Thematic Dynamics of Internet of Things (IoT): Impact on Digital Personalized Healthcare (PHC)

Bongs Lainjo (✉ bsuiru@icloud.com)

Cybermatic International, Montréal, QC Canada

Research

Keywords: Internet of Things, Data Mining, Artificial Intelligence, Machine Learning, Data Transformation, Personalized Healthcare, Thematic Trends, Thematic Dynamics

Posted Date: August 3rd, 2021

DOI: <https://doi.org/10.21203/rs.3.rs-704108/v1>

License: This work is licensed under a Creative Commons Attribution 4.0 International License.

[Read Full License](#)

Abstract

Background: Information technology has continued to shape contemporary thematic trends. Advances in communication have impacted almost all themes ranging from education, engineering, healthcare, and many other aspects of our daily lives.

Method: This paper attempts to review the different dynamics of the thematic IoT platforms. A select number of themes are extensively analyzed with emphasis on data mining (DM), personalized healthcare (PHC), and thematic trends of a select number of subjectively identified IoT-related publications over three years. In this paper, the number of IoT-related-publications is used as a proxy representing the number of apps. DM remains the trailblazer, serving as a theme with crosscutting qualities that drive artificial intelligence (AI), machine learning (ML), and data transformation. A case study in PHC illustrates the importance, complexity, productivity optimization, and nuances contributing to a successful IoT platform. Among the initial 99 IoT themes, 18 are extensively analyzed using the number of IoT publications to demonstrate a combination of different thematic dynamics, including subtleties that influence escalating IoT publication themes.

Results: Based on findings amongst the 99 themes, the annual median IoT-related publications for all the themes over the four years were increasingly 5510, 8930, 11700, and 14800 for 2016, 2017, 2018, and 2019 respectively; indicating an upbeat prognosis of IoT dynamics.

Conclusion: The vulnerabilities that come with the successful implementation of IoT systems are highlighted including the successes currently achieved by institutions promoting the benefits of IoT-related systems like the case study. Security continues to be an issue of significant importance.

Introduction

Technology has continued to shape contemporary trends. Advances in communication technology have impacted almost all sectors ranging from education, engineering, healthcare, and many other aspects of our daily lives. Specifically, the impact and importance of technology in healthcare have been well documented (e.g., Rahmani et al. 2015; Khan 2017; Alotaibi and Federico 2017). For instance, in their article that reviewed the impact of health information technology on patient safety Alotaibi and Federico (2017) observed that there should be no doubt that health information technology is an important tool for improving healthcare quality and safety. It was observed that the use of technology in its various forms, including the use of sensors, the internet, and monitoring systems, helped avoid medication errors, reduce adverse drug reactions, and improve compliance. These views have extensively been addressed in the literature and have even been adopted in healthcare policy (Storey 2013 and Frantzidis, et al. 2010).

While the application of technology in healthcare is commonplace, changing technologies continue to improve and stretch available possibilities. The current literature review supposes that the internet of things (IoT) is a somewhat new technology with considerable applications. According to Kulkarni and Sathe (2014), it was observed that the fields of computer science and electronics continue to interface to

produce revolutionary technologies. The IoT is viewed as having significant potential in a myriad of themes. Fundamentally, IoT facilitates communication between various electronic systems, especially sensors, to facilitate applications of different systems. IoT has been described as having the ability to connect everyday objects such as household electronics, sensors, and actuators through a network. The way such objects connect and intelligently communicate with each other has been otherwise illustrated as a web of objects with inimitable identifiers that can converse with each other with or without the assistance of a computer or the internet (Darshan & Anandakumar 2015 and Kulkarni & Sathe 2014).

