

1 Nilanjan Dey¹ and Amira S.Ashour²

2 ¹ Techno India College of Technology

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4

5 **Abstract**

6 Information technology advancement leads to an innovative paradigm called Ambient
7 Intelligence (AmI). A digital environment is employed along with AmI to enable individuals to
8 be aware to their behaviors, needs, emotions and gestures. Several applications of the AmI
9 systems in healthcare environment attract several researchers. AmI is considered one of the
10 recent technologies that support hospitals, patients, and specialists for personal healthcare
11 with the aid of artificial intelligence techniques and wireless sensor networks. The
12 improvement in the wearable devices, mobile devices, embedded software and wireless
13 technologies open the doors to advanced applications in the AmI paradigm. The WSN and
14 the BAN collect medical data to be used for the progress of the intelligent systems adapted
15 inevitably. The current study outlines the AmI role in healthcare concerning with its
16 relational and technological nature. Health monitoring and electronic patients? planning
17 assistance applications are reported in the present work. Lastly, the challenges tackled in the
18 AmI technology adaptation in the real world healthcare applications are highlighted.

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20 **Index terms**— ambient intelligence, wearable devices, wireless sensor networks, wireless body area networks,
21 wearable sensors, monitoring systems, computational int

22 **1 I. Introduction**

23 mbient intelligence is associated with the atmosphere at which emotional and rational intelligence is ubiquitous. In
24 the AmI environment, individuals are bounded by embedded intelligent devices' networks to collect information
25 nearby their physical places in order to provide services, and ubiquitous information. Intelligent devices are
26 accessible whenever required through interactions and acting independently to allow high quality information to
27 any user, at any time, on any device, and anywhere.

28 In healthcare environments, these devices are related to medical informatics, decision support, gathered
29 electronic health records, knowledge reasoning and representation, and telemedicine. Patients' medical reports,
30 radiological films, personal and medical information can be observed in remote places. Furthermore, remote
31 robotics can be used in telemedicine and surgery. Nevertheless, these healthcare applications are used for specific
32 clinical situations in certain services with explicit patient.

33 Distributed environment are used mainly with applications that are undeveloped to share actions and
34 knowledge. Recently, AmI is related to the exponential evolution of Internet. For user-friendly applications,
35 web browsers can support several features to allow users to use remote applications.

36 Recently, technology is moving towards Ambient Intelligence (AmI) environments to support inhabitants
37 the daily life . In physical world environments, AmI allows the human to interact in an inconspicuous and
38 intelligent way using computing devices with complete awareness of the people requirements and forecasting
39 behaviors. AmI environments are essentially local including hospitals, offices, homes, transports, and control
40 centers. Recently researchers are interested to comprise extra intelligence in the AmI environments for
41 superior access to the indispensable knowledge and decision making that support individuals. Typically,
42 AmI is related to several concepts, such as context awareness, embedded systems, and artificial intelligence
43 to incorporate with other techniques including computer graphics, automation and communications. Several
44 challenges faces the AmI requiring advanced techniques such as the artificial intelligence (AI), machine learning,
45 computational intelligence, computer vision, and intelligent robotics that have been used in several applications
46 [22][23][24][25][26][27][28][29][30][31].

AmI technologies have several features including transparent, adaptive, sensitive, intelligent and ubiquitous. Researchers reported different AmI definitions, namely i) AmI is an emergent multidisciplinary domain founded on ubiquitous computing that influences the protocols design, devices, systems, and communications [32], ii) AmI offers new interaction ways between people and technology to serve the environment and individuals' needs [33], and iii) AmI is a new research domain for non-intrusive, distributed, and intelligent software systems [34]. AmI is involved in several applications especially in the medical sector to improve the healthcare by developing for example an inclusive structured approach to electronic medical record (EMR) toward intelligent healthcare units [35]. Furthermore, it can be employed to augment smart hospital rooms that support both the medical staff and patients [36]. In order to achieve these features, wired/wireless sensor technologies are assimilated, tolerating the patient to interact and to control the19 Year 2017 () H

hospital services. In addition, the clinical oriented interface can allow vital sign monitoring. Such wireless technologies include the use of Bluetooth, Zigbee, WiFi, and RF (radio frequency along with intelligent sensors. This AmI environment integrates several software and hardware technologies for manually or automatically controls electronic and electrical devices.

