# Remote Health Monitoring System with Analytics Dashboard

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### **Agenda**

- Introduction to the project
- Literature Review
- Existing problem
- Proposed Solution
- Block Diagram
- Hardware/Software Design
- Experimental Investigation
- Flow Chart
- Result Analysis
- Advantages and Disadvantages
- Application
- Conclusion
- References

### Introduction

- This project addresses the current demand of wearable which could be developed further to function and meet the need of patients effectively.
- Utilization of sensors decreases the possibility of human mistake, ensures better care and treatment, reduces medical expenses, lessens the involved space of the room and improves overall performance.
- This system is much practical in maintaining social distancing and to avoid spread of the Covid-19 or such contagious diseases.

# **Purpose**

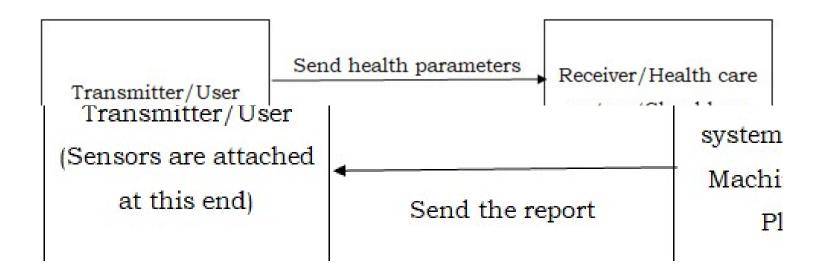
- The purpose of this project is to use ICT in healthcare.
- Bustling time schedule and unpredictable situations of life increases the probability of health risk, independent of the age of a person.
- Though we cannot replace the healthcare system with this, but this project ultimately supplements the existing healthcare system.

# **Existing Problem**

- Hesitation to move to a hospital.
- Busy schedule of the people. The denial will increase the health issue and subsequently results into a health hazard.
- The numbers of health professionals are limited, which increases the personal overhead to each health professionals. For example, a single doctor can check or treat up to a certain number of patients in specific time duration.
- Patients who met with an accident, patients at the time of child-birth (delivery cases) or such patients in critical condition need emergency attention. A wearable can be much helpful in observing the status of such patients. The doctor can observe remotely to such patients till they arrive the hospital.

# **Proposed Solution**

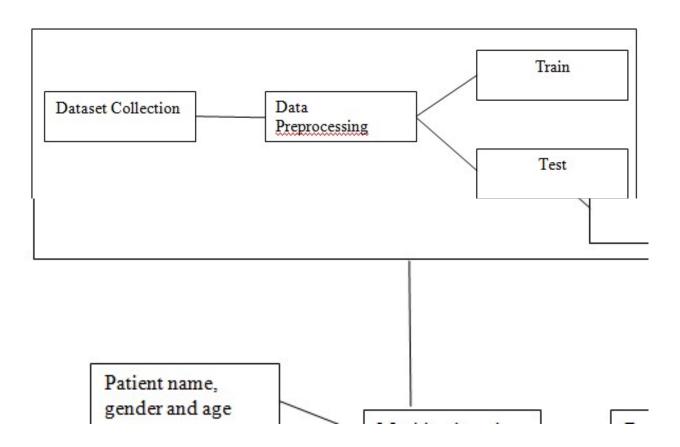
 Integrate sensors with handheld devices or wearable.



# **Block Diagram**

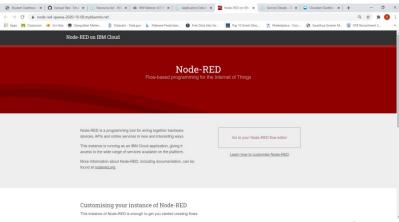
- The system is composed of three components such as,
  - User interface to enter name, gender and age
  - IoT interface to collect the health parameters
  - Machine learning model in the cloud to evaluate the data

# **Block Diagram (Cont...)**



# **Hardware/Software Design**

- IoT Service (IBM Watson IoT Platform)
  - Temperature Sensor
  - Blood Pressure Sensor
  - Pulse Sensor
- Node-RED user interface
- Machine Learning Model design
- Cloudant Database Design