Importantly, IoT systems enable new forms of intelligent communication between devices and people, the environment and devices through sensors, and between things themselves (Kulkarni & Sathe 2014). Enabled by a range of sensors and actuators, and linked through a network, IoT facilitates device systems that can gather and share information directly with each other and the cloud. Such a network makes it possible for supported systems to collect and record significant amounts of data. Gelogo, Hwang, and Kim (2015) even mentioned that such a capability for systems to communicate remotely over the internet allows devices to record and analyze new data streams faster and more accurately. Overall, and as collaborated on by Darshan and Anandakumar (2015), IoT facilitates the creation of smart objects; in a sense, they can gather data and give meaningful information or even facilitate various actions. The illustration in Figure 1 provides a typical process diagram of the components of IoT. Notably, the ecosystem leads to a decision-making tool or an information system: a conceptual and logical flow of IoT systems. Essentially, the illustration provides components that communicate to provide meaningful information. (Alotaibi and Federico 2017; Kulkarni & Sathe 2014; Gelogo, Hwang, and Kim 2015).

The current review supposes that data mining is at the core of IoT – artificial intelligence (AI), machine learning, and data extraction. IoT facilitates convenient and autonomous data mining making it ubiquitous and extensively applicable in many disciplines including manufacturing and supply chain management. For example, He, Xue, and Gu (2020) noted that the increasing adoption of IoT is triggering what is described as Industry 4.0. According to MIIT (2015) (As cited in He, Xue, and Gu 2020), Industry 4.0 entails the integration of next-generation information technology to form new production modes, new industrial forms, and business models. Fundamentally, it is claimed that IoT has changed the environment in which manufacturing, supply chains, resource management, social models, and networks operate (He, Xue, and Gu 2020; López, Ranasinghe, Harrison, and Mcfarlane, 2020).

In a review of the potential of IoT in smart city development, Simmhan et al. (2017) noted that smart cities and homes are a practical manifestation of IoT. IoT systems allow smart cities to achieve intelligent management of operations, efficient transportation, better security, and better access to information. Furthermore, the sensor network enabled through internet IoT continues to play a critical role in education and research by facilitating more accurate and low-cost data collection options and enabling data analytics. Notably, the data mining ability of IoT and the ability of IoT systems to communicate remotely provide significant potential to explore trends (Rainie & Anderson, 2017). Additionally, the ability of IoT to use a variety of sensors and complex algorithms to analyze data and share it through a network has facilitated various engineering, transportation, and research activities (Kimsey et al. 2015; Simmhan et al.

2017). For instance, Kimsey et al. (2015) mentioned that students continue to use low-cost, open-source IoT platforms to create their applications and conduct research.

IoT-based systems have also facilitated security and law enforcement applications. According to Tundis, Kaleem, and Mühlhäuser (2020), IoT systems – a combination of various monitoring systems that range from mobile applications with the capability to provide critical data, to body cameras for law enforcement officers – contribute to safer spaces. IoT-driven, remote, and non-invasive monitoring systems have been applied in specialized healthcare, research, supply chain management, and even sports (Kulkarni and Sathe 2014).

Notably, there has been a growing body of knowledge regarding the use of IoT in personalized healthcare (Deng et al. 2017; Darshan and Anandakumar 2015). For example, Darshan and Anandakumar (2015) observed that IoT-facilitated remote health monitoring systems have enormous benefits over customary health monitoring systems. Acquiring healthcare data, especially for personalized healthcare, can be challenging. While commenting on the matter, a study by Gelogo, Hwang, and Kim (2015) mentioned that IoT supports a variety of sensors and complex algorithms that can be used to analyze the data and share it remotely over the internet. Such an ability may mean that individuals can access medical care remotely. That is, they can be monitored from their respective homes, as an example.

Another example of how IoT can facilitate personalized care is through the use of body sensor networks (BSN). In their analysis, Deng et al. (2017) noted that personalized care can be availed to elderly people through wearable body sensors. Such sensors collect physiological data, as coded for different patients, and relay the data to a healthcare provider. These features have facilitated improved access to care, increased quality of care, and most importantly, reduced the cost of healthcare (Kulkarni and Sathe 2014).

While it is possible to appreciate the significant opportunities for ubiquitous IoT, personalized healthcare (PHC) is amongst the most notable successes of the technology. Advances in IoT-enabled PHC range from applications in personal health management by individual cases through the use of smart sensors to diagnose where body sensors are used to relay data to healthcare providers. For instance, in diabetes therapy management, Jara, Zamora, and Skarmeta (2011) demonstrate a case where IoT personal care devices are used to facilitate getting the right insulin infusion calculation needs for individual patients. It is observed that sometimes many individual factors and conditions can complicate getting the right dosage for patients. However, personal IoT devices (such as radio-frequency identification (RFID) tags) can be developed to assist and consider more individual factors in the insulin therapy dosage calculation.

