

5G SMART DIABETES TOWARDS PERSONALIZED DIABETES DIAGNOSIS WITH HEALTH CARE BIGDATA CLOUD

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Abstract - Recent advances in wireless networking and big data technologies, such as 5G networks, medical big data analytics, and the Internet of Things, along with recent developments in wearable computing and artificial intelligence, are enabling the development and implementation of innovative diabetes monitoring systems and applications. Due to the life-long and systematic harm suffered by diabetes patients, it is critical to design effective methods for the diagnosis and treatment of diabetes. Based on our comprehensive investigation, this article classifies those methods into Diabetes 1.0 and Diabetes 2.0, which exhibit deficiencies in terms of networking and intelligence.

Thus, our goal is to design a sustainable, cost-effective, and intelligent diabetes diagnosis solution with personalized treatment. In this article, we first propose the 5G-Smart Diabetes system, which combines the state-of-the-art technologies such as wearable 2.0, machine learning, and big data to generate comprehensive sensing and analysis for patients suffering from diabetes. Then we present the data sharing mechanism and personalized data analysis model for 5G-Smart Diabetes. Finally, we build a 5G-Smart Diabetes testbed that includes smart clothing, smartphone, and big data clouds. The experimental results show that our system can effectively provide personalized diagnosis and treatment suggestions to patients.

Key Words: Monitoring, Diabetes, Wearable, Big Data, Smart Systems, Machine Learning

1.INTRODUCTION

Diabetes is an extremely common chronic disease from which nearly 8.5 percent of the world population suffer; 422 million people worldwide have to struggle with diabetes. It is crucial to note that type 2 diabetes mellitus makes up about 90 percent of the cases [1]. More critically, the situation will be worse, as reported in [2], with more teenagers and youth becoming susceptible to diabetes as well. Due to the fact that diabetes has a huge impact on global well-being and economy, it is urgent to improve methods for the prevention and treatment of diabetes [3].

Furthermore, various factors can cause the disease, such as improper and unhealthy lifestyle, vulnerable emotional status, along with the accumulated stress from society and work. However, the existing diabetes detection system faces the following problems:

The system is uncomfortable, and real-time data collection is difficult. Furthermore, it lacks continuous monitoring of multi-

dimensional physiological indicators of patients suffering from diabetes [4, 5].

The diabetes detection model lacks a data sharing mechanism and personalized analysis of big data from different sources including lifestyle, sports, diet, and so on [6, 7].

There are no continuous suggestions for the prevention and treatment of diabetes and corresponding supervision strategies [8, 9].

To solve the above problems, in this article, we first propose a next generation diabetes solution called the 5G-Smart Diabetes system, which integrates novel technologies including fifth generation (5G) mobile networks, machine learning, medical big data, social networking, smart clothing [10], and so on. Then we present the data sharing mechanism and personalized data analysis model for 5G-Smart Diabetes. Finally, based on the smart clothing, smartphone, and big data healthcare clouds, we build a 5G-Smart Diabetes testbed and give the experiment results.

Furthermore, the “5G” in 5G-Smart Diabetes has a two-fold meaning. On one hand, it refers to the 5G technology that will be adopted as the communication infrastructure to realize high-quality and continuous monitoring of the physiological states of patients with diabetes and to provide treatment services for such patients without restraining their freedom. On the other hand, “5G” refers to the following “5 goals”: cost effectiveness, comfortability, personalization, sustainability, and smartness.

Cost Effectiveness: It is achieved from two aspects. First, 5G-Smart Diabetes keeps users in a healthy lifestyle so as to prevent users from getting the disease in the early stage. The reduction of disease risk would lead to decreasing the cost of diabetes treatment. Second, 5G-Smart Diabetes facilitates out-of-hospital treatment, thus reducing the cost compared to on-the-spot treatment, especially long-term hospitalization of the patient.