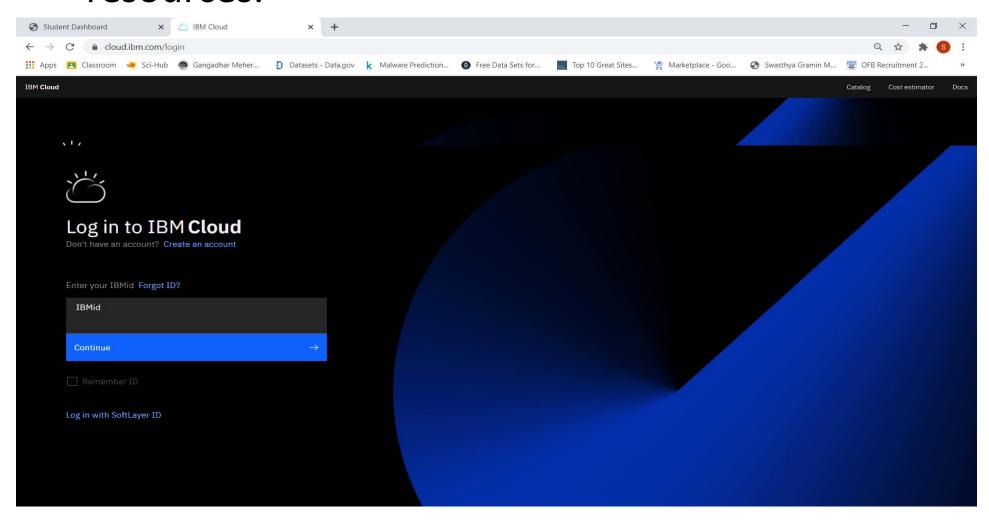


### Procedure for hardware/software design

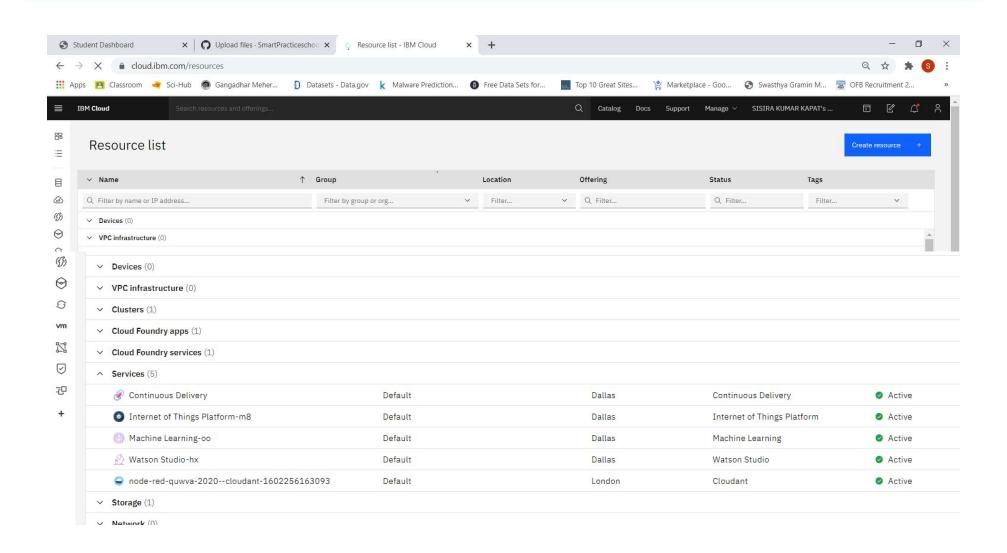
- IBM Academic Initiative account
- Create and launch Node-RED instance
- Create an IBM Watson IoT Platform
- Configure the IoT platform
- Configure the Watson Studio service
- Configure and connect the online simulator
- Build a machine learning model
- Create a node-RED flow to get data fro IBM IoT device
- Create HTTP requests
- Create UI to input user data
- Store the user data in the cloud database
- Create node-RED to display the detailed prediction