In similar cases, Serna, Pigot, and Rialle (2007) and Datta et al. (2015) discussed how IoT-enabled smart homes facilitated personalized healthcare. Serna, Pigot, and Rialle (2007) noted that smart homes, which are fundamentally enabled through IoT and data mining, provide support to cognitively impaired people. Giving a case of patients with Alzheimer's disease, the article presents findings on how IoT-based data mining models were used to simulate the progression of dementia of the Alzheimer's type by evaluating performance in the execution of an activity of daily living (ADL) in a smart home environment.

The foregoing cases demonstrate some of the capabilities of IoT-enabled data mining in developing personalized healthcare models and interventions. From near body sensors that collect physiological data to smart homes complete with networked smart devices, it is possible to argue that the potential of IoT in facilitating PHC will continue to grow. Most of the literature reviewed seemed to suggest that the integration of IoT in healthcare will continue to evolve. The documentation appears to emphasize the need to expand the knowledge in this area especially following the sensitivity of healthcare data. That is, increased use of interconnected devices in healthcare touches on the integrity of otherwise confidential health data.

Trends in IoT-enabled personalized healthcare appears to indicate that going forward, just like the way other sectors continue to limit personal interaction in service provision, IoT may enable remote diagnosis and other personalized healthcare services. For example, Amendola et al. (2014) demonstrated the use of IoT in data collection including such data as personal movements and gestures, and how such data can be used for human behavior analysis. The article also showcases the use of wearable and smart implants developed from RFID technology and supported through an IoT system, with the capability to provide services and enough read ranges to implement a network of sensors for tracking personalized human wellness and monitoring the quality of the local environment.

Despite considerable success in IoT development, there have been growing data security concerns (Chacko and Hayajneh 2018; Eken and Eken 2016). For example, Deng et al. (2017) explain that a malicious attacker could send incorrect sensing data to a medical reader causing an incorrect diagnosis. The authors highlight some of the security requirements for implementing secure IoT. Nonetheless, it is contended that IoT-based systems will continue to play a critical role in society and will also continue to face several challenges especially regarding data security (Deng et al. 2017; Chacko and Hayajneh 2018). Sensor technologies are continuously improving, indicating the increased potential for IoT-based systems. Also, developments in connectivity, especially the 5G network, is another indication of the potentials in IoT-based systems.

The general objective of this paper is to conduct a thematic analysis of the IoT dynamics including trends and rankings of the most number of apps over four years; descriptive statistics are also part of the research in addition to modeling highlights of an illustrative life digital personalized healthcare (PHC) study. A generic IoT ecosystem is also presented.

Method - Thematic Review

The methodology used in this paper is primarily using secondary data provided by Google Scholar. The data are used in conducting quantitative analyses as part of the study. The illustrative case study included is a successful implementation of an IoT system by the methododist hospital in Texas.

During the period of July 21 and 24, 2020, a four-year retrospective (2016-2019) database was compiled through searches using the Goggle Scholar framework. The objective was to develop a data set based on the number of IoT publications (apps) that will help in demonstrating the evolution of different IoT

themes. In the absence of reliable IoT predictive models, complicated by the lack of representative data sets, this also serves as a proxy for future IoT innovations. It also needs to be highlighted that the themes are not necessarily mutually exclusive. The four years were selected based on the likelihood of having access to the most recent and most updated compelling data sets. It was also important to have a minimum number of points to facilitate the identification of possible trends during that period.

In general, there were 99 themes in the review. Among these, and based on four-year means and the ranking conducted, the theme with the highest number of IoT-related publications (IP) was "social models and networks," with 251,000 publications. At the bottom, the theme with the least number of IP was "new human-device interaction," with two publications. "Healthcare" came in at 42nd with 12,878 IP. The 18 highest number of IP are presented in Figure 2. The relatively low ranking of "Healthcare" notwithstanding, there is available evidence attesting to the potential robust evolution. The driving force behind this potential and robust uptake is influenced by the implications of cost optimization effects.

