3D Real-Time Reconstruction Approach for Multimedia Sensor Networks

Ahmed Mostefaoui, University of Franche-Comte (LIFC), France
Benoit Piranda, LASELDI, France

ABSTRACT

Multimedia sensor networks have emerged due to the tremendous technological advances in multimedia hardware miniaturization and the application potential they present. However, the time sensitive nature of multimedia data makes them very problematic to handle, especially within constrained environments. In this paper, the authors present a novel approach based on continuous 3D real time reconstruction of the monitored area dedicated for video surveillance applications. Real-time 3D reconstruction allows an important network bandwidth reduction in context to sensor nodes sending descriptive information to the fusion server instead heavy video streams. Each node has to support additional processing in order to extract this descriptive information in real-time, which results in video sensors capturing tasks, data analysis, and extraction of features needed for 3D reconstruction. In this paper, the authors focus on the design and implementation of such sensor node and validate their approach through real experimentations conducted on a real video sensor.

Keywords: 3D Real-Time Reconstruction, Multimedia Sensor Networks, Video Capture Devices, Video Sensors, Video Surveillance

INTRODUCTION

Nowadays, Wireless Sensor Networks (WSN) have emerged as one of the most promising technologies (Akyildiz, Su, Sankarasubramniam, & Cayirci, 2002) by reason of the huge potential they represent in several real-world applications, ranging from health care applications to military (surveillance) applications. In fact, recent advances mainly in miniaturised hardware components, radio frequency communication technologies and low power embedded comput-

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addition, nodes are usually subject to frequent failures. Even though these limitations, wireless sensor networks are nowadays present in several applications that are delay tolerant and have low bandwidth requirements.

More recently, the availability of new miniaturized multimedia hardware (CMOS cameras and microphones for instance) at reasonable prices has encouraged the emergence of Wireless Multimedia Sensor Networks (WMSN (Akyildiz, Melodia, & Chowdhury, 2007)) for a large number of real applications (home automation, assistance to senior people, etc.). What differs WSMN from classical sensor network is their ability to collect multimedia content (video and audio streams, still images, etc.) in addition to scalar data.

It is obvious that the management of multimedia data (storing, retrieving, processing in real-time, fusing and correlating, etc.) within limited environments as those of sensor networks poses new issues at all levels of the applications because of the huge volume of the produced data (the size of videos or images is far away much more important than the one of scalar data) on one hand and on the other hand of the “continuous constraint”, also called real-time constraint, related to multimedia data i.e., their delivery is time sensitive. It is obvious that the management and the delivery of multimedia data require much more resources in terms of processing power as well as network resources, mainly bandwidth. Furthermore, the process of multimedia data correlation and fusion differs fundamentally from that of scalar data in the sense that the latter is primarily concerned with average computing while the first is much more complex: what does the fusion of two images or two videos mean? To the best of our knowledge, no general approach is actually available but only some specific approaches tailored for specific contexts as the one reported in Zhang (2007).

In this paper, we develop a new approach to handle the huge and voluminous multimedia sensor data generated in a video surveillance context. The key idea behind our proposition is to “continuously” construct a 3D representation of the monitored area, in which video streams originating from the video sensors are fused. In other words, the “views” of the sensor nodes are merged in the 3D scene of the monitored region. This approach presents many interesting advantages, in particular for resources limited environments like those of sensor networks.

The first important advantage of using 3D representation is its flexibility in the sense that it allows a comprehensive representation of the observed scene. In fact, it is commonly known that exploiting directly raw video data is very hard or even not possible if it is not pre-processed or annotated by experts/annotators (Mostefaoui, 2006). Moreover, it is not obvious to get comprehensive information from tens or hundreds of simultaneously delivered streams, unless focusing only on one or two streams at once. On the opposite, the 3D scene can deliver more comprehensive information to the observer, in particular if the concern is the surveillance purposes. Also, it is easier to track “objects” in 3D environment than on tens or hundreds of monitors!

Another advantage of 3D reconstruction is its ability to answer some spatio-temporal requests that are very hard to handle with raw video data. To illustrate, let’s consider the example of supermarkets in which managers are interested to know the impact of some new department dispositions on the behaviour of customers i.e., measuring the average waiting time of customers at the fish department for example. Unless “manually” measuring customers waiting times through using traditional video surveillance system, no automatic approach is conceivable to answer such queries. However, it is relatively easier to handle such queries in a system based on 3D reconstruction even though its precision is approximate (the application does not require person recognition). Moreover 3D reconstruction allows some “angles of views” of the monitored scene that raw data sensor cannot offer (see our practical example presented in in Figure 9).

In our 3D reconstruction system, each node performs two main tasks: (a) data extraction (processing phase) during which the node
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