# Hardware/Software Design (cont...)

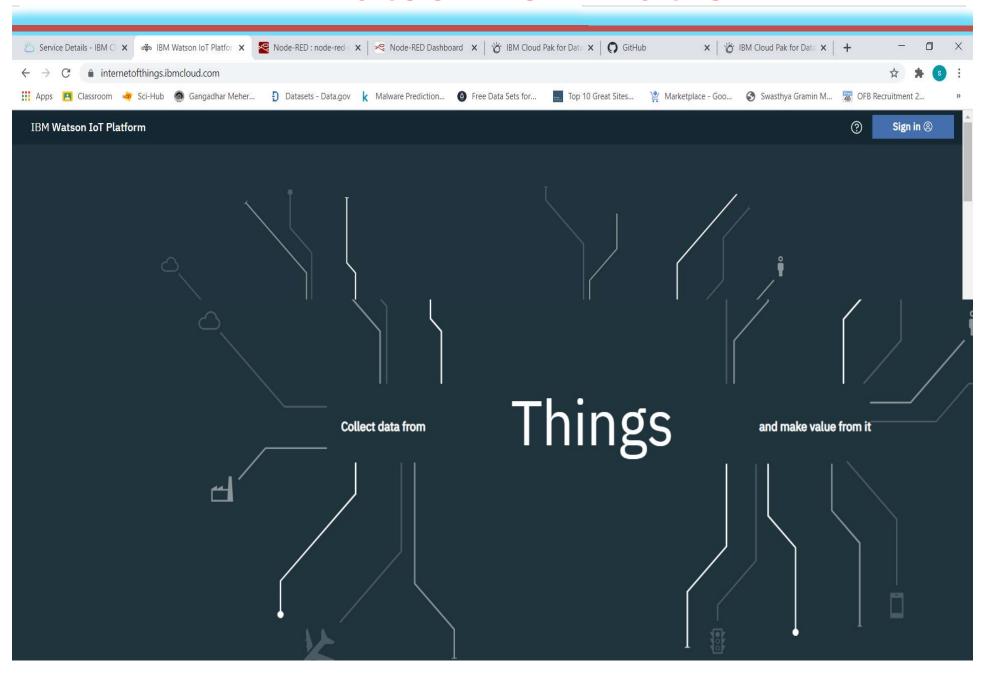
The system is designed by using IBM cloud online resources.



### Resources used



### **IBM Watson IoT Platform**



### **IBM Watson IoT Platform**

 This service is used to simulate the different sensors using IBM IoT.

```
Payload

Specify the event payload in the editor window or by uploading a CSV file.

0 {

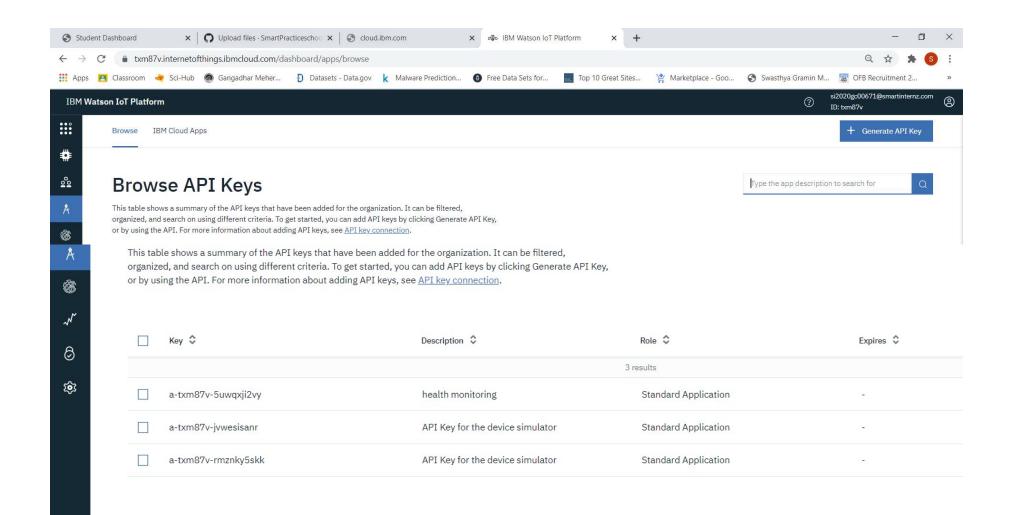
1  "temperature": random(0, 41),

2  "pulse": random(0, 200),

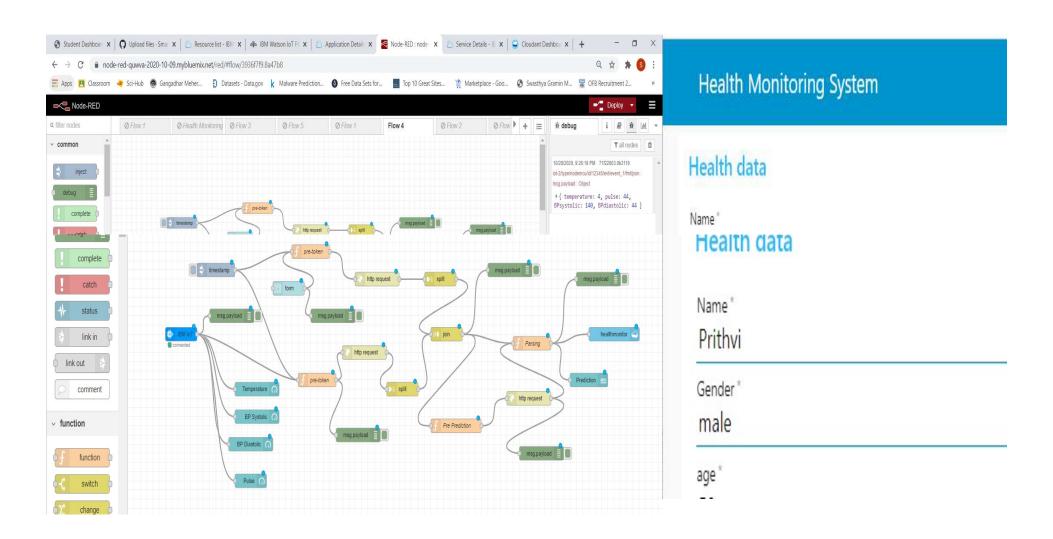
3  "BPsystolic": random(0, 200),

4  "BPdiastolic": random(0, 140)
```