The annual median IP for all the themes over the four-year period were 5510, 8930, 11700 and 14800 for 2016, 2017, 2018 and 2019 respectively. The positive overall trends, as shown in Figure 3a, suggest the evolving nature and compelling degree of interest in the general population.

As part of the review process, the author further analyzed the annual inter-thematic IoT-related publications: percentage differentials. The outcome (see Figure 3b) is a relative set of patterns of a select set of themes – eighteen in all, during the period from 2016 to 2019. Localization technologies portray a negative trend in publication, particularly in the period of 2016-17, where there is a greater percentage differential of 45 per cent, with 2017-2018 having the least with 24 per cent. Similarly, social models and networks also portrays a negative trend, more in the 2018-2019 year of publication as compared to 2016-2017 and 2017-2018, with the former having the least at 20 per cent. For smart cities, there is minimal to no interest in the three periods, showing a lacking interest within the three publication years.

Identity management similarly relays a negative trend where the 2018-2019 publication year is dominant with 49 per cent compared to the two periods; 2016-2017 and 2017-2018. With offices, there is a positive trend with all the three years of publication, with the 2016-2017-publication year portraying a larger percentage differential of 47 per cent. Similarly, sports also have a positive trend with an increased overall percentage differential compared to offices. The 2016-2017 year seems to dominate the other two publication years due to its larger percentage differential of 54 per cent, followed by 2018-2019 with 26 per cent.

Resource management shows a negative trend with the 2017-2018 year of publication, dominating the other two with a differential percentage of 55 per cent. Supply chains have the highest differential percentage compared to all the other IoT themes. This is probably because it is amongst the most researched themes. Here the year 2016-2017 is dominant with 49 per cent as more research seems to have been conducted during this period, followed by 2018-2019 at 27 per cent. Manufacturing shows a positive trend with more work coming from the year 2016-2017, followed by 2017-2018 and finally, 2018-2019 with 16 percent.

Digital Personalized Healthcare Study Model

Introduction

This section demonstrates an application of IoT in automating personalized healthcare. It serves as an illustration of the complexities and nuances involved in developing and implementing practical, effective, economical, and pragmatic strategies in an evolving healthcare environment. It also serves as a confirmation of how clinical staff, from nurses to physicians to support staff, can collaborate effectively while minimizing cost, improving productivity, and enhancing teamwork: an invaluable productive approach.

Model Objective

The objective of the model was to develop an automated prediction and diagnosis system. This system was intended to facilitate data access for hypothesis testing, cohort identification, data mining, risk prediction, and clinical research training.

Model Target Audience

The model targets physicians, caregivers, and other clinical staff, as well as patients. It is designed to integrate a clinical informatics framework that contains a data and logic storage EDW (enterprise data warehouse) and a clinical outcome prediction tool, SIA (software intelligence and analytics), for caregivers and allows the integration of various data mining applications that facilitates remote monitoring and control of a patient's physical state from data collected. For example, the MOCHA mobile application that is part of the developments under Methodist Environment for Translational and Outcomes Research (METEOR), uses mobile sensors and integrates with applications such as Fitbit to automatically track activity features like steps, distance, calories burned, and active minutes of cancer patients (Stubbins et al. 2018). (See Figure 2).

Inception Phase

The initiative was launched in July, 2012. The initial phase of the project, integration of health information system (HIS) and other dynamic patient data, was completed in June, 2013 after integration of vitals, medications, and orders. This provided a foundation for the development of other components.

Model Design

The Methodist Environment for Translational and Outcomes Research (METEOR) framework consists of two components: the enterprise clinical data warehouse (EDW) and a software intelligence and analytics

(SIA) layer for enabling a wide range of clinical decision support systems to enhance clinical research and outcome studies.

The EDW consists of organized, enterprise-wide longitudinal clinical, administrative, and research data. The SIA is embedded with functional modules capable of supporting evidence-based medicine across the healthcare enterprise and other customized software applications. Accordingly, like the case of MOCHA, it is possible to create analytics services and other applications on top of METEOR to meet user-specific needs. These details are illustrated in Figure 4 and Figure 5 below.