Smart environments and wireless technologies have an ultimate role to provide user-friendly tools to cope with the surrounding environment. Mahoney [37] established that smart environment can be considered ease the hospitals staff workload when supporting the AmI. These technologies endorse the clinical care quality with patient's independency, superior life quality and improving the health care quality. Several applications that support miscellaneous clinical requirements include condition unambiguous treatment and diagnosis, patient's remote monitoring and softcopy radiological film review. Moreover, assistive environments, such as RFID based smart hospitals, environmental sensors and monitoring cameras can be included to support healthcare. Such environments require integrated interconnect services in the automation systems, e-health systems and sensors through binding architecture [38].

The structure of the remaining sections is as follows. Section 2 introduces the wireless sensor networks role in the AmI systems. Section 3 reports several studies that have employed the artificial intelligence to develop advanced AmI systems. Conducting decision making, healthcare interactions and monitoring in AmI environments are addressed in section 4. The challenges face the AmI technology in healthcare are highlighted in section 5. Finally, the conclusion is considered in section 6.

2 II. Wireless Sensor Networks in AMI Systems

The Wireless Sensor Networks (WSN) is a compulsory technology for developing the AmI environment via providing the users by services based on their context. These networks offer flexible and dynamic structure for the acquired data transmission from the environment through sensors. The transmitted data is considered the base for developing the AmI services adjusted with the acquired information from the sensors system in charge of handling this information. The WSN allow information gathering about the environment and the user.

The foremost characteristic of the WSN is to transmit wirelessly the acquired data by sensors in diverse environments to other nodes for processing this data. Since the WSN consists of massive number of nodes, these nodes require special design characteristics, namely low power consumption, small size, low cost, and low complexity. Furthermore, network topology and protocol are considered during the WSN design to simplify the nodes functionality with less time consuming to reduce the power consumption. In the Wireless Body Area Networks (WBAN) for healthcare applications, the star topology is used mainly, where a central node coordinates the communication with outside the Body Area Networks (BAN) and the medical sensors [39]. Generally, the BAN has compact units responsible about transferring the vital signs from the patient's body and the physician or hospital. Several applications employed the BAN to monitor the patient's state, such as i) MobiHealth monitoring system for vital signs based on a BAN [40] and an m-health service stage using communication via Bluetooth between the central device and the intra-BAN, ii) WBAN VitaSens system including ECG, blood pressure, respiratory, and temperature sensors [41], and iii) Ubimon ubiquitous monitoring system for implantable and wearable sensors including accelerometer, ECG, humidity and temperature sensors [42].

Through the AmI paradigm, the way to offer the information to the society services and users is developed by including the Internet services. Services can be accessible by the user as well as by the system intelligence tolerate automatic delivery of the services. The user interaction with the services will be over interfaces. Consequently, in order to realize services based on this paradigm, the user context should be known requiring sensors in the patient's body and in the environment. Such facility can be acquired using AmI services to compromise the adopted service via the natural interfaces. Likewise, the interoperability and integration of these networks with Local, Personal, and other networks configurations become essential.

Hospitals and medical centers offer traditional healthcare services. For the scientific community, finding active methods to improve healthcare become challenging issue.

Post-surgery monitoring is considered one of the vital needs. Furthermore, patients need to contact their doctors easily. Nevertheless, traditional solutions for these aspects are inconvenient, costly, and inefficient for the patients for routine checks. Thus, E-health aims mainly to improve the health care quality, and to enhance the health care effectiveness. In order to direct the healthcare services from the hospital environment to the home, AmI environment become essential for personalized healthcare and for healthcare monitoring. AmI has

108 been involved in several healthcare applications, such as computer-assisted surgery systems to remote surgical
109 conduction with reduced risk [43], virtual reality systems to treat the anxiety disorders [44].

110 AmI allows a physical connection between the patient's daily practices and the e-health systems using wearable
111 medical devices, smart environments software techniques. The context embraces the environment and the users'
112 information, which contain different parameters, such as temperature, light, blood pressure, and heart rhythm.
113 Different WSNs technologies, such as ZigBee; Bluetooth or Radio Frequency Identification (RFID), are employed
114 to gather the AmI required information. Development of AmI systems requires also dynamic methods and
115 mechanisms based on artificial intelligence (AI).

116 3 III. Artificial Intelligence in AMI Systems

117 Typically, AmI environment is profound to the living creatures' existence in it, ropes their activities and
118 anticipates/remembers their actions [45]. Consequently, for health claims in AmI, data collected from vital-
119 sign sensors plays a significant role. Several computational methods based AmI in vital sign sensors are
120 developed. Sensor data analysis requires distributed/centralized models with the AmI systems [46]. Each sensor
121 has committed processing abilities to perform local computation before transferring the data to other nodes in
122 the WSN.