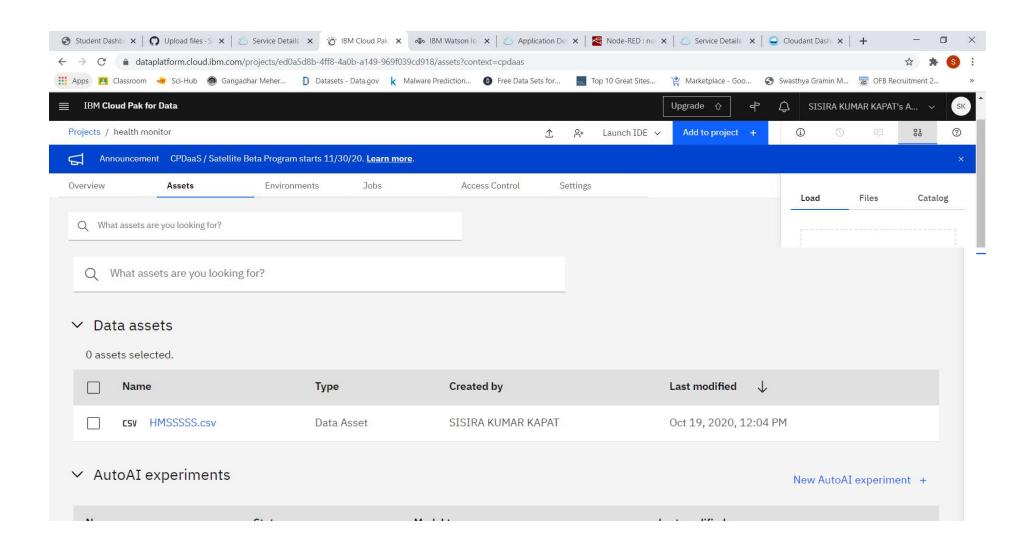
# **API keys**



# **Node-RED UI Creation**



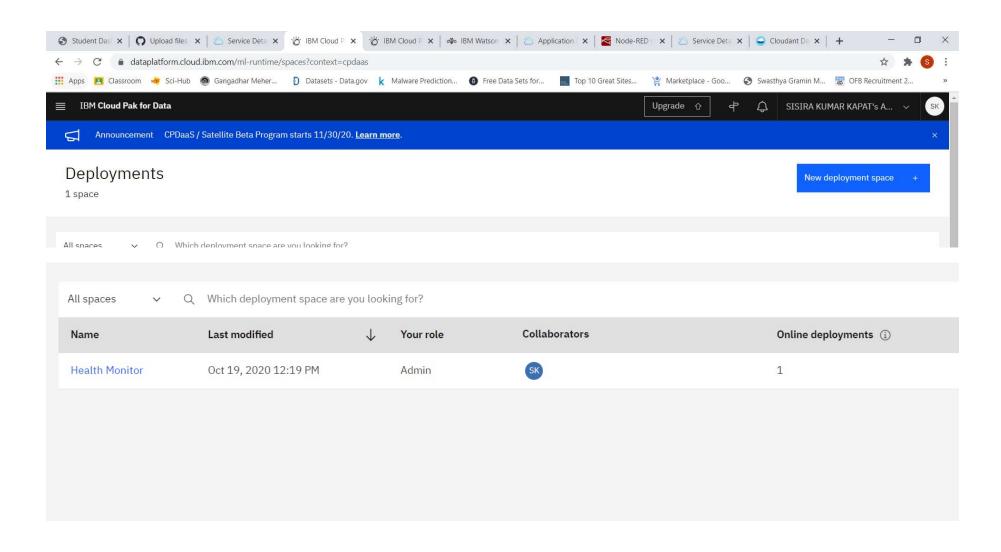
# **Machine Learning model Creation**



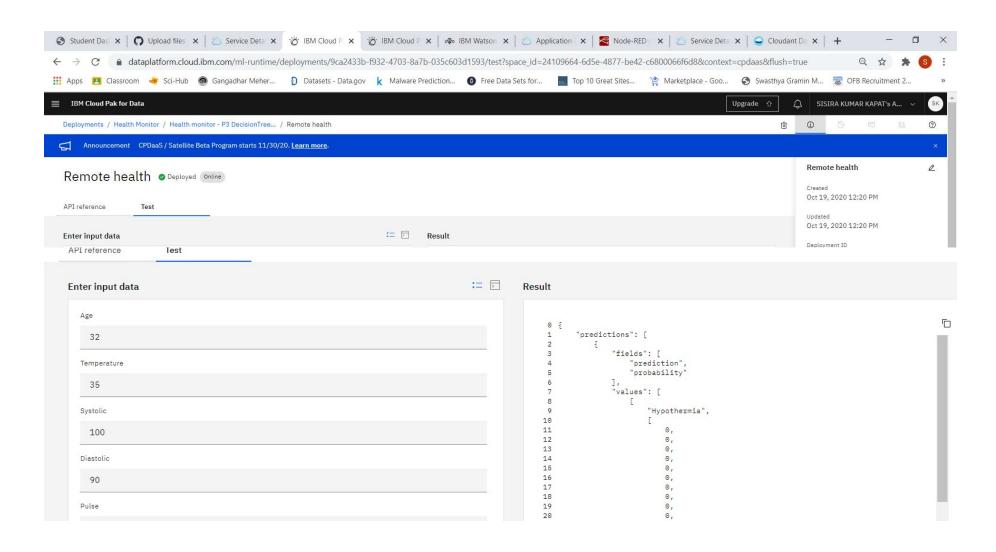
# **Dataset**

Attribute	Range					