Comfortability: To achieve comfort for patients, it is required that 5G-Smart Diabetes does not disturb the patients’ daily activities as much as possible. Thus, 5G-Smart Diabetes integrates smart clothing [3], mobile phones, and portable blood glucose monitoring devices to easily monitor patients’ blood glucose and other physiological indicators.

Personalization: 5G-Smart Diabetes utilizes various machine learning and cognitive computing algorithms to establish personalized diabetes diagnosis for the prevention and treatment of diabetes. Based on the collected blood glucose data and individualized physiological indicators, 5G-Smart Diabetes produces personalized treatment solutions for patients.

Sustainability: By continuously collecting, storing, and analyzing information on personal diabetes, 5G-Smart Diabetes adjusts the treatment strategy in time based on the changes of patients' status. Furthermore, in order to be sustainable for data-driven diabetes diagnosis and treatment, 5G-Smart Diabetes establishes effective information sharing among patients, relatives, friends, personal health advisors, and doctors.

With the help of social networking, the patient's mood can be improved so that he or she is more self-motivated to perform a treatment plan in time. **Smartness:** With cognitive intelligence toward patients' status and network resources, 5G-Smart Diabetes achieves early detection and prevention of diabetes and provides personalized treatment to patients. The remaining part of the article is organized as follows. We first present the system architecture of 5G-Smart Diabetes. Then we explain the data sharing mechanism and propose the personalized data analysis model. Furthermore, we introduce the 5G-Smart Diabetes testbed. Finally, the conclusion of this article is given. Because of the way that diabetes has an enormous effect on worldwide prosperity and economy, further developing strategies for the prevention is dire also, treatment of diabetes. Moreover, different elements can cause the sickness, like ill-advised and unfortunate way of life, weak feeling status, alongside the gathered pressure from society and work. However, the 5G-Smart Diabetes system, which incorporates cutting-edge technologies such as machine learning, medical big data, social networking, smart clothing, and so on, is our first proposal for a diabetes solution of the next generation [10]. Then, at that point, we present the information sharing system and customized information investigation model for 5G-Shrewd Diabetes. At long last, based on the savvy clothing, cell phone, and enormous information medical services mists, we construct a 5G-Shrewd Diabetes testbed and give the investigation results.

Besides, the "5G" in 5G-Shrewd Diabetes has a two-crease meaning. On one hand, it alludes to the 5G innovation that will be embraced as the correspondence framework to understand top notch and consistent observing of the physiological conditions of patients with diabetes and to give treatment administrations to such patients without limiting their opportunity. On the other hand, "5G" alludes to the accompanying "5 objectives": cost-effectiveness, ease of use, individualization, sustainability, and intelligence.

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their opportunity. On the other hand, "5G" alludes to the accompanying "5 objectives": cost-effectiveness, ease of use, individualization, sustainability, and intelligence.

2. LITERATURE SURVEY

Chen et al. proposed the 5G-Smart Diabetes system, which combined the state-of-the-art technologies such as wearable 2.0, machine learning, and big data to generate comprehensive sensing and analysis for patients suffering from diabetes. Then this work presented the data sharing mechanism and personalized data analysis model for 5G-Smart Diabetes. Finally, this work builds a 5G-Smart Diabetes testbed that includes smart clothing, smartphone, and big data clouds. The experimental results showed that the system can effectively provide personalized diagnosis and treatment suggestions to patients.

Rghioui et al. presented an intelligent architecture for monitoring diabetic patients by using machine learning algorithms. The architecture elements included smart devices, sensors, and smartphones to collect measurements from the body. The intelligent system collected the data received from the patient and performed data classification using machine learning to make a diagnosis. The proposed prediction system was evaluated by several machine learning algorithms, and the simulation results demonstrated that the sequential minimal optimization (SMO) algorithm gives superior classification accuracy, sensitivity, and precision compared to other algorithms. Turkish Journal of Computer and Mathematics Education Vol.14 No.02 (2023),99- 106 101 Research Article

Venkatachalam et al. motivated to develop a diabetes motoring system for patients using IoT device in their body which monitors their blood sugar level, blood pressure, sport activities, diet plan, oxygen level, ECG data. The data are processed using feature selection algorithm called as particle swarm optimization and transmitted to nearest edge node for processing in 5G networks. Secondly, data are processed using DBN Layer. Thirdly, this work shared the diagnosed data output through the wireless communication such as LTE/5G to the patients connected through the edge nodes for further medical assistance. The patient wearable devices are connected to the social network. The Result of this proposed system is evaluated with some existing system. Time and Performance outperform than other techniques.

