Enhanced Event Coverage and Redundant Data Minimization Employing Pentagonal Scalar Premier Selection in Wireless Multimedia Sensor Networks (WMSNs)

Sushree Bibhuprada B. Priyadarshini

Abstract— In the modern era, data redundancy has become one among the predominant ultimatums encountered in Wireless Multimedia Sensor Networks (WMSNs), which occurs because of event information reporting through the scalars residing at the superimposing zones of field of views (FoVs) of multiple camera sensors. As a result, same data is transferred many times, thus leading to redundancy in data transfer. Therefore, the aim is to select the representatives of scalar sensors called scalar premiers (SPs) that can report the event information in lieu of all the scalars while diminishing the redundant data transfer and improving the event coverage. We have proffered a pentagonal scheme of SP selection that chooses five SPs in each of the virtual compartments of the monitored zone efficiently. The chosen SPs operate as nominee of scalars for event information transmittal. Extensive experiments have been accomplished to affirm the efficiency of our proffered method. We changed the number of cameras deployed (noc) and the number of scalars deployed (nos). The results attained from the experimental studies in terms of number of camera sensors activated (nca), coverage ratio (cr), redundancy ratio (rr), event loss ratio (elr) and energy expenditure for camera actuation (eca) assert the superiority of our proffered approach over existing approaches.

Index Terms: Event Grade, Field of View, MCIM, MSIM, My Waiting List, Scalar Premier

1. INTRODUCTION

In this modern age of advanced technological procreation, the adoration of sensors is fundamentally owing to their dissonant perspective of relevance. However, such widespread applications give rise to diversified challenges. The most protrusive challenge in such networks is to attain an improved coverage of the area under consideration, while conjointly lowering the redundancy in data transmission which prevails owing to the superimposing of Field of Views (FoVs) of cameras. Consequently, the scalars residing at the overlapping zones convey the common information pertaining to the ongoing event to their concerned cameras as portrayed in Fig. 1.

The pink circle represents the event region. The tiny squares denoted by C1, C2, C3, etc. denote the camera sensors and the tiny filled circles denote the scalar sensors. Whenever, event takes place, it get ensnared through scalars and those scalars send the event information to the concerned camera sensors. Here, the triangular shapes represent the FoVs of the cameras. the scalars residing at superimposed areas of the FoVs report the same data number of times. As same data is reported several times, this results in redundant data transmittal, thereby leading to undesired energy and power expenses. Thus, the total repeated data transfer must be diminished so that only lowered number of camera sensors get turned on. The prime objective is that the activation of camera sensors has to be accomplished so that the amount of event region coverage get escalated conjointly with the depreciation of the repeated data transfer.

Fig. 1. A Scenario of Redundant Data Transmission

In this context, while thinking about the case of sensing environment efficient event tracking is both a popular and relevant aspect of investigation while talking about the case of a monitored area of examination as devised by Bhoi, S., etal.. in 2012 [1]. But, in case of failure of sensors it turns to be very hectic to track the prevailing event efficiently. Moreover, the major objective concerned in the method is to design a cluster through deploying density based clustering paradigm. Further, Bhoi, S., etal. (2012) [2] demonstrated another method while taking into account the case of wireless environment which employs a fault tolerance and conjointly achieves the required region coverage. Hence, for effective event tracking, sensors play a very significant role in ensuring reliability while monitoring the event zone under speculation.

In this paper, a new distributed scheme regarded as “Pentagonal Scalar Premier Neighborhood for Optimum Camera Actuation (PSPN-OA)” is demonstrated that elects scalar premiers among the scalars uniformly along all the directions of the virtual compartment. In this scheme, the

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elected Scalar Premiers (SPs) act as the delineative of the scalar sensors that report about the prevailing event to the concerned camera sensors. The main aim of our proffered approach PSPN-OA can be advocated as follows:

- The PSPN-OA approach divides the whole monitored area into number of smaller compartments known as Virtual Compartments (VCs). The proposed strategy engrosses the election of five SPs in each of the VCs individually.
- The SP election is carried out such that the camera sensors which are actuated owing to the elected SPs can camouflage greater amount of different portions of the prevailing event region uniformly, while conjointly diminishing the camera activation in concerned monitored zone.
- The activation of reduced number of camera sensors in case of our proffered PSPN-OA method leads to reduced amount of redundant data transfer with lower amount of superimposing among the FOVs of actuated cameras..
- In addition, the amount of energy exhaustion for camera to be turned on is diminished in the proffered scheme owing to activation of minimal count of cameras.