Age	0 to 80					
Temperature	0 to 41					
BP (Cistole)	0 to 200					
BP (Distole)	0 to 140					
BP (Distole)	0 to 140					

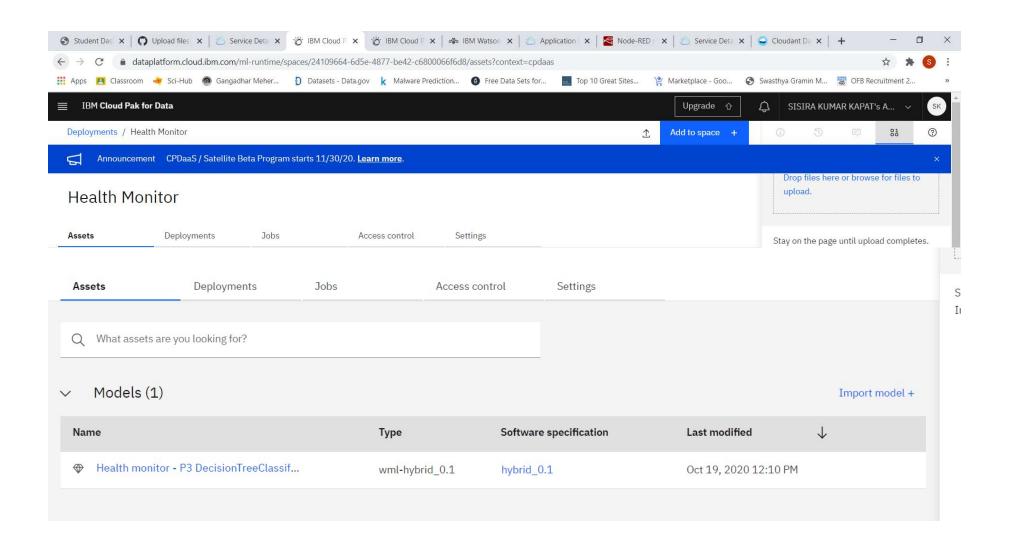
# **Machine Learning model Creation (cont...)**