Prakash et al. introduced a neural network-based ensemble voting classifier to predict accurately the diabetes in the patients via online monitoring. The study consists of Internet of Things (IoT) devices to monitor the instances of the patients. While monitoring, the data are transferred from IoT devices to smartphones and then to the cloud, where the process of

classification takes place. The simulation is conducted on the collected samples using the python tool. The results of the simulation show that the proposed method achieves a higher accuracy rate, higher precision, recall, and f-measure than existing state-of-art ensemble models.

Tsoulchas et al. proposed a model to monitor the health of people with diabetes melitus, a disease with high incident rates mainly at the elderly but also in younger people. Specifically, a study about the existing medically approved technologies for continuous measurement of diabetes is described. Subsequently, the model for monitoring patient's blood glucose levels is described. Whenever a patient's blood glucose levels are Low or High, the model triggers an alarm to a Cloud infrastructure in order remote medical staff to provide immediate cure to the patient. Furthermore, to assure the immediate response of the remote medical staff, the proposed model is deployed upon a 5G wireless network architecture. Huang et al. proposed a 5G-based Artificial Intelligence Diabetes Management architecture (AIDM), which can help physicians and patients to manage both acute complications and chronic complications. The AIDM contains five layers: the sensing layer, the transmission layer, the storage layer, the computing layer, and the application layer. We build a test bed for the transmission and application layers. Specifically, this work applied a delay-aware RA optimization based on a doublequeue model to improve access efficiency in smart hospital wards in the transmission layer. In application layer, this work builds a prediction model using a deep forest algorithm.

3. ANALYSIS OF A SYSTEM

Analysis of the Existing System, helps in designing problem statement of Proposed system. In the following section based on the analysis of Existing System, the requirements of Proposed System has defined.

Existing System:

As there is no staff available in unmanned restaurants, it is difficult for the restaurant management to estimate how the concept and the food is experienced by the customers. Existing rating systems, such as Google and TripAdvisor, only partially solve this problem, as they only cover a part of the customer's opinions. These rating systems are only used by a subset of the customers who rate the restaurant on independent rating platforms on their own initiative. This applies mainly to customers who experience their visit as very positive or negative.

Proposed System:

In order to solve the above problem, all customers must be motivated to give a rating. This paper introduces an approach for a restaurant rating system that asks every customer for a rating after their visit to increase the number of ratings as much as possible. This system can be used unmanned restaurants; the scoring system is based on facial expression detection using pertained convolutional neural network (CNN) models. It

allows the customer to rate the food by taking or capturing a picture of his face that reflects the corresponding feelings. Compared to text-based rating system, there is much less information and no individual experience reports collected. However, this simple fast and playful rating system should give a wider range of opinions about the experiences of the customers with the restaurant concept. In proposed work, we are using Decision Tree, SVM, Artificial Neural Network algorithms from python to predict patient condition from his data. To train these algorithms we are using diabetes dataset. To predict data efficiently author is using Ensemble Algorithm which is combination of Decision Tree, SVM and ANN algorithm. Training model of all these three algorithms will be merging inside Ensemble Algorithm to get better accuracy and results.

Personalization: In this technique one patient can share his data with other patient based on distance between cloud servers they are using to store data. Here we are using dataset so sharing is not possible but I am making all predicted test data values to be open so all users can see or share it.

Smartness: this technique will be considered as smart as it required no human effort to inform patient about current condition. Here we have designed two applications to implement the above technique.