The subsequent section elaborates the related work carried out in our field of study. Later on, we discuss the proposed scheme along with the whole methodology associated with the proposal. Subsequently, the results attained from the experimental investigation has been elaborately discussed. Finally, we conclude the paper with direction towards the future mission.

II. RELATED WORK

Various work have been carried out in past for lowering of the amount of repeated data communicated. In this connection, Distributed Collaborative Camera Actuation based on Scalar Count (DCA-SC) [3] represents one popular scheme for lowering the repeated data communicated. Here, the cameras collectively make decision regarding their activation. Likewise, the approach discussed in [4] devised a Non-Heuristic (N-H) scheme aimed at lowering the Energy as well as Power expenditure prevailing in case of WMSNs. This strategy is focused on making the camera sensors activated unnecessarily in off condition and at the same time diminishing the amount of redundant data transmittal owing to such cameras.

Moreover, Distributed Collaborative Camera Actuation Scheme based on Sensing-Region Management (DCCA-SM) [5] represents a method which effectively segregates the whole region into a number of sensing zones. Afterwards, in every region, the cluster heads get determined those report the cameras about the happening of an event. Besides, the notion of directional coverage method [6] focuses on independent targets attached with differentiated importances. Moreover, the article addresses the “priority-based target coverage” scheme. It attempts to select the minimal subset of directional sensors which ensnare all targets, thereby pacifying the expected priorities.

Similarly, S. Sundhar Ram (2007) [7] advocated a novel strategy of path coverage and the examination of the entire coverage phenomenon cajoled on 1-dimensional path through sensor network get modeled as a 2-dimensional Boolean model. Moreover, the strategy demonstrated in [8] that puts focus on directional cover sets issue of arranging the sensor directions into a gathering of various cover sets in the objective of prolonging the life-time. Paper [9] has given a strategy that advances in two passes for the goal of repeated data minimization. A local elimination has been accomplished which expels the repeated messages locally in every state of the automaton. Further, the approach proffered in [10] depicts a methodology where a method is proposed to estimate the similarity between the data gathered to the base station.

Apart from that, the approach [11] elaborates for sensor placement. Further, the proffered formulation assists in framing of larger Wireless Sensor Networks (WSN), which are established in effective way so that this preserves proper balance between lowering congestion while the data packets follow the shorter routes. Besides, a redundant positioning framework discussed in [12] depicts a new design for processing huge deal of data from various pervasive appliances. Moreover, a new fangled method has been given in [13] where provides desired k coverage, where every point in a field get camouflaged through minimum k number of sensors. The proffered scheme constructs maximal number of layers, in which every layer is k-covered.

Besides, an indepth study has been carried out in [14] from perspective deformation domain. First of all, the deformation of the point is segregated into 3 parts. Again, noise immunity of every portion associated with several FOVs is elaborated for justifying the prime reason behind the hettiness in the surveying of constricted cameras. Moreover, method given, in 2008, depicted a problem with the objective of prolonging the lifetime of network through scheduling [15] the sensors into various sets so that every sensor preserve target coverage as well as connectivity among all the active sensors and the destined sink. Further, the in [16] uses a cooperative multi-camera target ensnaring approach for those sensor networks, in which the target ensnaring [16] has been done through the sensors coordination based on the automatic node determination.

A view coverage method given in [17] estimates the coverage quality [17] along with a finer granularity for the objective of recognizing face. Similarly, the work carried out in [18] deploys a method which uses differential evolution optimizer as the search strategy. Likewise, the work in [19] used a dynamic itinerary for the certainty map. Furthermore, the approach in [20] has stated an analysis of various techniques concerned with coverage predicaments. Similarly, the method in [21] demonstrated the coverage goal as a measure of the quality of service that is afforded through a specific network construction. In like manner, a novel grid coverage methodology for efficient tracking and target position in distributed sensor networks is described in [22]. The sensor placement issue for a planar grid zone get formulated as the combinatorial optimization predicament as considered in [23]. Further, the method is concerned with the
goal of prolonging the total detection probability.

