Artificial intelligence model for continuous, in-home, posture and health monitoring including user feedback and predictions of clinical assessment.

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## Introduction

According to [1] in 2020 alone, musculoskeletal disorders (MSDs) had been ranked 2nd as the leading non-fatal disability which has been affecting more than a billion people worldwide. In Finland, MSD had taken the spotlight as being the leading cause of temporal disability within the nation, through which a lot of resources allocated towards the health services [2]. It might be misconceived that only the elderly are the only ones that suffer from this condition. However, a report by [3] has concluded that quite a number of individuals across different age groups are currently suffering from it. It was reported that MSDs can often originate during the childhood stage mainly due to adoption of abnormal postures and low physical activities, which subsequently lead to long-term chronic pain, discomfort, and physical limitations. Traditional examination and treatment procedures most often consist of regular clinical visits and are currently viewed as being inconvenient and costly. According to [4], MSDs have said to have cost the European Union (EU) over 2% of its gross domestic product (GDP), which is estimated to be over €240bn each year. There is no doubt that this is a steadily growing concern that needs to be properly addressed.

Nowadays, most of the in-office work requires workers to be in a seated position for an extended period, which is said to have adverse effects to one’s health. According to [5] and [6], prolong sitting is one of the leading causes of MSDs among office workers. It is therefore recommended that the users should go for small walk breaks after every few hours. The integration of smart sensing chairs that would be actively monitored and provide feedback on one’s health and activity levels would be deemed quite useful.

Furthermore, with the rapid advancement in data sensor technology and Artificial Intelligence in this present age, there should be new and commercialized solutions out there in the market for continuous posture and health monitoring. There is no doubt that these types of systems have the potential of contributing towards the idea of personalized healthcare and improving the quality of life, especially for individuals that are suffering from MSDs.

With that in mind, various research studies have investigated the development of posture monitoring systems which are aimed at detecting bad sitting postures to help the end user in maintaining the right sitting posture at every given time. These types of systems are named as “smart sensing chairs” which goes all the way back to a research study done by [7], which fitted a chair with a pressure distribution sensor in order to classify a user’s sitting postures which was just first of many. Furthermore, with a lot of research papers being published in this field, this literature review aims to evaluate related studies and identify research gaps that can pave the way for further investigation into this study. By exploring existing studies, it is possible to gain a better understanding of the current state on the implementation of a smart sensing chair for posture classification and health monitoring.

## Research Methodology

### This literature review is aimed at conducting a systematic review of similar research studies done on smart sensing chair technology. The research that would be used is the \_\_, which was pioneered by \_\_\_.

### Research Questions

1. What are the sensors that is being used by similar studies?
2. What methods are being used to classify different sitting postures?
3. What are the limitations and research gaps seen with existing studies?
4. How useful was the implemented user feedback mechanism?

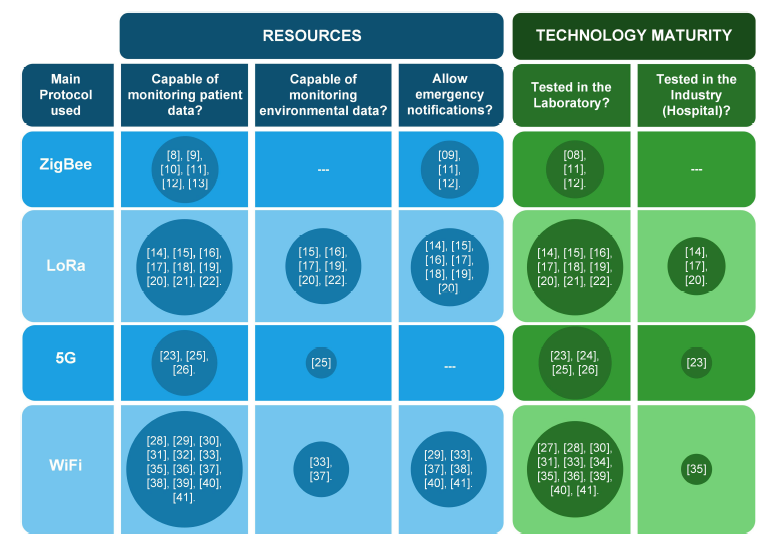
### Search Strategy

Articles that were examined came from various online publication sources such as [**Semantic Scholar**](https://www.semanticscholar.org/), [**Google Scholar**](https://scholar.google.com/), [**IEEE Explore**](https://ieeexplore.ieee.org/Xplore/home.jsp)**, and** [**MDPI journals**](https://www.mdpi.com/about/journals)to name some of them. To aid in the search for the relevant articles though different database systems, a list of important keywords was clearly defined to ensure that the most relevant papers came in the search results. Additionally, some of these “keywords” were combined to achieve better search results. Below are some of the search terms that was used.