1) Cloud Application: This application act like a cloud server and storage and train dataset model with various algorithms such as decision tree, SVM and ANN and Ensemble algorithms.

2) User Application: In this application we will upload some test data and will be consider as user sense data and this data will be sent to cloud server and cloud server will apply decision and SVM and ANN model on test data to predict patient condition and send resultant data to this application. As we don't have sensors to sense data, so we consider uploaded test data as sense data. Here we don't have user details to share data so i am keeping all predicted data to be open so all users can see and share.

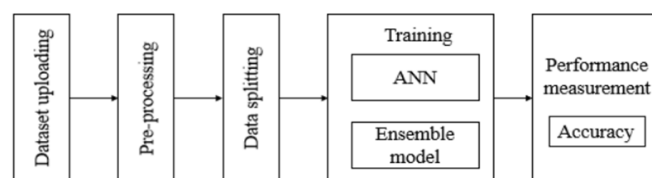


Fig. 1: Block diagram of cloud application.

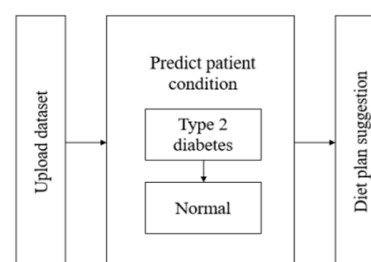


Fig. 2: Block diagram of user application.

4. METHODOLOGY

The input design is the link between the information system and the user. It comprises the developing specification and procedures for data preparation and those steps are necessary to put transaction data in to a usable form for processing can be achieved by inspecting the computer to read data from a written or printed document or it can occur by having people keying the data directly into the system. The design of input focuses on controlling the amount of input required, controlling the errors, avoiding delay, avoiding extra steps and keeping the process simple. The input is designed in such a way so that it provides security and ease of use with retaining the privacy. Input Design considered the following things:

Objectives:

1.Input Design is the process of converting a user-oriented description of the input into a computer- based system. This design is important to avoid errors in the data input process and show the correct direction to the management for getting correct information from the computerized system.

2.It is achieved by creating user-friendly screens for the data entry to handle large volume of data. The goal of designing input is to make data entry easier and to be free from errors. The data entry screen is designed in such a way that all the data manipulates can be performed. It also provides record viewing facilities.

3.When the data is entered it will check for its validity. Data can be entered with the help of screens. Appropriate messages are provided as when needed so that the user will not be in maize of instant. Thus, the objective of input design is to create an input layout that is easy to follow. The input design is the link between the information system and the user. It comprises the developing specification and procedures for data preparation and those steps are necessary to put transaction data in to a usable form for processing can be achieved by inspecting the computer to read data from a written or printed document or it can occur by having people keying the data directly into the system. The design of input focuses on controlling the amount of input required, controlling the errors, avoiding delay, avoiding extra steps and keeping the process simple. The input is designed in such a way so that it provides security and ease of use with retaining privacy. Input Design considered the following things:

Methods for preparing input validations and steps to follow when error occur.

4.Input Design is the process of converting a user-oriented description of the input into a computer- based system. This design is important to avoid errors in the data input process and show the correct direction to the management for getting correct information from the computerized system.

5.It is achieved by creating user-friendly screens for the data entry to handle large volume of data. The goal of designing input is to make data entry easier and to be free from errors. The data entry screen is designed in such a way that all the data manipulates can be performed. It also provides record viewing facilities.

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OUTPUT DATA:

A quality output is one, which meets the requirements of the end user and presents the information clearly. In any system results of processing are communicated to the users and to other system through outputs. In output design it is determined how the information is to be displaced for immediate need and also the hard copy output. It is the most important and direct source of information to the user. Efficient and intelligent output design improves the system's relationship to help user decision- making.

1.Designing computer output should proceed in an organized, well thought out manner; the right output must be developed while ensuring that each output element is designed so that people will find the system can use easily and effectively. When analysis design computer output, they should Identify the specific output that is needed to meet the requirements.

