Deployment of an IoT Solution for Early Behavior Change Detection

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Abstract. Today, numerous factors are causing a demographic change in many countries in the world. This change is producing a nearly balanced society share between the young and aging population. The noticeable increasing aging population is causing different economical, logistical and societal problems. In fact, aging is associated with chronic diseases in addition to physical, psychological, cognitive and societal changes. These changes are considered as indicators of aging peoples’ frailty. It is therefore important to early detected these changes to prevent isolation, sedentary lifestyle, and even diseases in order to delay the frailty period. This paper presents an experiment deployment of an Internet of Thing solution for the continuous monitoring and detection of elderly people’s behavior changes. The objective is to help geriatricians detect sedentary lifestyle and health-related problems at an early stage.

Keywords: Behavior change · Internet of Things · Frailty

1 Introduction

Aging is often related to significant changes in physical activities, mobility, nutrition, social life and cognitive status. These changes considerably affect elderly people quality of life. According to the World Health Organization (WHO) [1], the biggest health risk for seniors is the adoption of a sedentary lifestyle that causes isolation, depression and many other diseases such as cardiovascular disease, obesity, high blood pressure, etc.

In this paper, we present our experience in deploying an Internet of Things (IoT) solution for the continuous monitoring and detection of elderly people’s behavior changes. The objective is to help geriatricians detect sedentary lifestyle and health related problems at early stage, without the need to perform classical psycho-geriatric tests that have many limitations like assessment inaccuracies.
and the difficulty for elderly people to recall past events. The work was performed as part of the European project City4Age based on 6 pilot sites: Athens, Birmingham, Lecce, Madrid, Montpellier and Singapore. In this paper, we will focus and detail the deployment performed in the pilot site of Montpellier.

2 Literature Review and Related Work

Early detection of ageing people behavior change can improve medical assessments and enable proactive intervention. In fact, aging-related health problems generate long-term behavior changes, such as possible instabilities, variations, impairments, declines, increases or improvements [2]. Nowadays, geriatricians use psycho-geriatric scales and questionnaires to analyze behavior and investigate possible changes [3,4]. These psycho-geriatric approaches are insufficient to monitor patients on a daily basis [5]. Thus, geriatricians need technological services to acquire new objective observations that complete their medical observations. Monitoring technologies can help follow-up elderly people at home and in city, in order to early detect possible health changes [6].

Contrary to existing technological solutions that target Short-term Health Change Detection [7], retrospectively investigate possible changes after change occurrence [8] and use intrusive technologies to capture video sequences, collect daily questionnaire-based information and record physical data using body sensors [9], our proposed approach analyzes overall behavior over long periods, in order to detect long-term changes in health status. These long-term changes require weeks and months to emerge, and are difficult to detect due to normal continuous variation in human behavior [10]. In addition, it continuously analyzes monitoring data on a daily basis, in order to early detect possible changes. This proactive change detection provides opportunity for daily assessment and subsequent intervention. Finally, it uses unobtrusive monitoring technologies that are embedded in our real environment or in objects of daily living, do not interfere with natural behavior of elderly people and do not change their daily habits.

3 Montpellier Pilot Setup

Montpellier pilot site goal is to quickly and unobtrusively detect possible aging people’s behavior changes. Detected behavior changes are afterwards analyzed and confirmed by collaborating geriatricians to provide adequate intervention.

The pilot site is Coordinated by the French National Center for Scientific Research (CNRS) and is collaborating with local authorities such as Montpellier Metropolis, the ETAPE (health autonomy pole) association and healthcare professionals from Beausoleil Clinic and Saint Vincent De Paul nursing home.

The proposed solution deployed in the pilot site of Montpellier consists of a set of sensors deployed in the participants’ homes and in the city as shown in Fig. 1. For indoor monitoring, the pilot uses motion sensors, contact sensors
and bed sensors. Outdoor monitoring is based on Smartphones carried by participants and beacons deployed in places of interest of participants in the city. These sensors allow to collect raw data and objective information in real time, which are processed to detect behavior changes.

Fig. 1. Montpellier pilot site global setup

4 Recruitment and Engagement

Each of the six pilot sites followed a specific ethical approval process, related to specific countries’ regulation, in order to recruit participants. For Montpellier pilot site, the ethical process consists in sending the specifications of the deployment to CNRS’s Data Protection Correspondent (Correspondants Informatique et Libertés: CIL). In parallel, an application was also submitted to the Institutional review board (Comités de Protection des Personnes: CPP).

After obtaining the ethical approval, we started the recruitment process. We have approached around 40 potential participants with the help of ETAPE association and Saint Vincent De Paul nursing home. In fact, we have presented the solution in several local events to promote the project and identify interested people to be included in the study. As shown in Fig. 2, briefing and presentation sessions were organized in the Montpellier pilot site to better promote experimentation objectives, quickly launch recruitment process and keep recruited participants involved in the experimentation. These events allowed to have in depth discussions with interested aging people who accepted to visit a demonstration house, see a live demonstration of the system and have their feedback. The demonstration house highlights the unobtrusiveness of the technological solution proposed for the potential participants. Interested people also observe sensor events in real-time and examples of real data over weeks and months indicating significant changes in health status. Potential participants asked for
information about employed technological solutions, real benefits of adopting them at home and possible risks.

19 participants accepted and have been equipped with the City4Age solution. An initial interview with included participants allowed to collect some indications on their social and health profiles. Participants have diverse medical and social profiles such as educational level, dependence level, habits and health status. Table 1 presents regular habits and health info of some participants.

