eHealth and Smart Solutions framework for health monitoring in the course of the pandemic

János Dávid Balogh\textsuperscript{a}, Attila Adamkó\textsuperscript{b}

\textsuperscript{a}Doctoral School of Informatics University of Debrecen, Debrecen, Hungary
balogh.janos.david.official@gmail.com

\textsuperscript{b}Department of Information Technology University of Debrecen, Debrecen, Hungary
adamko.attila@inf.unideb.hu

Abstract. Our main objective focuses on the different design principles for applications that can receive and store various data from smart devices, e.g., smart bands, smart watches, and smart homes, and could gain more information about its users to some extent to their health. We would like to collect this information, store it, then create understandable diagrams and show them to the user in a modern, responsive environment. We placed great emphasis on the underlying architecture, which holds the necessary features (scalability and extensibility). We have followed the design process’s standards, recommendations, and protocols. The result is an application framework that fulfills the duty of an XXI. century smart solutions application that could monitor people’s health and even help them live healthier lives by avoiding chronic disorders, e.g., obesity and/or diabetes. We focused on predicting and classifying COVID-19 disease according to the collected information and research.

Keywords: smart solutions, smart devices, healthcare, architecture, data gathering, data analysis, machine learning, covid, data model

AMS Subject Classification: AMS Subject Classifications

1. Data collection

Problem 1.1 (Data collection). The first problem is finding a solution to gather health data from various IoT devices and send them to the main application. Every device has its own structure, so we had to find a way to handle that.
1.1. Smartwatch with SDK

Many smartwatches have more precise sensors to measure heart rate, distance, stress level, and much more health information. These watches come out with operating systems from the factory, and the manufacturers provide SDK. We could develop an application that collects the above-mentioned health information over time into one file and send it to our architecture via a REST API endpoint.

Also, these devices have default smartphone applications where we can export data easily without our own application, and the users can share manually with us.

**Example 1.2.** The Figure 1 shows the data we collected from our test device, a Samsung Galaxy Watch 2. This spreadsheet focuses on heartbeat rate and displays minimum and maximum heartbeat and the current measurements. Besides the measurements, it informs us of the measurement date, device, and more useful health information.

![Figure 1. Health information from smartwatch.](image)

1.2. Smart bands

We talked about the smartwatch solution, but these devices still are more expensive than an average person could afford. That is why we shifted our focus to smart bands, which are more affordable for larger crowds.

These devices are also equipped with sensors, maybe not the same precision level as the smartwatch sensors, but they still could get information about the health information we want to have.

The greatest disadvantage of smart bands is the lack of an open operating system and SDK. We had to find a solution to collect the data from the devices. Modern Xiaomi and Huawei devices offer a REST API endpoint where we authorize ourselves and get the requested information. We used that endpoint in an application where the user just adds their API key and authorization information so we could get our hands on the data and send it to our architecture.
Example 1.3. As we can see in the Figure 2, the smart bands send less information about the heartbeat rate compared to the smartwatch, but we still get what is necessary, the date and the measurements, which are crucial to our research.

<table>
<thead>
<tr>
<th>date</th>
<th>time</th>
<th>heartRate</th>
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<tbody>
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<td>91</td>
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</tr>
<tr>
<td>2023.03.14</td>
<td>21:16:00</td>
<td>75</td>
</tr>
</tbody>
</table>

Figure 2. Health information from smartband.

1.3. Renaming

As we work with many types of devices, we are bumping into one great obstacle. Different devices have various data storing structures containing the same type of data, but the structure is different. If we would like to work with them effectively, we should get them brought into a common structure. The above-mentioned figures show what the problem is, the smartwatch heartbeat rate measurement has a long specific name, but the smart band heartbeat rate is a simple one. We dedicated a service which is between the receiving endpoint and the storing to handle these various formats and produce them into our structure, but we do not lose important information.

2. Database and storing

2.1. Database

We were taking account of the difference of smart devices and sensors, they gather and deliver more or less the same data structure, but we wanted to make sure that the varying data structure will not cause problems in storing these data. That is the reason why we decided to use NoSQL database instead of the standard regular SQL databases. The choice of ours was the MongoDB. It offers a flexible document data model database which stores data in JSON format. Ad-hoc queries and secondary indexing are supported which makes truly powerful ways to access our data. The
database does not contain any business logic, on how to process the data. It has only one job and it fulfills that perfectly. In our application we implemented a service which handles the repository. The repository has the connection with the database and is responsible for the CRUD operations and various queries. This book [4] guided us to implement a data model which will contain our sensitive data and is memory effective.

2.2. Sorting out false measurements

When we are wearing the smart devices during our life the devices automatically measure our health and sometimes they could get a false information when the sensors are not touching well the user or any kind of problem. If this happens they could record abnormally high heartbeat rate or zero value and we cannot work with these false information, we need to find a way to get rid of these data.

We used the Simple Moving Average (SMA) to find out false measurements. SMA attaches a value to every i-th element using the average of sum of all of i elements. When this value is too high or low we label it as false measurements and do not work with anymore.

Example 2.1. As the Figure 3 shows it means a lot when we are using SMA. The red line represents the original measurements with false values and it clearly points out what happens when it contains false information. The big spike is when the heartbeat rate is too high and the downhill when it failed to measure any heartbeat. The blue line is after SMA and it is a more consistent representation of the measurements so we could work with valid and supposedly true values after storing in the database.