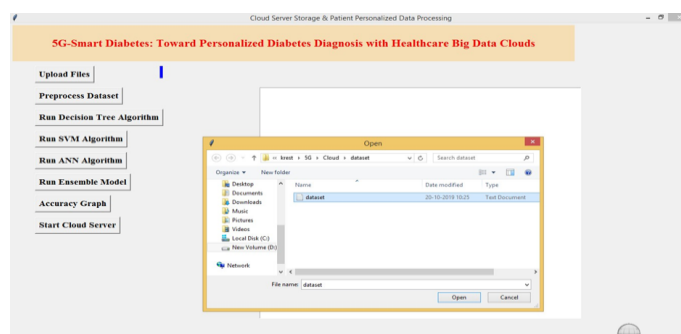
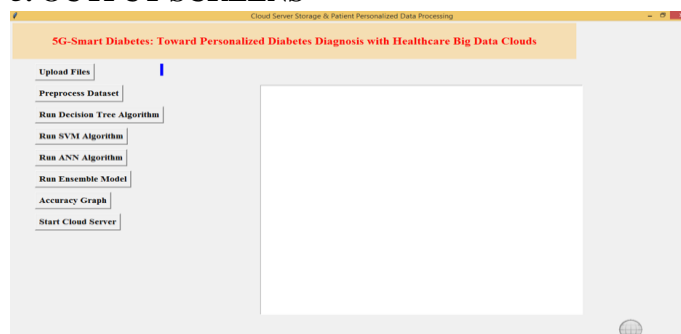
2.Select methods for presenting information.

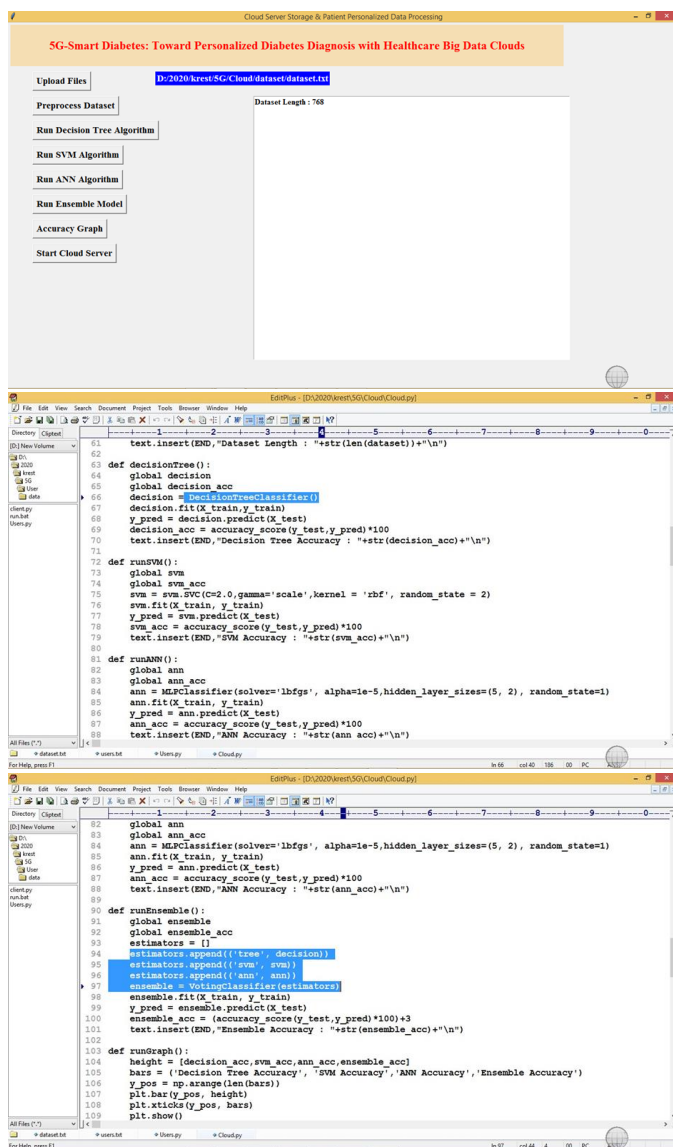
3.Create document, report, or other formats that contain information produced by the system.