123 Numerous data mining and artificial intelligence (AI) techniques are used to analyze the sensors data in the
124 AmI systems, including fuzzy logic rules, neural networks, machine learning and decision making. Such techniques
125 assist AmI in healthcare monitoring. Several studies for developing AmI systems in healthcare have been carried
126 out. Activity recognition system based on artificial neural network (ANN) has been conducted to regulate the
127 falls occurrence using single sensor positioned on the individual's chest [47]. The results established that the
128 ANN entails more tuning factors compare to the support vector machines (SVM).

129 A GerAmI system has been settled with the Alzheimer Santísima Trinidad Residence of Salamanca that
130 used sensors to record the patients'/users' data, where the user wore an armband holding a RFID chip for tracking
131 individuals [48]. In the case of required assistance, a message containing the patient's name, the occurred problem
132 and information about the paramount way to handle this situation is directed to the staff members PDA. In
133 patients with Parkinson's disease, in order to predict clinical scores of data severity obtained from wearable
134 sensors, a SVM has been implemented [49]. On a single environment, an AmI application has been outfitted with
135 sensors and deliberated to improve the resident experience in the environment [50]. A Hierarchical Task Network
136 (HTN) planner has been employed to produce actions sequences and eventuality plans to realize the aim goal of
137 the AmI system [51]. The AmI system may react to a sensed health necessity by calling the medical professional
138 and transferring health vitals through any communication device/tool such as email, or cell phone.

139 In smart health environments, a study has been carried out to proof the architecture concept for emotion
140 regulation and detection of the patients through the analysis of their facial expressions, behavior and vital
141 signals [52]. Another study has been applied using the insulin dosing, glucose levels, sleep state, and physical
142 activity data gathered from body-wearable sensors to detect type 1 diabetes [53]. An open research domain
143 is directed toward the techniques used in order to acquire self-reported data and to integrate sensor-provided
144 information from the sensor networks in the AmI system. Subsequently, a platform for collecting and integrating
145 data from service providers and sensors into one cohesive format for further use in the experience sampling
146 methods (ESM) has been implemented [54].

147 For cognitive-related pathologies, a game has been designed for the analysis and evaluation of the frontal
148 brain activity via the videogame mechanics identification that include EEG brain activities associated to some
149 cognitive skills [55]. Furthermore, another study, on the brain signals to recognize emotions, has been proposed
150 using the neural network [47]. Several machine learning methods have been used to evaluate a predictive system
151 performance that deals with the in-hospital patients' mortality. These patients undergo overhaul of an abdominal
152 aortic aneurysm.

153 4 IV. Decision Making, Healthcare Interactions and Monitoring 154 in AMI

155 In order to conduct fully automated AmI applications, decision-making techniques are employed. Several studies
156 have been carried out to implement decision making based AmI systems. Temporal reasoning has been used with
157 a rule-based system in order to recognize hazardous states with decision making that resolve this situation and
158 return the environment to a nonviolent status while communicating the place's residents [56]. Added, deleted,
159 and modified fuzzy rules have been learned via observing the resident behavior in the iDorm application to adapt
160 the environment according to the changing behavior.

161 Another AmI system based on decision making has been conducted to design a planning system supported
162 by artificial intelligence to remind entities by their next daily activities as well as the incomplete tasks [57]. A
163 hierarchical task network planner based AmI system has been proposed to produce plans of actions' sequences for
164 responding to sensed health requirements by contacting medical professional and sending health vitals through
165 email, cell phone, or fax [51]. Patients suffer from Alzheimer's disease and other disorders are also supported
166 by developing an AmI system that help individuals to perform their regular errands through sensing their

167 location/environment and offer decision making to forewarning caregivers in the critical situations [58]. In
168 medicine, to in order to regulate optimal decisions sequences, Markov-based 21 Year 2017 () H

169 method has been used to describe the dynamic sequential decision making process [59].

170 Based on communication technologies and information, ambient systems can assist and enhance the life quality
171 of individuals at home and anywhere. This promotes the services/infrastructures development toward autonomous
172 life through the incorporation of the communication technologies and information through ambient intelligence
173 in healthcare applications. In such domain, ubiquitous systems, wearable sensors and secure mobile can be
174 engaged to improve the life health quality. Automated AmI systems require universe technologies to achieve the
175 interaction between the patients and the physicians.