A novel approach [24] that works efficiently for considering multiple events which occur concurrently in the geographic monitored zone, at the same time lowering the count of undesired activated cameras sensors thereby curtailing the amount of redundant data reporting. Similarly, a maximal far-flung methodology [25] uses a distributed method that involves maximum far-distant scalar premier election for camera to be turned on. This way selection reduces the possible superimposing among the FoVs, hence lowering the amount of repeated data transmittal prevailing due to it. Likewise, in another investigative work in [26] employs a distributed technique based on election of scalar leaders for the purpose of improved coverage of the occurring event zone. This manner of scalar leader determination is done such that the leaders get systematically arranged in a hierarchical manner where every child node is located minimum at a length of double of Depth of Fields (DoFs) chosen and the elected leaders act as the entity for event information transmittal to the sensors [26].

Despite all the aforesaid methods try to lower the amount of repeated data transfer, redundancy still persists in case of data transmittal. Hence, the goal is to activate merely the desired count of camera sensors so that the amount of repeated data transfer is reduced, while conjointly attaining enriched coverage of the concerned area.

III. PROPOSED PSPN-OA METHOD

This paper proffers a novel method known as “Pentagonal Scalar Premier Neighborhood for Optimum Camera Actuation (PSPN-OA)” which activates reduced number of cameras while conjointly affording enhanced event coverage while diminishing the redundant data transmittal.

Following are the steps encountered in our proffered PSPN-OA method:

Step 1. Random Placement of Sensors

At the beginning all the scalars and cameras are arbitrarily deployed in the monitored zone. Initially, the ids of all the camera sensors get retained in a list namely, Current Waiting List (CWL). A list namely Event Grade Id List (EGIL) is preserved that maintains the ids of the camera sensors that finally undergo activation. Cameras and scalars broadcast My Camera Information Message (MCIM) and My Scalar Information Message (MSIM) successively that retain the id as well as position information of the respective sensors.

Just after the receipt of the two messages all the sensors become aware regarding the position and ids of all the sensors. Now each of the camera sensors finds its distance from scalar sensors individually. If the concerned Euclidian distance is lower than the DoFs of the corresponding camera; then it is considered that the particular scalar sensor comes under the purview of the concerned camera and hence the id of the particular scalar now get added to the id list of a table called Recent Field of View (RFOV) table.

Step 2. Selection of Scalar Premiers (SPs)

The monitored zone get divided into number of Virtual Compartments (VCs) for effective determination of Sps. The division is carried out following the same strategy as used in paper [25] such that the total monitored region of size 500*500 get divided into 100 VCs. Further, within each of the virtual compartments (VCs), five SPs are determined among the scalars present within that particular VC as portrayed in Fig. 2.

Such thing is accomplished so that the initial SP get selected towards the right side of the X-axis that passes through the centre of VC. Such premier is selected such that it is the maximum distant scalar that resides within the particular virtual compartment at farthest distance from the centre of the particular VC but within the concerned VC. If no such scalar sensor is available owing to random deployment of sensors; then the scalar sensor having sensing range spanning such virtual X-axis is determined as the initial premier. If no such scalar still covers the virtual axis, at that time the farthest scalar sensor that lies at maximal distance from the centre of VC along the counter clockwise direction from concerned virtual axis get selected as the initial scalar premier.

![Fig. 2. A Scenario of Scalar Premier (SP) Determination](image-url)
cameras getting actuated owing them can camouflage the event region more effectively. In this way the SPs get selected in each of the VCs of the region to be tracked.