Model Development

The system development was launched at the Houston Methodist Hospital in 2012 through the integration of data from systems throughout the hospital into one comprehensive clinical data warehouse of patient information. By 2013, the integration of HIS and other health data was completed after the consolidation of vitals, medications, and orders. Pilot studies were then started and by 2015, an evaluation of the model found that it had significant benefits (Puppala et al. 2015).

Model Pilot Testing

Several pilot projects were conducted. For instance, one pilot project using web apps that allowed tracking of diabetes and transplant surgery was conducted in June, 2015. Other studies assessed which general (abdominal) surgery cases were associated with acute kidney injury. Another assessed which variables in patient care and health status contributed to the length of stay after transplant surgery. Furthermore, Stubbins et al. (2018) evaluated the integration of wearable devices and the mobile phone-enabled data collection into METEOR for personalized care of cancer patients. The study demonstrated real-time IoT-enabled data collection, which informed behaviour modification healthcare for individual patients.

Model Evaluation

The system was evaluated in 2015, according to a study by Puppala et al. (2015). Data and usability analysis performed as a preliminary evaluation indicated that METEOR addressed significant niches in the clinical informatics area and provided a powerful means for data integration and efficient access in supporting clinical and translational research. The evaluation concluded that the model improved outcomes, enabled coordinated care, and support health analytics and clinical research at the Houston Methodist Hospital.

Model Revision

The system has undergone constant development and improvements since its inception at the Houston Methodist Hospital. The platform has allowed for the development of user-specific applications and associated IoT-enabled data mining for personal healthcare. The case of MOCHA is an example where more contemporary gadgets such as wearable Fit bits and mobile scanners have been integrated into METEOR.

Model Implementation

The system has been operational and has supported various studies. Using the METEOR clinical data warehouse, and integrating with IoT-enabled data collection applications, the platform has facilitated various studies. In a study by Puppala et al. (2015), the platform enabled a retrospective analysis of fall patients' medical records. Also, using MOCHA, the METEOR was used for a personalized patient care study for cancer patients (see Figure 4). Patient data were collected with the help of wearable gadgets and mobile phone-enabled applications. The data was integrated with other patient data in the METEOR database to inform diagnosis.

Model Launch

The system has been operational since 2013 and continues to grow by the incorporation of data from ancillary systems such as radiology, cardiology, and operation room systems. Also, there have been developments of other applications like Nutritionix Track and integration with sensors and wearables such as Fitbit, which collect data in real-time to inform personal healthcare (see Figure 6).

Challenges

Even with available policies, guidelines, and regulations around data use in healthcare systems, privacy and security breach remain key issues for electronic healthcare systems. The security of patient data has been addressed through various innovations. In the case of MOCHA, for example, mobile applications do not store any patient data. All data is transferred to the METEOR database in real-time. However, concerns around data security have influenced considerable resistance to technologies.

System Vulnerability

The very connectedness of the IoT leaves it open to security and safety vulnerabilities. Every connected thing is susceptible to attack or misuse. For example, in September 2016 at DEF CON, one of the world's largest security conferences, 47 vulnerabilities affecting 23 IoT-enabled items (door locks, wheelchairs, thermostats and more) from 21 manufacturers were disclosed. Soon after, there was a massive Distributed Denial-of-Service (DDoS) attack on Oct. 21, 2016, against Dyn, an internet performance management company. The attack was accomplished when tens of millions of IoT-connected devices

like printers, DVRs, cable set-top boxes, webcams, and baby monitors were used to launch the DDoS and block Dyn's ability to connect internet users to the web addresses they hoped to access, such as Twitter, Amazon, PayPal, Spotify, Netflix, HBO, the *Wall Street Journal*, and the *New York Times* (Rainie & Anderson, 2017). A simple software program called Mirai was used to create the botnet that initiated the attack. After the Dyn attack, a report in the *New York Times* called the IoT a "weapon of mass disruption." While that assault amounted to nothing more than a short-lived slowdown of a large portion of the internet, it showed how vulnerable connected devices are to hacking and exploitation. In May 2017, a ransomware attack named WannaCry affected computers in 150 countries, and its creators demanded payments from those whose computers were compromised before releasing their files. Experts pointed out how dramatically this attack highlighted the vulnerabilities of the IoT.