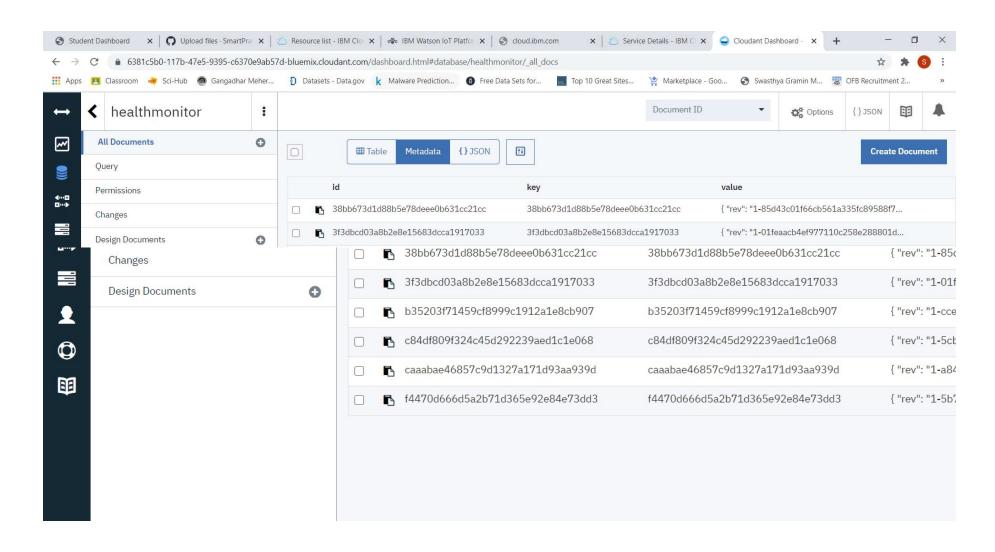
### Machine Learning model Testing(cont...)



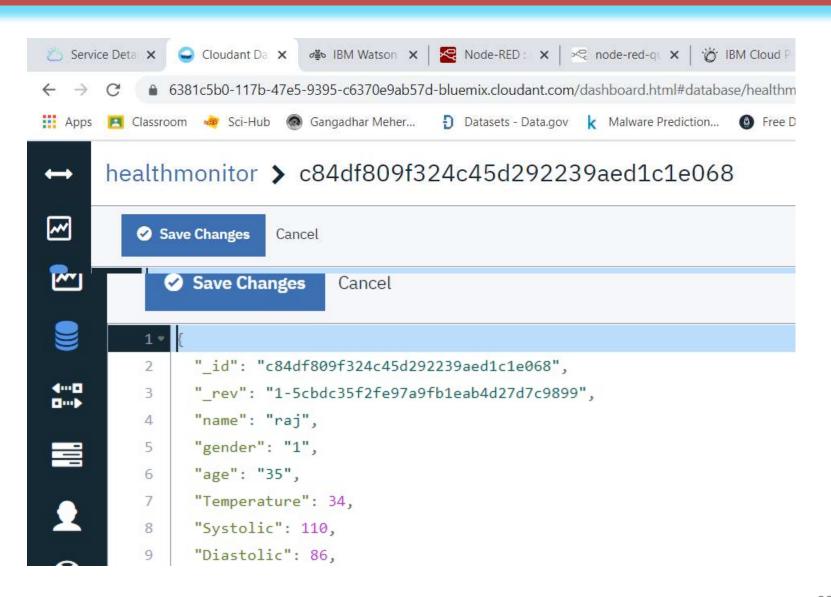
### **Machine Learning model Creation (cont...)**



# **Database Creation**



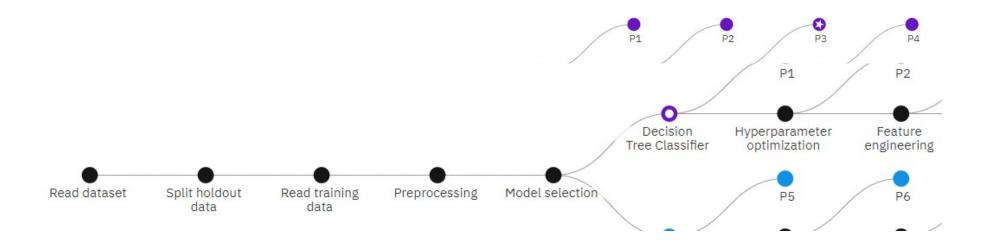
### Data in the database



### **Experimental Investigation (Progress map)**

Progress map ①

Prediction column: status



# Pipeline Leaderboard (CV)

#### Pipeline leaderboard

Rank ↑	Name	Algorithm	Accuracy (Opt	Enhancements	Build time
* 1	Pipeline 3	Decision Tree Classifier	0.959	HPO-1 FE	00:00:36
2	Pipeline 4	Decision Tree Classifier	0.959	HPO-1 FE HPO	00:00:13
	ripeline	4 Decision Tree Class	itier (	לכל.נ	HPO-1 FE HPO-
3	Pipeline	1 Decision Tree Class	ifier (	0.958	None
4	Pipeline	2 Decision Tree Class	ifier (	0.958	HPO-1
5	Pipeline	7 Random Forest Clas	sifier (	0.953	HPO-1 FE
6	Pipeline	8 Random Forest Clas	sifier (	0.953	HPO-1 FE HPO-

### Some Terminologies associated with ML

 The accuracy of the classifier is the probability of correctly classifying the records in the test dataset.

$$Accuracy = \frac{True\ Positive + True\ negative}{Total}$$

 In multi-class classification, the true positive is the sum of all the true positive case of all the pipelines and similarly false positive, false negative and true negative is calculated.