176 A theoretical framework has been proposed to support this interaction process [60]. In the AmI design,
177 different measuring tools for the patient's rendezvous in technology progress and for testing the effectiveness
178 of AmI prototypes have been used. Other interactive environment has been proposed for people rehabilitation
179 with physical disabilities [61]. Due to the intuitive interaction of users, direct, and natural features of the AmI
180 systems, it can be employed to recognize and predict the individuals' activities and can be involved with services
181 and applications embedded in the surrounding environment. Based on inertial wearable sensors, a collective
182 dataset for human gait has been gathered of further analysis [62]. A passive vision-based system to estimate the
183 measurements of gait using light sensor along with 3D point-cloud has been proposed in order to explore the gait
184 analysis of the wearable system that has 2 wireless sensors for acceleration fixed on the ankles [63].

185 AmI systems have been also supported the daily life healthcare through homecare assistance. A computational
186 detection technique to quantify the changes in physical activity patterns using wearable sensor data has been
187 proposed [64]. This technique can be applied to detect inadvertent changes in the individuals' patterns
188 performance to validate the effect of any new healthy behavior on the individual's lifestyle. Another ambient
189 system has been framed to promote social commitments and activities of the elderly individuals using in-home
190 sensors by linking the information inferred with the social network [65]. This provides the old individuals with the
191 chances to make new social networks. In the daily life environment, another AmI system application to support
192 old people has been proposed to assess the fall risk assessment with preventing its occurrence using wearable
193 sensors [66].

194 5 V. Wearable Sensors Devices in AMI Systems

195 AmI is an evolving restraint that passes intelligence to the individuals' daily life environments and creates sensitive
196 environments to the human needs. Its main idea depends on enriching the surroundings with technology, namely
197 interconnected devices to a network and sensors leading to a system that can take useful decisions to the users.
198 Such decisions depend on the gathered real-time historical information and accumulated data. Based on the
199 development in several technologies and areas, including networks, sensors, artificial intelligence, human computer
200 interfaces and ubiquitous computing, AmI environment grows quickly.

201 AmI has several applications in the healthcare domain, including: I) Human fall detection by evaluating the
202 wearable sensors data such as that obtained from the accelerometers. In this application, numerous algorithms
203 and sensor's data transmission and localization are developed. In addition, an automated call to the person's
204 relatives can be started with the fall situation detection. II) Human activity recognition that identifies the user's
205 activities if he/she wears accelerometer(s). Such system requires machine learning procedures for analyzing the
206 data received from the sensors. Numerous procedures for studying the sensor body locations in different positions,
207 such as the ankles, chest, wrists, and thighs can be involved. III) Human stress detection, which detects the
208 users' stress level using also machine learning procedures for data analysis data is considered one of the important
209 AmI applications. IV) Automated human energy expenditure estimation can be carried out by analyzing the
210 data from different wearable sensors, including the galvanic skin response, heart rate, and accelerometer. This
211 system is also requires machine learning methods to estimate the energy expenditure.

212 From the preceding addressed applications of the AmI systems in the healthcare domain, it is obvious that the
213 wearable sensors have the main impact in all systems. Wearable sensors have monitoring as well as diagnostic
214 applications using their biochemical/ physiological capabilities. Physiological monitoring can be involved in
215 diagnosis and enduring treatment of a massive number of patients, who suffer from cardiovascular, neurological,
216 and pulmonary diseases including hypertension, seizures, asthma, and dysrhythmias. Sensors are positioned and
217 arranged in consistent with the clinical application under concern. For example, sensors for monitoring vital
218 signs, such as the respiratory rate and the heart rate, can be arrayed when the monitored patients with chronic
219 disease or congestive heart failure suffering clinical intrusion [67]. Sensors for capturing movement data can be
220 used in monitoring applications effectively, such as in the homebased therapy interventions in stroked patients
221 or to detect the elderly people mobility. Figure ?? demonstrated the monitoring system structure based on
222 wearable sensors that attached to the users' body. In order to transmit the collected sensors data, wireless
223 communication is trusted to convey the patient's data to any access point or mobile phone and communicate
224 the information to an inaccessible center through the Internet. Detecting emergency situations using data
225 processing systems are implemented to send an alarm message to the emergency service center for instantaneous
226 assistance to the patients. Consequently, the wearable system consists of main parts, namely i) data sensing and
227 collecting hardware sensor device, communication system of software/hardware to transmit the sensed data to
228 a remote center, and finally the data analysis and processing technique based on machine learning in order to

229 extract the significant and clinical information. Subsequently, development in the telecommunication, electronic
230 circuits, sensors technology, microelectronics, and data processing and analysis methods is directly reflected to
231 an improvement in the wearable systems for several healthcare applications.