Researchers have been showing how easy it is to hack cars, voting machines, and power plants. They have demonstrated ransomware exploits against home thermostats and exposed vulnerabilities in implanted heart pacemakers. In one paper, "IoT Goes Nuclear," analysts showed how a flaw in the design of smart lightbulbs could be used for a "bricking attack" that would effectively kill all of a city's traffic lights. Within the past year, Bryan Johnson (Kernal), Elon Musk (Neuralink), and Mark Zuckerberg (Facebook's Building 8) have announced initiatives to create an effective consumer-grade brain-computer interface; thus, of course, hacking a person's brain could also be a future security issue.

Result

The Internet of Things (IoT) has made digital personalized care more efficient. Sensors and wearable devices like inertial sensors, ECG (electrocardiogram), EEG (electroencephalogram), and GPS are able to observe and record multiple type health data, including heart rate, location, weight, blood pressure, and user-context information of an individual patient; thus, enabling physicians to monitor patients remotely (Krishna, Gurumoorthy and Obaidat 2018). IoT has provided a platform for people to seek personalized healthcare. For instance, through smart mobile apps, a physician can send the patient their medical information, and discuss the diagnosis and chart a way forward (Wason et al. 2020).

Prognoses

IoT will increasingly facilitate personalized healthcare; hence speeding up the delivery of services. As such, it will become increasingly easy for patients to access and track their health information and establish flawless communication with their providers (Ismail 2019). This will see IoT adoption in healthcare increase exponentially as more healthcare organizations adopt its use.

Further Research

The increase in the adoption of IoT as the delivery model changes to patient-centric care means that healthcare organizations will be handling big amounts of data. It will be vital to develop a sole source of truth to create meaningful and correct analytics (Vega-Barbas and Seoane 2019). Therefore, there is the need for further research on how healthcare organizations can better leverage the use of big data, while simultaneously optimizing staff productivity. One challenge among others is system vulnerability. Cybercrimes are increasingly becoming a major concern. More research is required to devise intervention strategies that mitigate system disruptions as a result of electronic hacking. There is also a need to improve access to personal data. Institutions need to be encouraged through evidence to implore policy makers to enact appropriate, effective, and inclusive policies.

Discussion

While the Covid-19 virus has presented unprecedented proportions of lifestyle challenges, it has also served as a catalyst in advancing medical technological achievements. As communities continue to anticipate more and more positive outcomes – vaccine arrivals, distribution and utilization of new vaccines and more – institutions have also continued to innovate. And remote medicine serves as one sector where significant positive outcomes are a new normal, a development that continues to escalate across different health-related themes. One silver lining of the morbidity and mortality of the pandemic is that it created an environment for lessons learned with a learning curve that affected every level of managing the different landscapes. It also gave institutions an opportunity to “pilot test” innovated telemedicine platforms: modelling, patient appointments, monitoring, and more, during the first wave. These improved and compelling innovations are currently being implemented during the second phase. And one can only speculate that more and more developments are yet to come.

The IoT technology has revolutionized the healthcare industry. Monitoring of patients through IoT devices gives the healthcare industry many advantages. For instance, physicians are now able to pick up health indicators that show patients are in danger and hence act on a timely basis to prevent the deterioration of health or even the loss of life (Bhatt, Dey, and Ashour 2017). Therefore, the relevance of digitized healthcare services cannot be adequately emphasized. The evidence-based advantages notwithstanding, suppliers (medical staff) and others need to be constantly reminded that while the system by itself is necessary, it is not sufficient. Effective application and utilization by the relevant users remain invaluable. As the saying goes, “garbage in garbage out (GIGO),” so an efficient system must start with appropriate, relevant, timely, valid, and cost-effective data correctly transformed, analyzed, and implemented.

Conclusion

IoT will continue to facilitate personalized healthcare and make the delivery of healthcare services increasingly more efficient with time. Healthcare organizations that want to focus more on patients are therefore advised to adopt IoT so as to benefit from it. IoT is the future. With access to more data and

appropriate technology, digitized personalized healthcare will not only be revolutionized, it will also introduce a more effective strategy of improved healthcare practices with patients serving as the driving force. And such dynamics will, without a doubt, contribute to rendering a win-win outcome for both the service provider and the patient.