# Some Terminologies associated with ML

 The precision of the classifier is the probability of records actually being in a class if they are classified to be in that class.

$$Precision = \frac{True\ Positive}{True\ Positive + False\ Positive}$$

 The recall of the classifier is the probability that a record is classified as being in a class if it actually belongs to that class.

$$Recall = \frac{True\ Positive}{True\ Positive + False\ N}$$

 The F-measure is the harmonic mean of precision and recall.

$$F - measure = \frac{2 \times Recall \times Pre}{Recall + Precise}$$

# **Detailed Accuracy (CV)**

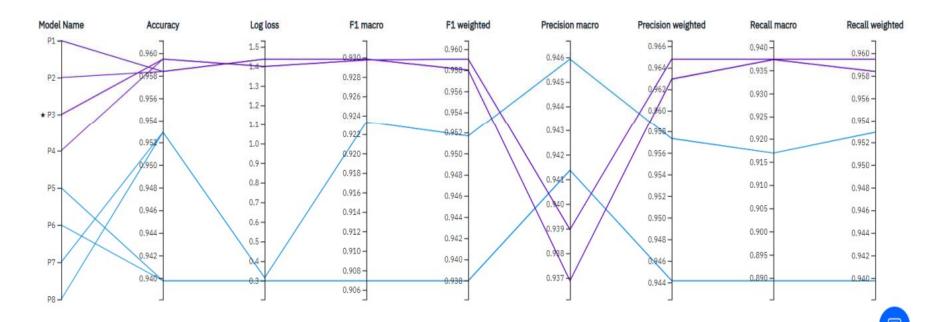
#### Pipeline leaderboard

Rank ↑	Name	Algorithm	Accuracy (Optimiz	F1 macro	F1 micro	F1 weighted	Log loss	Precision m	Precision m	Precision w	Recall macro	Recall micro	Recall weig
<b>*</b> 1	Pipeline 3	Decision Tree Classifier	0.959	0.930	0.959	0.959	1.400	0.939	0.959	0.965	0.937	0.959	0.959
2	Pipeline 4	Decision Tree Classifier	0.959	0.930	0.959	0.959	1.400	0.939	0.959	0.965	0.937	0.959	0.959
3	Pipeline 1	Decision Tree Classifier	0.958	0.930	0.958	0.958	1.437	0.937	0.958	0.963	0.937	0.958	0.958
3	Pipeline 1	Decision Tree Clas	ssifier	0.95	8	0.930	0.9	58 0	.958	1.437	0.937	0.95	8 0
4	Pipeline 2	Decision Tree Clas	ssifier	0.95	8	0.930	0.9	58 0	.958	1.437	0.937	0.95	8 0
5	Pipeline 7	Random Forest Cla	assifier	0.95	3	0.923	0.9	53 0	.952	0.315	0.946	0.95	3 0
6	Pipeline 8	Random Forest Cla	assifier	0.95	3	0.923	0.9	53 0	.952	0.315	0.946	0.95	3 0