1. Smart Sensing Chair
2. Sitting Posture Recognition
3. Posture Classification
4. Sitting Posture Classification using AI/Machine Learning
5. Sitting Posture Monitoring

### Paper Screening

### Systematic Mapping



DO SOMETHING SIMILAR TO THIS (Reno et al)

## Discussion

## Sitting Posture Monitoring Systems

As previously stated, the development of a sitting posture monitoring system is not an entirely new concept, rather it is an area that has been explored by multiple researchers in the past until this present day. This section would be going over different research projects that developed their variation of smart sensing chairs. To efficiently conduct this literature review, at least 30+ relevant research literatures were carefully selected and examined as seen in Table 1. have been published focusing on the of the use of unobtrusive means for the classification of different sitting positions. Systematically examining these papers would surely some shed light on the most common machine learning algorithms and sensors being used to be able to classify various sitting postures.

### History of Smart Sensing chairs

As previously stated, [7] was the first research seen to pioneer the idea of a smart sensing chair that is capable of detecting one’s posture by using pressure distribution sensors integrated into the chair. Over the past few years, various research studies have implemented different variations of these smart sensing chair concepts ranging from different sensors to various classification methods to posture detection as shown in Figure 1. Furthermore, a literature connection map (on similar studies) done on smart sensing chairs was constructed as shown in Figure 2 below. This figure gives a rough visualization of the amount of research being done in landscape of smart sensing chair technology.