Declarations

Ethical Approval and Consent to participate:

Granted

Availability of Data and material:

Granted if requested

Author's Contribution:

Research design, analyses, and research and manuscript

Acknowledgments:

NA

Funding:

Personal Funds

Competing Interest:

No competing interest

References

1. Alotaibi, Y. K., and Federico, F. 2017. "The impact of health information technology on patient safety." *Saudi Med Journal*, 173-1180.
2. Amendola, S., Lodato, R., Manzari, S., Occhiuzzi, C., and Marrocco, G. 2014. "RFID Technology for IoT-Based Personal Healthcare in Smart Spaces." *IEEE Internet of Things Journal*, 1(2), 144-152.
3. Bhatt, C., Dey, N., and Ashour, A. S. 2017. *Internet of Things and Big Data Technologies for Next Generation Healthcare*. Springer.

4. Chacko, A., and Hayajneh, T. 2018. "Security and Privacy Issues with IoT in Healthcare." *EAI Endorsed Transactions on Pervasive Health and Technology*, 4(14), 1-7.
5. Darshan, K. R., and Anandakumar, K. R. 2015. "A Comprehensive Review on Usage of Internet of Things (IoT) in Healthcare System." *015 International Conference on Emerging Research in Electronics, Computer Science and Technology (ICERECT)* (pp. 132-136). Mandya: IEEE.
6. Datta, S. K., Bonnet, C., Gyrard, A., Ferreira da Costa, R. P., and Boudaoud, K. 2015. "Applying Internet of Things for Personalized Healthcare in Smart Homes." *IEEE 2015 24th Wireless and Optical Communication Conference (WOCC) - Taipei, Taiwan (2015.10.23-2015.10.24)] 2015 24th Wireless and Optical Communication Conference (WOCC)*, (pp. 164-169).
7. Deng, Y.-Y., Chen, C.-L., Tsaur, W.-J., and Tang, Y.-W. 2017. "Internet of Things (IoT) Based Design of a Secure and Lightweight Body Area Network (BAN) Healthcare System." *Sensors*, 17(12), 2919.
8. Eken, C., and Eken, H. 2016. "Security Threats and Recommendation in IoT Healthcare." *Proceedings of the 9th EUROSIM & the 57th SIMS* (pp. 369-374). Oulu, Finland: EUROSIM.
9. Frantzidis, C., Bratsas, C., Klados, M., Konstantinidis, E., Lithari, C., Vivas, A., and Papadelis, C. 2010. "On the Classification of Emotional Biosignals Evoked While Viewing Affective Pictures: An Integrated Data-Mining-Based Approach for Healthcare Applications." *IEEE Transactions on Information Technology in Biomedicine*, 14(2), 309-317.
10. Gelogo, Y. E., Hwang, H. J., and Kim, H.-K. 2015. "Internet of Things (IoT) Framework for u-healthcare System." *International Journal of Smart Home*, 9(11), 323-330.
11. Google, 2020. Viewed July 2020.. <https://scholar.google.com/>.
12. He, L., Xue, M., and Gu, B. 2020. "Internet-of-things enabled supply chain planning and coordination with big data services: Certain theoretic implications." *Journal of Management Science and Engineering*, 5(1), 1-22.
13. Ismail, Y. 2019. *Internet of Things (IoT) for Automated and Smart Applications*. BoD – Books on Demand.
14. Jara, A. J., Zamora, M. A., and Skarmeta, A. F. 2011. "An Internet of Things–based Personal Device for Diabetes Therapy Management in Ambient Assisted Living (AAL)." *Personal and Ubiquitous Computing*, 15(4), 431–440.
15. Khan, S. F. 2017. "Health Care Monitoring System in Internet of Things (IoT) by Using RFID." *6th International Conference on Industrial Technology and Management (ICITM)* (pp. 198-204). Cambridge: IEEE.
16. Kimsey, T., Jeffords, J., Moghaddam, Y., and Rucinski, A. 2015. "An IoT Based Service System as a Research and Educational Platform." In D. Barbucha, N. N., & B. J., *New Trends in Intelligent Information and Database Systems. Studies in Computational Intelligence*. Cham: Springer.
17. Krishna, P. V., Gurumoorthy, S., and Obaidat, M. S. 2018. *Internet of Things and Personalized Health Care Systems*. Springer.
18. Kulkarni, A., and Sathe, S. 2014. "Healthcare Applications of the Internet of Things: A Review." *International Journal of Computer Science and Information Technologies*, 5(5), 6229-6232.