C. Step 3. Event Reporting and Camera Sensor Activation

During the time of event occurrence, the elected SPs which come within the prevailing event area ensure the event and inform about the ongoing event to the corresponding cameras within whose RFOV table they lie by sending a DETECTION message that retains the concerned SP’s id and the prevailing event information. Being reported from the SP, the camera sensor now updates its Event Reporting Scalar Premier List (ER-SPL) that retains the ids of merely those SPs which ensure the occurring event. In this context, the determination of event point and event radius by the scalar get accomplished by the following way:

(i) When any event occurs, the scalars that detect the event acquire signal having definite intensity. Again, the signal strength diminishes with rise in distance from the point of happening of the prevailing event. At the instant, a scalar attains a signal, it sends DETECTION message which retains the intensity of the signal which this got. Now, in like manner all scalar sensors can know the intensities of signals received by the other sensors. The scalar sensor can recognize the direction along which the signal intensity increases through exchange of information with its neighbors.

Following are the notations used:

**Notations Used:**

- $I_{sp}$: intensity of signal (IOS) at the point of occurrence of event
- $ER$: radius of occurring event
- $I_{O}$: IOS at the point other than the occurring event point, at which any scalar $s$ is residing
- $d_{i}$: decelerating rate of the signal intensity
- $p$, $q$, $p$ is the scalar sensor handy to the occurring event than scalar sensor $q$
- $U$, $V$: IOSs acquired by $p$ and $q$ respectively
- $\Delta t$: time appropriated by the signal to move from $p$ to $q$

which is indicated as follows:

Mathematically

\[
\Delta t = t_2 - t_1
\]  

(1)

$t_1$ and $t_2$ are representing the time when the signals are acquired by $p$ and $q$ successively.

As $p$ is closer to the prevailing event than $q$, thus, the IOS ($V$) received by $q$ is lower than IOSs ($U$) acknowledged by $p$ and can be interpreted as shown below:

\[
V = U - d_1 \times \Delta t
\]  

(2)

(since the decelerating rate of the signal intensity $d_i$ is negative)

Now, $d_1$ can be estimated by the following equation

\[
d_1 = (U - V)/\Delta t
\]  

(3)

Thereafter, the scalar computes the position of the other scalar (let it be point $N$ having coordinate $(X_n, Y_n)$), at which IOS is the maximal from the received DETECTION message. This intensity is the value of $I_{sp}$.

Now, $I_{O} = I_{sp}^2 - 2d_i S_1$

(4)

where $I_{O} \leq I_{sp}$

$S_1 = (I_{sp}^2 - I_{O}^2)/2d_i$

(5)

Where, $S_1$ is the distance of event point from any sensor $s$. (ii) Now the position of minimum signal strength accepting scalar get attained from by the DETECTION message attained through the various scalars. Let this point be $P$ having coordinate position ($X_p$, $Y_p$). During the period of sprinkling of sensors, a threshold value for the IOS attained by the scalar sensor is set for determining whether the intensity of the signal received is significant for deciding that the event information to be tracked is needed or not.

(iii) Through DETECTION message the location of the scalar sensor which received the IOS value equal or just higher than the considered threshold value is selected. Let such point be $B$ having coordinate ($X_b$, $Y_b$). The Euclidian distance from point $N$ to $B$ is the value of the event radius $ER$.

(iv) After determination of event point and event radius, all the camera sensors now calculate the sum of event informing SPs that lie within their DOFs known as Event Reporting Scalar Premier Estimate (ER-SPE). The estimation of ER-SPE value is done as follows:

- Assign sum to zero (0)
- For each camera sensor $a$ and any scalar premier $b$, estimate the Euclidian distance (Edist$_{ab}$) between them.
- If Edist$_{ab} < DOF$, then sum = sum +1;
- For each camera sensor $a$, execute this calculation for every scalar premier $b$.
- Now update the sum value every time, if their corresponding Edist$_{ab} < DOF$ of corresponding camera sensor.
- Further, the updated sum value represents ER-SPE value of camera $a$.

- After the estimation of the ER-SPE value is over, the camera sensors append their ids to either Multi Premier Camera Ballot (MP-CB) or Single Premier Camera Ballot (SP-CB) based on the ER-SPE value. Further, MP-CB and SP-CB retain the ids of camera sensors in increasing sequence possessing $ER-SPE \geq 2$ and $ER-SPE = 1$ successively. The ids for the cameras are kept in MP-CB and SP-CB in the same sequence as that of CWL.

- Now, the camera sensor whose id appears first in MP-CB gets actuated first and its id is included in AL and gets expunged from CWL.