Figure – Timeline Map of Similar Literatures

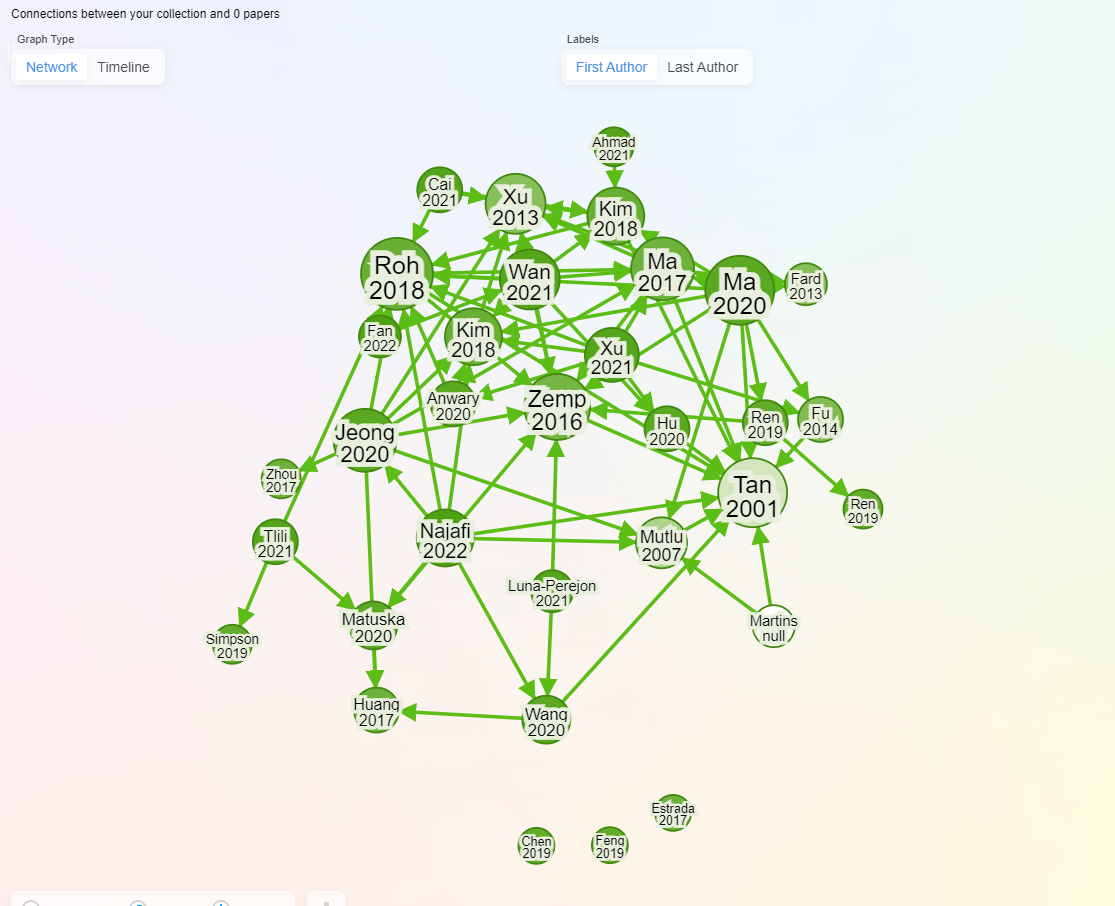


Figure - A Map of Similar Studies on Smart Sensing Chairs

## Sensor systems

As anticipated, various scholarly papers use different types of sensor devices to detect different sitting postures. In summary, they can be divided into 4 overarching categories:

* Smart Sensing Chairs using Pressure Sensors
* Smart Sensing Chairs using Flex Sensors
* Smart Sensor Chairs using Mixed Sensor Systems
* Smart Sensing Chairs using Computer Vision Systems

### Sensing Chair using Pressure Sensors

##### Force Sensing/Sensitive Sensor (FSR)

Force Sensing Resistors are also known as force sensors which are commonly used to measure the forces applied to its surface area. These sensors work by varying their output resistance based on the pressure being applied to it. Typically, the overall resistance decreases as more pressure is applied [8]. To be able to get the reading from this sensor, it normally connected directly to a microcontroller such as an Arduino or like get its reading. Figure 1 shows an example of how a FSR sensor commonly looks like.



Figure - A Force Sensing Resistor

Focusing on studies that integrated FSRs in to their studies, [9] integrated 19 different FSRs into the seating cushion and used the SimpleLogistic ML algorithm to achieve 78% accuracy in classifying 10 different postures. [10] used 13 pressure sensors classify 10 sitting postures and was able to achieve an accuracy of 99.10% using the SVM ML algorithm. [11] applied 8 sensors (4 on the seating cushion and 4 on the back rest) and used EMN algorithm to classify 8 sitting posture and achieved an accuracy of 91.68%. In addition to this, there was a Desktop Graphical User Interface (GUI) application which displayed the senor reading in real-time. [12] add 6 sensors placed on the seating cushion and resulted in an 81.5% classification accuracy using SOM (ISOM-SPR) ML algorithm.

##### Load Cells

Load cells are another variation of force sensor which is commonly used to measure monitor sitting postures. Under the hood, it works by converting the mechanical force being applied to it into digital signals which can be read by microcontrollers. [13] developed a smart chair by integrating 4 load cell sensors within the chair sitting cushion to classify 6 sitting postures. An accuracy of 97.94% was achieved using a SVM (RBF kernel) ML model. Similarly, [14] distributed 3 load cells across the seat’s cushion in order to classify 8 sitting postures; overall they were able to a classification accuracy of 98.50%.

##### Textile Pressure Sensor

A textile-based pressure sensor is normally composed of a soft fabric material. This sensor consists of a conductive thread pattern placed over a dielectric material that serves as a substrate between the threads. Figure 4 shows an example of how each layer within the textile pressure sensor is structured.



Figure - Textile Pressure Sensor composition [15]

A few research studies were found to have used textile sensors to classify sitting postures. One of which was [16], who developed a washable textile pressure sensor and incorporated it into their chair system to classify 7 sitting posture using a decision algorithm. [17] proposed a “eCushion” device which is made up of a textile pressure array sensor that is capable of detect 7 different sitting postures at 85.9% accuracy. [18] also developed something similar by using 10 presence textile capacitive sensor (embroidered) sensors.

