[i],MRICclst[i], RSSIclst[i];
// TTLnew= Time To Live new node	Calculate C (i1), C (i2), C (i3);
// TTLclst=Time To Live cluster	Calculate GCmin;
{ If ((MDIC nous) MDICalet) & & (DSSInous (DSSIalet) & &	If (newly arrive node GCmin< cluster head GCmin)
If ((MRICnew>MRICclst) && (RSSInew <rssiclst) &&="" (bandnew="">Bandclst))</rssiclst)>	t Assign new node as cluster head;
// MRICnew= Multicast Routing Information Cost	}
//MRICclst=Multicast Routing Information Cost of cluster	else
//RSSInew=Reverse Signal Strength Indication of new	{
node	Join cluster ();
//RSSIclst =Reverse Signal Strength Indication of new	}
node	}
//Bandnew= Level Bandwidth new node	
//Bandclst= Level Bandwidth cluster	Step 2: calculation of maximization clustering
{ No norm and one init Chastan ();	ClusterHeadAssignment ()
No new node can join Cluster ();	Total no of node = n;
else	For (cluster 1 to cluster n) $(1 - 1)$
{	
Join cluster;	For (i=0; i <n; i++)<="" td=""></n;>
Join cluster; }	{
Join cluster; } }	{ /*initialize parameter for cost evaluation*/
Join cluster; } }	{ /*initialize parameter for cost evaluation*/ Bcon[i] = { }; /* Battery consumption at node i*/
Join cluster; } } }	{ /*initialize parameter for cost evaluation*/ Bcon[i] = { }; /* Battery consumption at node i*/ Apr[i] = { }; /*Present Cluster battery level at node i*/
} } }	{ /*initialize parameter for cost evaluation*/ Bcon[i] = { }; /* Battery consumption at node i*/ Apr[i] = { }; /*Present Cluster battery level at node i*/ Pm[i] = { }; /*Present memory at node i*/
<pre>} } step 2: cluster head selection on the basis of expectation</pre>	{ /*initialize parameter for cost evaluation*/ Bcon[i] = { }; /* Battery consumption at node i*/ Apr[i] = { }; /*Present Cluster battery level at node i*/ Pm[i] = { }; /*Present memory at node i*/ Tm[i] = { }; /*Total cluster memory*/
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<pre>} } step 2: cluster head selection on the basis of expectation clustering for cluster1 to cluster k { for node1 to node n { sump=0;</pre>	<pre>{ /*initialize parameter for cost evaluation*/ Bcon[i] = {}; /* Battery consumption at node i*/ Apr[i] = {}; /* Present Cluster battery level at node i*/ Pm[i] = {}; /* Present memory at node i*/ Tm[i] = {}; /* Total cluster memory*/ TTLnew[i] = {}; /* Time to Live new node*/ TTLclst[i] = {}; /* Time to Live cluster*/ MRICnew[i] = {}; /* Multicast Routing Information Cost at node i*/ MRICclst[i] = {}; /* Multicast Routing Information Cost of cluster*/</pre>
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$\begin{cases} \\ \text{step 2: cluster head selection on the basis of expectation clustering for cluster1 to cluster k \\ \{ \\ \text{for node1 to node n} \\ \\ \{ \\ \text{sump=0; } \\ \text{llh=0; } \end{cases}$ $for (i=0; i<=n; i++) \\ \{ \\ \delta_{ij} = (D_i - C_j)^t R^- (D_i - C_j) \\ P_i = \frac{W_j}{(2\pi)^{p/2} R } e^{\left[\frac{-1}{2}\delta_{ij}\right]} \\ P_i = \frac{W_j}{(2\pi)^{p/2} R } e^{\left[\frac{-1}{2}\delta_{ij}\right]} \\ \\ \text{Sump = sump + } P_i; \\ \\ \\ \text{llh = llh + ln(sump); } \\ c' = c' + Y_i X_i; \end{cases}$	<pre>{ /*initialize parameter for cost evaluation*/ Bcon[i] = {}; /* Battery consumption at node i*/ Apr[i] = {}; /* Present Cluster battery level at node i*/ Pm[i] = {}; /* Present memory at node i*/ Tm[i] = {}; /* Total cluster memory*/ TTLnew[i] = {}; /* Time to Live new node*/ TTLclst[i] = {}; /* Time to Live cluster*/ MRICnew[i] = {}; /* Multicast Routing Information Cost at node i*/ MRICclst[i] = {}; /* Reverse Signal Strength Indication of new node*/ RSSIclst[i] = {}; /* Level Bandwidth new node*/ Bandnew[i] = {}; /* Level Bandwidth cluster*/ /* now calculate the cost for each node */ C (i1) = (Apr[i]* TTLnew[i] /Bcon[i]* TTLclst[i])) C (i2) = (Pm[i]* RSSInew[i] /Tm[i]* RSSIclst[i] } </pre>

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min = 0; max = 0;

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If (max< GCmin)

else if (min> GCmin)

max = GCmin;

 $\min = GCmin;$

(i1) = $(Apr[i]^*)$

GCmin = C(i1) + C(i2) + C(i3);

C(i2) = (Pm[i] * RSSInew[i] / Tm[i] * RSSIclst[i])

(MRICnew[i]*Bandnew[i]/MRICclst[i]*Bandclst[i]}

(i3)

for (i=1 to n)

TTLclst[i]))

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BIOGRAPHIES



Gohil Sonal is pursuing BE in Computer Engineering from IIST, rajpur, kadi, Gujarat, India. She is currently doing her 8th semester project in .Net language. Her interest of area is networking in WSN.



Sumit Chaudhary is working as Head of Department in CSE at Indrashil Institute of Science & Technology, Cadila Group, Rajpur, Ahmedabad (Gujarat). He is pursuing Ph.D. from Uttaranchal University, Dehradun (Uttarakhand). He worked with various institutes like

Uttaranchal Institute of Technology (UIT), Dehradun, Shri Ram Group of colleges, Muzaffarnagar (U.P.), IIMT Institute of Engineering & Technology, Meerut (U.P.), INDIA including all that he has more than 7 year experience in teaching. He obtained his M-Tech (Computer Science & Engineering) with Hons from Shobhit University and B-Tech (Computer Science & Engineering) from SCRIET, Meerut (U.P.). During this short period of time, he has been supervised several dissertation of M.Tech students. He has been member of several academic and administrative bodies. During his teaching he has been coordinated many Technical fests and National Conferences at Institute and University Level. He has attended several seminars, workshops and conferences at various levels. His many papers are published in various national, international journals and conferences. His area of research includes Cloud Computing, Wireless Sensor Network (WSN), Network Security, Neural Network, Artificial Intelligence and MANET (Mobile Ad-Hoc network) ...



Dr. S.B. Batra is engaged as Dean (academic) with IIST, Rajpur, Mehsana, Gujarat, India. Heholds long experience of working with Indian AirForce Aeronautical Engineer (Electronics).After obtaining Pre-mature retirement in 1999, he is engaged in the

field of academic. He has served number of professional universities/engineering collages in India and abroad as director/principal/dean/professor. His many papers are published in various national/international journals.

IV. CONCLUSION

This paper applause for better and efficient techniques for cluster head selection and calculation of nodes of data cube on the basis of weight, memory and like that other things using clustering expectation and clustering maximization algorithm. It provides efficient techniques for energy saving in clustering with CECM algorithm in WSN with data cube aggregation.

FUTURE SCOPE

The proposed work provides efficient methods for energy in data aggregation but the field of data mining has various heterogeneous behavior in cluster node with ongoing advancement.

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