232 The prominence of incorporating large scale wireless tele-communication tools, including the WiMAX, Wi-Fi
233 Mesh, and 3G with tele-medicine attracts several researchers. Such technologies integration can be employed
234 for endless people's monitoring who suffer from cognitive disorders, such as Parkinson's, and Alzheimer's.
235 Furthermore, there is an important research studies that focused on the tiny wireless sensor devices development
236 that integrated into wearable materials, fabrics or can be entrenched in the human body. Currently, the range of
237 implantable and wearable biomedical devices increases due to the developments in the digital electronics, wireless
238 communications, and micro-electro-mechanical systems (MEMS) technology. This provided multi-functional, low
239 power and low-cost sensor nodes having small size and can interconnect through short distances. In addition,
240 tiny sensor nodes will be applicable that have the advantage of the sensor networks depending on cooperative
241 strength of a large nodes' number.

242 6 VI. Ambient Intelligence Challenges in Health Care

243 Ambient intelligence based on wireless communication and wearable devices has a significant role in several
244 applications [68][69][70][71]. The ability to resolve to individuals requirements along with bridging the gap
245 between devices, environment and individuals with widespread practice in changes management, innovation and
246 knowledge sharing, lead to the presence of several challenges in the AmI environment. More challenges and
247 limitations are raised in the healthcare area. Since AmI is considered a conception at which the environment
248 ropes the individuals using embedded sensors and processors.

249 Wearable devices and handheld devices considered as interface between the system and user that allow the
250 system to adapt based on the user's behavior. In order to realize higher quality health care environment including
251 hospitals, homes, and medical centers, more effective healthcare systems based on the AmI technology become
252 compulsory to handle inbed/wheel chair patients and many other critical cases. Massive progress in medicine
253 and living circumstances increased the life expectations.

254 Intensifying healthcare cost and lack of healthcare experts poses a difficulty in today's society. Sensor
255 equipment and communication equipment are engaged mainly to handle the challenges in the healthcare
256 environment. In addition, artificial intelligent can produce a self-satisfying life style. In AmI systems, the
257 detection the life threatening situations and rapid the time response in emergency cases. The information
258 technology progress supports healthcare institutions including healthcare centers and hospitals to operate
259 competently with saving cost.

260 Identifying the shortage in healthcare information management is considered one of the critical issues to
261 develop AmI systems for based on advanced AI techniques. Furthermore, proposing active communication
262 system to handle time-critical situations is considered one of the challenging problems in the AmI environment
263 to improve healthcare delivery using distributed intelligence systems. Designing wearable electronic devices to
264 assist chronic disease patients to know the correct medication, to give reminder, and to contact relatives in the
265 critical situations as well as to suggest the appropriate diet in order to manage their health conditions become
266 one of the new research challenges.

267 Transcription problem of can be considered as a problematic of transforming one information form to an
268 alternative or from one storing system to another. Thus, human resources must be allotted to copy the record
269 from one format to an alternative, which is time consuming, error prone, costly and challenging for the real
270 world AmI in healthcare. Another challenging aspect is the end user acceptance and usability of the complicated
271 designed user interface. For potential assistance for the healthcare provision, AmI systems have been identified
272 for all individuals' categories, however old peoples have limited experience to adopting technologies. Thus, in
273 advance of organizing developed technologies, it is important to assess their acceptance. Moreover, the intensive
274 care unit can be considered a complex system that includes massive health information and severe healthcare
275 system components, including environments, patients, and tasks. Thus, in order to improve the patient outcomes
276 and to enhance the health information technology, critical care delivery competence and the patients' safety,
277 superior interaction in the intensive care units systems with the different healthcare information technology
278 components become significant.