### Sensing Chair using Flex Sensors

Flex sensors are another variation of sensors that is being used by various studies to classify different sitting postures. A flex sensor, also known as a bend sensor works by measuring the degree of displacement resulting from the bending action being applied to the sensor [19].

It was seen that the primary use of flex sensors in the classification of sitting postures is not a widely popular approach among various studies. Overall, there were only 2 studies identified that utilized this method for sitting posture detection. The first was by [20] who developed a smart sensing chair using 6 flex sensors and a 2-layer Artificial neural network (ANN) for detecting 7 sitting postures and achieved an accuracy of 97.43%. The second was by [21] which also developed a similar system without the use of an ML model which aimed at detecting 7 different sitting postures.



Figure - Flex Sensor

### Sensing Chair using Mixed Sensors

While most studies utilize a singular type of sensor for posture detection, there are a selected few study that involved more than one type of sensor into their proposed smart chair system. With this method, the different sensors would theoretically work hand in hand to achieve the best classification outcome.

[22] utilized 6 pressure sensors (placed on the seating cushion) along with 6 Infrared Reflective Distance Sensors (placed on the back rest). Using the K-Nearest Network (KNN), they were able to classify 11 different sitting postures while achieving an accuracy of 92%. This study also highlighted one of the main limitations seen with other smart sensing systems. It was stated that the main limitation of entirely relying on pressure sensors is that the angle of spinal trunk rotation can’t be detected, which is an important aspect of a sitting posture. [23] developed a smart seating cushion which employed the use 6 FSR sensors for detecting different sitting postures and an Inertial measurement unit (IMU) sensor to monitor user activity.

### Sitting Posture Classification

Taking an in-depth look at Table 1 it was seen that across all the gathered research papers, there are varying number of postures being classified. The minimum was by [24] who used RFID tag to classify 3 sitting postures (a. Sitting straight, b. Leaning Forward, c. Leaning Backward). On the other hand, [25] looked at detecting up to 15 different postures which was the highest among other studies. Upon further analysis, it was quite evident to see that the more sitting postures that are being classified, the less accuracy its classification accuracy would be. Hence, that is one of the main reasons why most studies limit the number of postures to 5-7.

### User Feedback System

The integration of a feedback system into a smart sensing chair is an integral component. From the end user’s perspective, they should be alerted whenever an improper sitting posture is being detected in real-time. It was seen that most studies focus on the classification aspects and leave out the implementation of a feedback platform. So far only 24% (9) of the studies were found to have incorporated a feedback platform that would encourage the user to maintain a correct posture as seen in Figure 6. A mobile application was seen as the most used platform of notifying a user [26–29]. Another common method was the use of a Desktop application which was done by some studies [10,25,30]. Alternatively, instead of implementing an interactive platform, [31] proposed on the use of an haptic motor system integrated into the seating which would vibrate whenever an incorrect sitting posture is being detected. To even make the system as unintrusive as possible, [32] used an RGB bulb that would change colors if incorrect postures are being detected.

Figure - Feedback System Percentage

### (Internet of Things) IoT Integration and Connectivity

Over recent years, IoT has gained in popularity and has become a game changer within certain industries. [23] emphasized on the use of IoT-Based technology in it being capable of seamless monitoring seating activities.

### Challenges and Limitations

### Research Gaps

* Most research stops at the posture classification phase.
* Not much effort is focused on the user feedback aspects.

### The use and impact of Mobile apps in the healthcare sector

The use of mobile phones in the healthcare sector has rapidly been gaining in popularity in recent times. Mobile Health (MHealth) apps are mobile applications that are mostly tailored towards assisting both medical professionals and patients in the aspects of health management. According to (Lohnar, 2016), the number of mHealth apps are expected to be on an upward trend.

According to various studies, there are a series of hurdles and issues that must be put into consideration during the implementation of a mHealth app.