Figure 3. Diagram showing the sorting out.

3. Architecture

3.1. Goals

As we stated before one of our goals is a flexible data model which could be a base of a general data managing platform. For this purpose, we relied on NoSQL database MongoDB that is a document data model storing data in BSON objects. A BSON object is a binary JSON object, this format helps in data storage and
network transfer. We could store sensor data with different structure easily and even process them. In the backend we used the bson package and the Document data type to work with the database.

Our other goal was to create an architecture which satisfies our needs. We wanted to design a layered architecture that is scalable, modular, expandable.

Scalable, we would like to work with larger data inputs and more users so we must be ready to receive more requests. To do that we must scale our architecture without a problem, or we would suffer heavy data loss.

Modular, anytime we can decide which module is needed or not. We could wire or unwire any module with ease.

Expandable, if any new requirements come in the picture, we could meet them by adding new features and modules to the architecture.

For fulfilling these properties, we decided to implement a microservice architecture. In our project we expanded it to see how it will work out. We studied [1] to get more understand in modern software architecture engineering and [6] to learn developing in Spring Boot which will support our goals.

### 3.2. Microservices

The microservices architecture is a collection of services that are highly maintainable and testable and independently deployable. Spring offers an opportunity if we would like to wire many services together. All our services share the same database, but they have different endpoints with REST API interface and business logic.

We chose this architecture because it is easy to expand later when we are adding more and more services to the application.

![Figure 4. The architecture.](image-url)
As the Figure 4 shows, we have two external entry point: the smart watch and a browser. The smart watch sends the data through the REST API interface and its service to the database.

The browser can send data as exported file or/and manually. We provided different services to them, and they handle the rest to the database. We query the database in the creating diagrams, tools and machine learning services and their results are shown in the browser.

3.3. Creating graphs

After we got data from the users and they are stored in our database, finally we could work with them. First, we need to check if they are normalized. It means none of them has a different format. If they do, we have to normalize them. The main goal of the application is to provide value added services, give feedback and motivation about the lifestyle change. For this reason, our illustrations are easily readable. To create diagrams, we used the R script language. R is famous for its data processing and plotting abilities, it is widely used in analytic, statistic works, and Spring can invoke third party programs. Our script needs a dataset to work with it. For this purpose, we wrote a query which collects the user’s relevant data from the database. The script waits for a csv file, so a simple CSV writer class writes the result into a csv file. The newly created csv file is passed into the script as argument and runs. Creates diagrams such as calorie burn by date, heart rate rating, graph of steps, calorie intake and burn by date and many more. These diagrams are exported as images, and they are shown to the user by the frontend part of the application.

3.4. Machine learning

The greatest feature should be our machine learning classification which will be using our collected data to classify the user or the patient if they are infected with any disease. Our personal approach was from our experience with the COVID-19 which caused lots of loss and trouble in our world. [5] helped our research to get started to develop a model in machine learning in Python. The architecture will pass the input values to the script and call it and will be waiting for it’s output. We used the most popular algorithms to train our model and test it with the data that we had and it looks like it was satisfying. According to this research [2] we saw that if a person with COVID-19 produces more heartbeat even if the person is sitting or in resting position. Because we collected heartbeat rate and more health information we could classify if the user is dealing with COVID-19. This works as a binary classification where we import the user’s data which contains heartbeat rate and activity position and we calculate average heartbeat by day when they are resting or sitting. The average heartbeats by days will be used as input data and if they show increasing values day by day and they are out of normal person heartbeat rate in resting position interval we could suppose they are infected with...
the disease, we label them with true value. If the average values within the interval and shows no increasing values they are healthy and we label them with false value.

In the future we would like to work with infected people to wear smart devices to collect more precise information about their heartbeat and symptoms to improve our machine learning model to predict their condition.

3.5. HL7

Because our project works with healthcare, we found it worthwhile to use the Health Level Seven standards. The Health Level 7 standards are a set of international standards for transfer of clinical and administrative data between software application used by various healthcare providers.

The name comes from using the application layer which is layer 7 in the OSI model. The HL7 Standards are produced by Health Level Seven international and used by bodies e.g., American National Standards institute and International Organization for Standards. [3] showed a way to implement and use the HL7 standards to achieve communication between our architecture and healthcare providers.

We dedicated for a HL7 a new service which satisfies the standards for communication between healthcare software applications. It is heavily relied on XML documents because HL7 newer versions store the necessary data in it. It includes the patient’s ID and status like body weight or other measurements such as heart rate.

We want to develop a service in our general architecture where medical approaches could benefit from it, helping the digitalization of healthcare. We are still working on it to be more precise with the data output from the smart watches, but we have already found a way to process data from them and convert them into XML to use HL7 Standards.
We are trying to build a relationship with the Clinic of the University of Debrecen to see how will work our solution in live environment where doctors take part in our research. They will get 24/7 healthcare information about the patient and if any trouble could occur they could take steps to prevent disease like the COVID or give aid if necessary.

4. Conclusion

While developing this application we gained more knowledge and insight into modern application architecture, responsible websites and even healthcare a bit. We would like to invest more in this project because we are dedicated in our goals to improve people’s health using the instruments of technology.

We are planning to invite more people to the project using different smart devices to collect more data. If we have more data, we can continue the machine learning service because it requires more precise information about diet and service and their results.

References