The output form of an information system should accomplish one or more of the following objectives.

Convey information about past activities, current status or projections of the Future. Signal important events, opportunities, problems, or warnings. Trigger an action. Confirm an action.

5. OUTPUT SCREENS





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1: Cloud Server Storage & Patient Personalized Data Processing
2: 5G-Smart Diabetes: Toward Personalized Diabetes Diagnosis with Healthcare Big Data Clouds
3:
4: Upload Files
5: Preprocess Dataset
6: Run Decision Tree Algorithm
7: Run SVM Algorithm
8: Run ANN Algorithm
9: Run Ensemble Model
10: Accuracy Graph
11: Start Cloud Server
12:
13: Dataset Length: 768
14:
15: text.insert(END, "Dataset Length : "+str(len(dataset))+"\n")
16:
17: def decisionTree():
18:     global decision
19:     global decision_acc
20:     decision = DecisionTreeClassifier()
21:     decision.fit(X_train, y_train)
22:     y_pred = decision.predict(X_test)
23:     decision_acc = accuracy_score(y_test, y_pred)*100
24:     text.insert(END, "Decision Tree Accuracy : "+str(decision_acc)+"\n")
25:
26: def runSVM():
27:     global svm
28:     global svm_acc
29:     svm = svm.SVC(C=0.0, gamma='scale', kernel='rbf', random_state=2)
30:     svm.fit(X_train, y_train)
31:     y_pred = svm.predict(X_test)
32:     svm_acc = accuracy_score(y_test, y_pred)*100
33:     text.insert(END, "SVM Accuracy : "+str(svm_acc)+"\n")
34:
35: def runANN():
36:     global ann
37:     global ann_acc
38:     ann = MLPClassifier(solver='lbfgs', alpha=1e-5, hidden_layer_sizes=(5, 2), random_state=1)
39:     ann.fit(X_train, y_train)
40:     y_pred = ann.predict(X_test)
41:     ann_acc = accuracy_score(y_test, y_pred)*100
42:     text.insert(END, "ANN Accuracy : "+str(ann_acc)+"\n")
43:
44: def runEnsemble():
45:     global ensemble
46:     global ensemble_acc
47:     estimators = []
48:     estimators.append((tree, decision))
49:     estimators.append((svm, svm))
50:     estimators.append((ann, ann))
51:     ensemble = VotingClassifier(estimators)
52:     ensemble.fit(X_train, y_train)
53:     y_pred = ensemble.predict(X_test)
54:     ensemble_acc = (accuracy_score(y_test, y_pred)*100)+3
55:     text.insert(END, "Ensemble Accuracy : "+str(ensemble_acc)+"\n")
56:
57: def runGraph():
58:     height = [decision_acc, svm_acc, ann_acc, ensemble_acc]
59:     bars = ('Decision Tree Accuracy', 'SVM Accuracy', 'ANN Accuracy', 'Ensemble Accuracy')
60:     y_pos = np.arange(len(bars))
61:     plt.bar(y_pos, height)
62:     plt.title('Accuracy Comparison')
63:     plt.show()
64:
65: # Run the algorithms
66: decisionTree()
67: runSVM()
68: runANN()
69: runEnsemble()
70: runGraph()

```

6. CONCLUSION

In this, we first propose a 5G-Smart Diabetes system that includes a sensing layer, a personalized diagnosis layer, and a data sharing layer. Compared to Diabetes 1.0 and Diabetes 2.0, this system can achieve sustainable, cost-effective, and intelligence diabetes diagnosis. Then we propose a highly cost-efficient data sharing mechanism in social space and data space. In addition, using machine learning methods, we present a personalized data analysis model for 5G-Smart Diabetes. Finally, based on the smart clothing, smartphone and data center, we build a 5G-Smart Diabetes testbed. The experimental results show that our system can provide personalized