### Commercialization

### Future Directives and Trends

A computer screen shot of a computer

Description automatically generated

## Conclusion

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## Appendixes

Tables 1 – Sitting Posture Monitoring Research Papers

|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
| **Reference** | **Sensors** | **Number of Postures** | **Classification Method** | **Classification Accuracy** | **Test Subjects** | **Limitations** | **User Feedback System** | **Is Realtime** |
| [14] | Load Cell - Posture Conductive Nappa - ECG | 5 | K-NN | 98.50% | 10 | Not used in a real-life setting | |  |
| [33] | Screen Printed Pressure units (16 Array) | 4 | LightGBM | 99.03% | 32 | Not used in a real life setting | |  |
| [34] | 52 by 44 Piezo-Resistive Sensor Array | 8 | ANN | 92.20% |  | No accuracy |  |  |
| [18] | 10 Presence textile capacitive sensor (embroidered) | 8 |  |  | 5 |  |  |  |
| [27] | 6 Flexible Force Sensors | 9 | Average Standard deviation with 3 Threshold values to determine good/bad postures (Non AI) |  | 12 |  |  |  |
| [11] | 8 Force Sensing Resistors | 8 | EMNM | 92.68% | 40 |  |  |  |
| [29] | A502 Force Sensor & Flex Sensor | 5 |  |  |  | No accuracy |  |  |
| [31] | Pressure Array (IMMM00014, I-MOTION | 7 | 5 Layer Artificial Intelligence | 97.07% | 100 |  |  |  |
| [13] | 4 Load Cells | 6 | SVM using RBF kernel | 97.94% | 24 | No accuracy |  |  |
| [35] | Textile Pressure Sensors (Conductive Ni-Ti alloy fiber) | 7 |  |  |  | No accuracy |  |  |
| [24] | RFID tags | 3 | RF | 99.27% | 14 | Small number of postures | |  |
| [20] | 6 Flexible Force Sensors | 7 | 2 Layer ANN | 97.43% | 11 |  |  |  |
| [22] | 6 Pressure Sensors & 6 Infrared Reflective Distance Sensors | 11 | K-NN | 92% | 36 |  |  |  |
| Martins et al. 2013 | 8 | ANN | 93.40% | 30 |  |  |  | |
| [9] | 10 | SimpleLogistic | 87% | 46 |  |  |  | |
| [36] | 5 | Decision Tree (J48) | 99.47% | 12 |  |  |  | |
| [37] | 16 Force Sensor | 7 | Random Forest | 81% - 98% | 41 |  |  |  |
| [10] | 13 pressure sensors (FSR-406) | 10 | SVM (Linear) | 99.10% | 20 |  |  |  |
| [35] | 8x8 Pressure Sensor | 5 | CNN | 95.30% | 10 |  |  |  |
| [12] | 6 Force Sensitive Resistors (FSR) | 7 | ANN | 81.00% | 12 | Low accuracy |  |  |
| [28] | 3x3 Flexible Array Pressure Sensor | 6 | SOM (ISOM-SPR) | 95.67% | 40 | Few Test Samples | Mobile App |  |
| [38] | 44 × 52 Pressure Sensor Array | 5 | CNN | 99.82% | 8 | \_ Few Test Samples \_ Lack of detection of spine curvatures | N/A | YES |
| [30] | Astra3D Sensor |  | CNN | 90% |  | Privacy issue with Camera Can't work in bad lighting Lack of proper user feedback system for posture correction | PC Screen (Correct/Wrong) | YES |
| [23] | Pressure array | 5 | Decision Tree | 89% |  |  |  |  |
| [39] | 64 Pressure Sensors Array (40x50) cm2 | 4 |  |  | 5 | Limited number of subjects | N/A | NO |
| [40] | 6 Square-Type force Sensing Resistors |  | ANN |  |  |  | RGB LED | YES |
| [25] | 2 Pressure Sensors Array (FSR) | 15 | SNN (LSM) | 88.52% | 19 | Lack of focus on the User feedback aspects Not used in real-life setting to prove it's usefulness | Desktop App | YES |
| [17] | Electrical Textile | 7 | Naive Bayes Network | 85.90% | 14 | The mobile just visualizes the sitting pressure distribution, however the value of this is not really seen to the end user's persepective. No sort of recommendation system. | Mobile App | Yes |
| [41] | Web Camera | 6 | RCNN & CNN | 92.50% |  | No good user feedback/recommendation system | N/A | YES |
| [42] | 8 Force Sensing Resistors (FSR) FSR 406 |  | HMM Decision Tree |  |  | Lacks Proper testing  There's a need to implement a feedback system | N/A | YES |