- When a camera undergoes actuation, it transmits a Update Scalar Premier (USP) message to the remaining camera sensors. USP retains the ids of SPs residing within the DoFs of camera sensors, that are estimated based on the estimated Euclidian distance between the actuated camera sensor and the concerned SP.

- As soon as USP message is received, the Update Message id List (UM-IL) get updated which retains the ids of SPs contained in the corresponding USP message. Consequently, the subsequent camera coming immediately in the MP-CB (say, $b$) matches the ids of SPs contained in its ERSP with the ids of SPs maintained in USP message, transmitted by the actuated camera.

- Now the following scenario prevails:
if (scalar premier ids retained by USP message and that maintained in ERSPL list match totally)  
Then {No need to actuate the camera b}  
else  
Then {Camera b has to be activated}  

- Likewise, the remaining cameras maintained successively in MP-CB list decide to go to turned on condition based on comparing their corresponding SP ids contained in their ERSPL list with the SP’ ids contained in USP message based on the previous two situations.  
- Whenever, a camera get actuated its id is included in CAL and gets removed from the CWL. At the time when the corresponding cameras residing in MP-CB list get actuated, then the ids of scalar sensors lying in USP messages of already actuated cameras, contained in UM-IL for MP-CB list are compared with the SP ids contained in ERSPL list of cameras lying in SP-CB list. Further, in case the scalar ids contained in ERSPL of camera (s) lying in SP-CB list completely matches with SP ids retained in updated UM-IL, at that time no necessity to actuate the corresponding camera sensor.  
- However, if a conflict is noticed, that camera pertaining to SP-CB gets actuated. The actuated camera sensor of SP-CB list broadcasts USP message.  
- Such process persists till all the cameras decide either to be actuated or to be preserved in off condition. Thereafter, EGIL is updated. The total count of camera sensors lying in EGIL denotes the Event Grade.  

IV. RESULTS AND DISCUSSION  

We have conducted the performance analysis of our proffered PSPN-OA method by taking results of three other approaches currently emerged recently [3, 4, 5]. They include: DCA-SC [3], N-H[4], DCCA-SM[5] methods and proffered PSPN-OA.  

Simulation Setting  

Our paper has been implemented using C++. During the experimentation, all the scalars sensors and camera sensors are considered to be arbitrarily sprinkled where as the camera sensors are considered to have predefined fixed locations. Omni-directional camera sensors are used in case of our proffered strategy as they can trap panoramic image of objects uniformly. The number of cameras (noc) and number of scalars (nos) are varied independently and marked the impact on performance metrics as described below:  
1. number of cameras actuated (nca)  
2. energy expenditure for camera actuation (eeeca)  
3. coverage ratio (cr): The cr is depicted as “the portion of the area of events that is covered by all the activated cameras with respect to the whole regions of the occurring events”[3].  
4. event loss ratio (elr): The elr is depicted as “the portion of area of events that are not covered with respect to whole region of occurring events”.  
5. redundancy ratio(rr): rr can be defined as “the ratio of total portion of superimposed portions of FOVs of actuated cameras to the total unique portions of event region that covered by the concerned cameras”.  

Results Discussion  

- Impact of variation noc  
- The noc was varied and its impact was marked on nca, rr, cr, elr and eeca respectively in DCA-SC, N-H, DCCA-SM schemes as well as proffered PSPN-OA method as displayed in Table 1, Table 2, Table 3 and Table 4 successively. From all the tables it is crystal clear that with a rise in noc, the nca gradually rises in case of all the approaches. Further, the value of nca is obtained to be the lowest in proffered PSPN-OA method.  
- Likewise, with a hike in ncd, cr hikes in the other schemes. It is owing to the cause that with rise in noc, greater count of camera sensors come under the ambit of prevailing event and hence greater number of cameras undergo actuation. Thus, greater portions of event area gets covered that gives rise to increasing value of coverage ratio.  
- Moreover, the cr value is attained to be the maximum in proffered PSPN-OA scheme and event area coverage is attained to be maximal of 86% in the proffered strategy.  
- Similarly, with a hike in noc, the rr goes on rising in case of all the methods as with rise in noc, the amount of superimposing zone among the Fovs hiker in every scheme. Further, the redundancy in percentage obtained in the proffered method is the minimal at 39% while varying the ncd. Moreover, the eeca is found to be the minimal in proffered PSPN-OA scheme than the other existing schemes as least count of camera sensors undergo actuation in case of the proffered scheme.