**Table 1.** Examples of Montpelier pilot site participants habits and health status

| Patient | Regular habits | Health info |
|---------|----------------|-------------|
| 98      | Wakes up at 8 h. Home aid 4 times per day. Stays most often at home. Sometimes goes out with daughter or caregiver | Alzheimer, Diabetes, Vision and audition problems |
| 101     | Wakes up at 7 h30–8 h. Home aid visits 3 times per day (morning, midday and evening). Niece and neighbor visits during the day. Sleeps earlier than before (at 20 h, and before at 22 h) | Alzheimer, Some falls and hospitalizations |
| 102     | Wakes up at 6 h–7 h. Home aid visits each day in the morning. Lives alone. Daughter house is nearby. Monthly visits to and from daughter | Heart problems, Urinary infection |

5 Technologies and Data Collection

The system proposed in Montpellier pilot site uses indoor and outdoor technologies to monitor daily living activities of participants. For indoor monitoring, the pilot is proposing a set of sensors (motion sensors, contact sensors and bed sensors). These sensors operate discreetly and allow to collect raw data and objective information in real time. The objective is to be able to accurately determine the “habitual behaviours” of people by collecting data over time. Outdoor monitoring is based on beacons deployed in the participants’ places of interest in the city.
Deployment of an IoT Solution for Early Behavior Change Detection 31 (e.g. bus and metro stations, cinemas, restaurants, etc.) and smartphones with dedicated mobile application carried by the participants. Thus, it is possible to collect data on the activities of our participants’ even when they are in outdoor in the city. Figure 3 showcases some deployed sensors in Montpellier pilot site.

![Indoor and outdoor sensors deployment](image)

Fig. 3. Indoor and outdoor sensors deployment

All indoor and outdoor data are collected in Montpellier pilot site local server where local treatment for Low Elementary Actions (LEAs) and measures identification is performed. Low Elementary Actions are basic participants’ actions which are inferred from received sensors’ events (e.g. Start Moving, Stop Moving, Change Room, Visit Restaurant, etc.). Measures are quantified data extracted from LEAs (e.g time in the bedroom/day, number of toilet visits /day, number of shops visits/week, etc.). Later, these information are transferred to the City4Age repository and analytic algorithm where further treatments are performed in order to produce visualisations for the geriatrician. The complete architecture of the City4Age solution and the performed deployment and data analysis in Montpellier pilot site is presented in Fig. 4.

![Complete architecture of Montpellier pilot site's deployment](image)

Fig. 4. Complete architecture of Montpellier pilot site’s deployment
The deployment of this system in France allowed to collect around two years of real data. We have collected 310,590 of Low Elementary Actions (LEA) and 49,659 of Measures for 19 participants.

6 Data Interpretation

The goal of the system proposed in Montpellier pilot site is to detect possible behavior changes that will be analysed and confirmed by collaborating geriatricians to provide adequate intervention. A behavior change tracker service [11] was developed allowing to detect changes in participants’ behavior using statistical algorithms. Collected data are analyzed by this developed algorithm and presented to collaborating geriatricians from clinic Beausoleil. Figure 5 showcases some behavior changes detected by the ChangeTracker service for one of the participants in the pilot site. In Fig. 5, 3 consecutive decreases on 2017-02-15, 2017-06-20 and 2017-10-25 are detected for participant 91. Participant and family doctor confirm mobility impairments and increased risk of dependence in managing activities of daily living. Professional caregiver helps with medication taking and household from 2017-05-04.

![Activity level of participant 91](image)

**Fig. 5.** Detected changes in activity level of participant 91 due to mobility impairments

7 Intervention Process

To perform intervention, the pilot site is providing a framework with visualizations about participants activities and statistics about their daily routines and habits. The ChangeTracker service is integrated in this framework allowing to automatically detect possible changes that can be confirmed by the caregivers and the geriatrician. The caregivers and geriatricians can navigate the data provided by these visualizations and decide on the type and form of intervention when needed. As an example, after detecting a decrease in outdoor and indoor activities for participant 96, nursing home stuff decided to initiate home assistance.
8 Validation

8.1 Detection Process Validation

The ChangeTracker service has being validated with the help of our stakeholders. Detected changes are being correlated with medical observation and health records to validate the used algorithms and their performances.

The ChangeTracker service sends change notifications. Change notifications provide opportunity to confirm that detected changes are really permanent and investigate possible correlations with geriatric observations. Regular review meetings with elderly people, family members and family doctors allow to accurately investigate possible causes of detected changes. Review meetings investigate multi-dimensional correlations of detected changes, such as identifying parallel decreases in activity level and time out home related to mobility impairments, and consecutive increases in sleep interruptions and toilet entries after treatment change. Figure 6 shows some results of changes detected by ChangeTracker service and their correlation with medical observations and participant feedback.

![Fig. 6. Detected behavior changes by ChangeTracker and corresponding participant feedback](image)

8.2 Results and Performance

In total, we have detected 340 changes for all participants with an average of 0.97 change per month. Participants show diverse changes in monitoring period that are associate with diverse medical reasons, such as physical problems (45.59% of detected changes), health improvements (36.76%), nutritional problems (8.82%), personal changes (7.35%) and social problems (1.47%).
9 Conclusion

Montpellier pilot site was a proof of concept of our IoT proposed solution for the early identification of behavior changes. In total, 340 changes have been detected for all participants. These changes have been validated and classified with the help of local geriatrician and by correlation with Medical Observations. This solution could be of great value for the geriatricians. In fact, the technological observations provided by the proposed solution enrich their medical observation for better assessments of frailty and MCI. The solution is also valuable for nursing homes. In fact, nursing homes need to maintain independent living at home and identify elderly people at risk who really require entry to nursing home.

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