diagnosis and treatment suggestions to patients. Significant security threats and concerns faced by 5G-enabled smart healthcare systems. It is necessary to reduce these issues and threats to understand the security needs of such systems. Because smart healthcare devices have limited scalability, resource constraints, single-point-of-failure, high cost, and standard security techniques are unable to meet all of the security requirements of 5G-enabled smart healthcare. Many technologies such as Blockchain and Artificial intelligence have recently brought a new healthcare security and privacy era. This paper presents 5G smart healthcare architecture and key enabling technologies. This study looked at various

technological features and services related to 5G smart healthcare security, including authentication, confidentiality availability, non-repudiation and integrity. We also discussed many security threats in 5G smart healthcare connectivity poses and available solutions. Finally, open issues and future research directions are presented for young researchers.

7. REFERENCES

1. S. Mendis, "Global Status Report on Noncommunicable Diseases 2014," WHO, tech. rep.; <http://www.who.int/nmh/publications/ncd-status-report-2014/en/>, accessed Jan. 2015.
2. F. Florencia et al., IDF Diabetes Atlas, 6th ed., Int'l. Diabetes Federation, tech. rep.; <http://www.diabetesatlas.org/>, accessed Jan. 2016.
3. M. Chen et al., "Disease Prediction by Machine Learning over Big Healthcare Data," IEEE Access, vol. 5, June 2017, pp. 8869--79.
4. O. Geman, I. Chiuchisan, and R. Todorean, "Application of Adaptive Neuro-Fuzzy Inference System for Diabetes Classification and prediction," Proc. 6th IEEE Int'l. Conf. E-Health and Bioengineering, Sinaia, Romania, July 2017, pp. 639--642.
5. S. Fong, et al. "Real-Time Decision Rules for Diabetes Therapy Management by Data Stream Mining," IT Professional, vol. 26, no. 99, June 2017, pp. 1--8.
6. B. Lee, J. Kim, "Identification of Type 2 Diabetes Risk Factors Using Phenotypes Consisting of Anthropometry and Triglycerides Based on Machine Learning," IEEE J. Biomed. Health Info., vol. 20, no. 1, Jan. 2016, pp. 39--46.
7. M. Hossain, et al., "Big Data-Driven Service Composition Using Parallel Clustered Particle Swarm Optimization in Mobile Environment," IEEE Trans. Serv. Comp., vol. 9, no. 5, Aug. 2016, pp. 806--17.
8. M. Hossain, "Cloud-Supported Cyber-Physical Localization Framework for Patients Monitoring," IEEE Sys. J., vol. 11, no. 1, Sept. 2017, pp. 118--27.
9. P. Pesl, et al., "An Advanced Bolus Calculator for Type 1 Diabetes: System Architecture and Usability Results," IEEE J. Biomed. Health Info., vol. 20, no. 1, Jan. 2016, pp. 11--17.
10. M. Chen et al., "Wearable 2.0: Enable Human-Cloud Integration in Next Generation Healthcare System," IEEE Commun. Mag., vol. 55, no. 1, Jan. 2017, pp. 54--61.
11. E. Marie et al., "Diabetes 2.0: Next-Generation Approach to Diagnosis and Treatment," Brigham Health Hub, tech. rep.; <https://brighamhealthhub.org/diabetes-2-0-next-generation-approach-to-diagnosis-and-treatment>, 2017, accessed Feb. 2017.
12. M. Chen et al., "Green and Mobility-Aware Caching in 5G Networks," IEEE Trans. Wireless Commun., vol. 16, no. 12, 2017, pp. 8347--61.
13. C. Yao et al., "A Convolutional Neural Network Model for Online Medical Guidance," IEEE Access, vol. 4, Aug. 2016, pp. 4094--4103.