19. López, S., Ranasinghe, D., Harrison, M., and Mcfarlane, D. 2020. Adding Sense to the Internet of Things. *Personal and Ubiquitous Computing*, 16(3), 291-308.
20. OECD. <https://www.oecd.org/health/digital-health.htm>. Accessed November 2020.
21. Puppala, M., He, T., Chen, S., Ogunti, R., Yu, X., Li, F., Wong, S. T. 2015. "METEOR: An Enterprise Health Informatics Environment to Support Evidence-based Medicine." *IEEE transactions on bio-medical engineering*, 62(12), 2776-2786.
22. Rahmani, A., Thanigaivelan, N. K., Gia, T. N., Granados, J., Negash, B., and Liljeberg, P. 2015. "Smart e-Health Gateway: Bringing Intelligence to Internet-of-Things based ubiquitous healthcare systems." *2015 12th Annual IEEE Consumer Communications and Networking Conference (CCNC)*. Las Vegas, NV.
23. Rainie, L. & Anderson, J. 2017. "The Internet of Things Connectivity Binge: What are the Implications?" Available at: <http://www.pewinternet.org/2017/06/06/theinternet-of-things-connectivity-binge-what-are-theimplications/>
24. Serna, A., Pigot, H., and Rialle, V. 2007. "Modeling the Progression of Alzheimer's disease for Cognitive Assistance in Smart Homes." *User Modeling and User-Adapted Interaction*, 17, 415–438.
25. Simmhan, Y., Ravindra, P., Chaturvedi, S., Hegde, M., and Ballamajalu, R. 2017. "Towards a Data-Driven IoT Software Architecture for Smart City Utilities." *Software: Practice and Experience*, 1-27.
26. Storey, J. 2013. "Factors Affecting the Adoption of Quality Assurance Technologies in Healthcare." *Journal of Health Organisation and Management*, 498 - 519.
27. Stubbins, R., He, T., Yu, X., Puppala, M., Ezeana, C. F., Chen, S., Patel, T. 2018. "A Behavior-Modification, Clinical-Grade Mobile Application to Improve Breast Cancer Survivors' Accountability and Health Outcomes." *American Society of Clinical Oncology*, 1-10
28. Tundis, A., Kaleem, H., and Mühlhäuser, M. 2020. "Detecting and Tracking Criminals in the Real World through an IoT-Based System." *Sensors*, 20(13).
29. Vega-Barbas, M., & Seoane, F. 2019. *Transforming Ergonomics with Personalized Health and Intelligent Workplaces*. IOS Press.
30. Wason, R., Goyal, D., Jain, V., Balamurugan, S., and Baliyan, A. 2020. *Applications of Deep Learning and Big IoT on Personalized Health Care Services*. IGI Global.

Figures

Figure 1

Internet of Things (IoT) Ecosystems Source: (Author)

Figure 2

2016-2019 Averages of the 18 highest number of IoT Publication Themes Source: Google Scholar and Author

Figure 3

a: 2016 – 2019 Median Distribution of number of IoT-related Publications Source: Google Scholar and Author
b: Select IoT Inter-Annual Thematic Trends from 2016-2019 Source: Google Scholar and Author

Figure 4

METEOR Illustration

Figure 5

Architectural design for Methodist Hospital Cancer Health Application (MOCHA) and its integration in the METEOR model Source: Puppala et al. 2015

(a)

(b)

(c)

Figure 6

Illustration of METEOR-Based MOCHA Application studies (a) the frequency of use (b) a case of Personal Healthcare based on real-time patient data integrated into METEOR through a network and enabled by IoT (c) Study results indicating the frequency of use of MOCHA against weight loss for various participants. Source: Stubbins et al. (2018).