279 The AmI healthcare applications can be categorized into personal healthcare/wrist-worn monitoring devices,
280 and institutional healthcare providers' aspects, including hospital environment sensor localization. In environ-
281 mental intelligence, although in healthcare applications, research is going ahead, it does not grasped yet the
282 maturity level due to the challenges in healthcare domain raised by the computer scientists as well as the
283 difficulty to handle the critical situations at which errors are intolerable. Several research openings can be
284 directed to explore the role of the AmI systems in several applications to support healthcare, including:

285 ? Data analysis for health AmI environments Furthermore, one of the main hurdles to the sensing technology
286 implementation, expressly for the wearable applications, is the sensors' size and the frontend electronics, which
287 used to collect the movement and physiological data in the applications of long-term monitoring. Modern
288 progresses in the microelectronics domain, allowed the engineering and researchers to improve minute circuits
289 with front-end amplification, sensing ability, radio transmission, and microcontroller purposes. Improvements in
290 the manufacture technology of the micro-electromechanical systems (MEMS), enables the miniaturized inertial

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sensors progress, which can be involved in the detection of the motor activities as well as other health cases monitoring systems. In addition, batch fabrication methods are challenging and have substantial reduction in the sensors' cost and size and cost. Moreover, microelectronics is recently depending on the integration of several components, including the radio communication circuits and the microprocessors, into a single unified circuit leading to the implementations of System-on-Chip [72].

7 VII.

8 Conclusions

The AmI technology is considered a new paradigm for upcoming applications in the information society proposing intelligent services based on the user context through interactive interfaces. Ambient intelligence has an emerging role in healthcare. The paper presented a snapshot of the AmI system related technologies as well as some empirical studies in healthcare. Robust foundation for the integrated AmI systems implementation is clearly reported based on the wearable sensors, information and communication technologies advancement to support healthcare. Several challenges are highlighted to inspire the researchers toward the AmI technology as a starting point for progress to provide effective AmI systems in healthcare. Such challenges include elder users' acceptance to this new technology, intelligence ¹

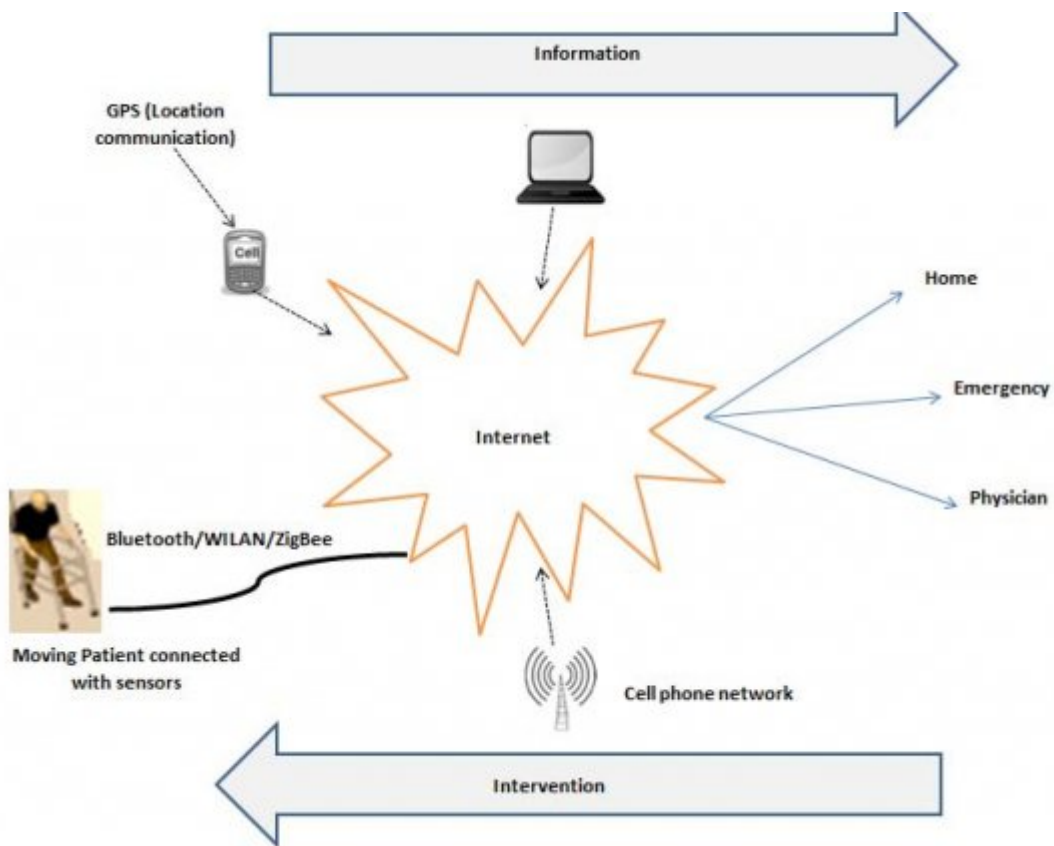


Figure 1: AmbientFigure1:

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[Note: communication]

Figure 2:

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8 CONCLUSIONS

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