Impact of variation of nos  
- The number of scalars has been varied and its effect has been marked on several performance metrics in case all the approaches as presented in Table 5, 6, 7 and 8 successively.  
- The tables clearly depict that with a hike in nsd, the nca rises in all the approaches because with gradual rise in nsd, the count of event ensnaring scalars also goes on rising. Moreover, the nca is obtained to be the minimal in our proffered scheme.  
- Similarly, with a hike in nsd, the value of cr hikes in all the methods and is attained to be the maximum in the proffered method. The initial rise in cr is due to the fact that with rise in nsd greater number of scalar premiers are going to ensnare the prevailing event. Further, maximal coverage of 92% and minimal event loss of 8% is attained in case of our PSPN-OA scheme.  
- However, the rr found in our proffered scheme is the minimal that ensures the effectiveness of our proffered PSPN-OA scheme.  
- With hike in number of scalar the eeca also increases. However, it is observed that eeca is obtained as the least in our proffered PSPN-OA method in comparison of others.
### Table 1. Impact of Varying $noc$ in DCA-SC

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### Table 2. Impact of Varying $noc$ in N-H

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### Table 3. Impact of Varying $noc$ in DCCA-SM

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### Table 4. Impact of Varying $noc$ in proposed PSPN-OA

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### Table 5. Impact of Varying $nos$ in DCA-SC

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<td>0.75</td>
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### Table 6. Impact of Varying $nos$ in N-H

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<th>rr</th>
<th>elr</th>
<th>eeca (joule)</th>
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### Table 7. Impact of Varying $nos$ in DCCA-SM

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<th>rr</th>
<th>elr</th>
<th>eeca (joule)</th>
</tr>
</thead>
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Table 8. Impact of Varying nos in proposed PSPN-OA

<table>
<thead>
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<th>rr</th>
<th>elr</th>
<th>eeca (joule)</th>
</tr>
</thead>
<tbody>
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<tr>
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V. CONCLUSIONS AND FUTURE SCOPE

This article depicts a newfangled method known as PSPN-OA that actuates minimal count of cameras at the same time conjointly diminishing the amount of repeated data reporting. Such selection of scalar premiers avoids multi-event informing by all the scalar sensors. Moreover, the uniform selection of scalar premier give rise to more even coverage of region tracked. Further, extensive experiments have been conducted to assess the effectiveness of our proffered scheme while assessing it with other existing approaches in the literature. The noc and nos were varied independently and the aftereffect were marked successively in case of all the methods. The numerical results make it obvious that the nca and eeca are attained as the minimal in our proffered PSPN-OA method than the other methods. Likewise, the cr is found to be the maximal while the rr is the minimal in case of our proffered PSPN-OA method. While varying the noc, it is clear that the event coverage is attained to be the maximum at 86% in proffered method. Likewise, while varying the noc the minimized value of rr and elr are obtained as 39% and 14% respectively in our proffered method. Similarly, during the variation of nos, the event coverage is found to be the maximal in case of our approach at 73% where as the redundancy and event-loss are marked to be minimal at 42% and 34% respectively. Although our proposed system reduces the amount of redundant data transmission, still it has some limitations. We have used camera sensors those are having fixed positions by keeping the real life random deployment scenario of sensors in mind. However, as a direction towards our future work, mobility can be added to the cameras for increasing the adaptability of entire system. Further, we are planning to map our proposed system to three dimensional scenarios as an insight towards the future investigation.

VI. ACKNOWLEDGEMENT

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She has been invited and selected as the Keynote Speaker of International Conference of Innovative Applied Energy, St. Cross College, Oxford University. Besides, She is the Member of IAENG, SDIWC, CSTA. She is also working as reviewer in International Journal of Engineering Research & Technology and IEEE Consumer Electronics journal. She has around 26 publications in various international journals and Conferences including IET, IEEE, Elsevier, Springer, IGI Global etc.

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