#### Comparative analysis of different pipelines of cross validation

#### Metric chart ①

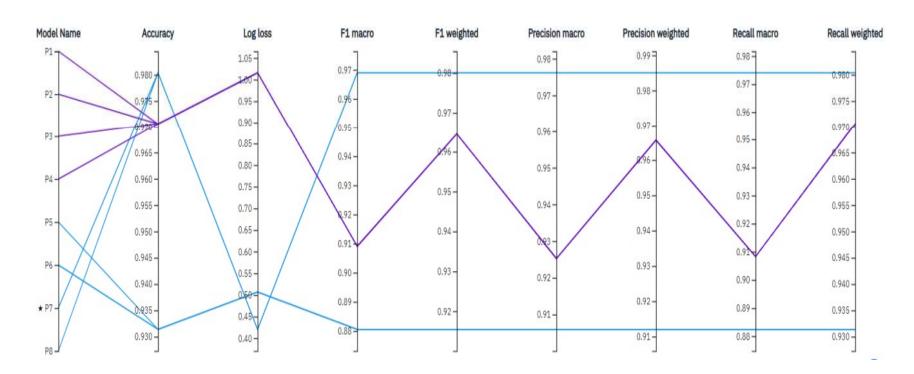
Prediction column: status



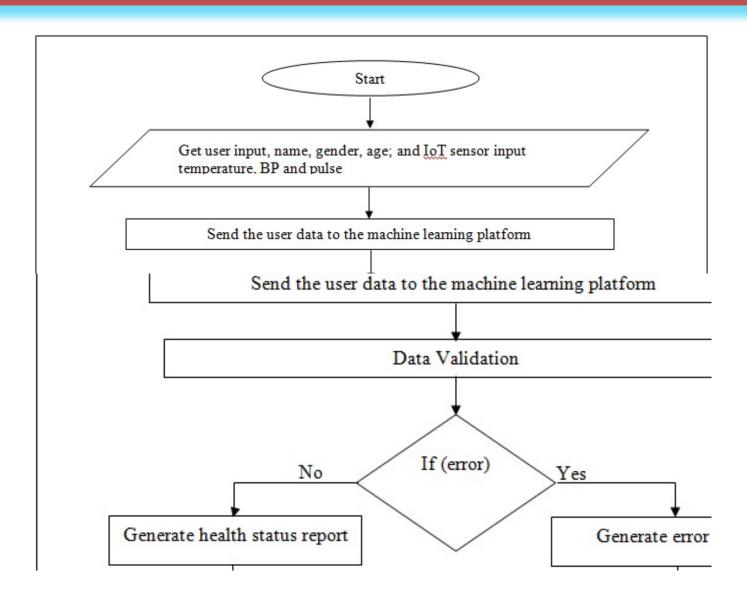
### Comparative analysis of different pipelines of Holdout

#### Metric chart ①

Prediction column: status



### **Flowchart**



# **Result Analysis**

- Input
  - Form input
  - Sensor input
- Output

# Form Input

### **Health Monitoring System**

### Health data

Name "

Prithvi

Name \*

Prithvi

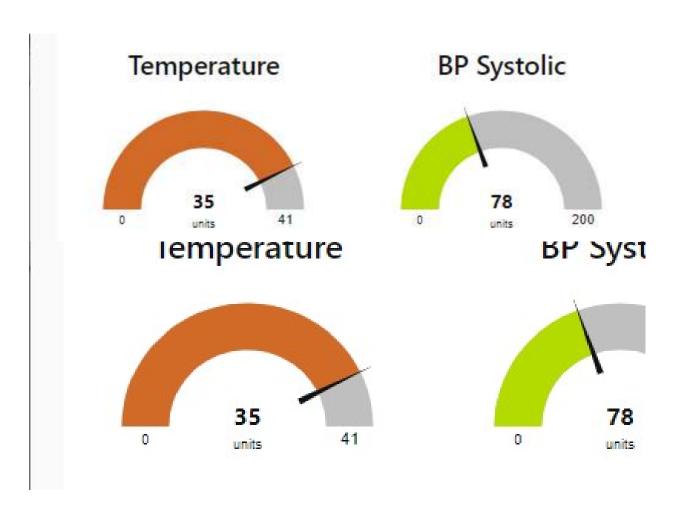
Gender\*

male

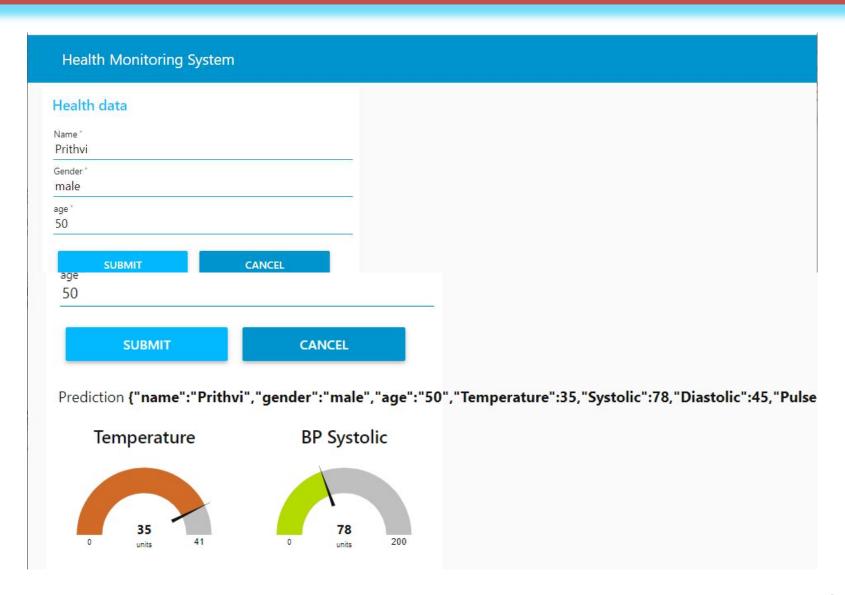
age \*

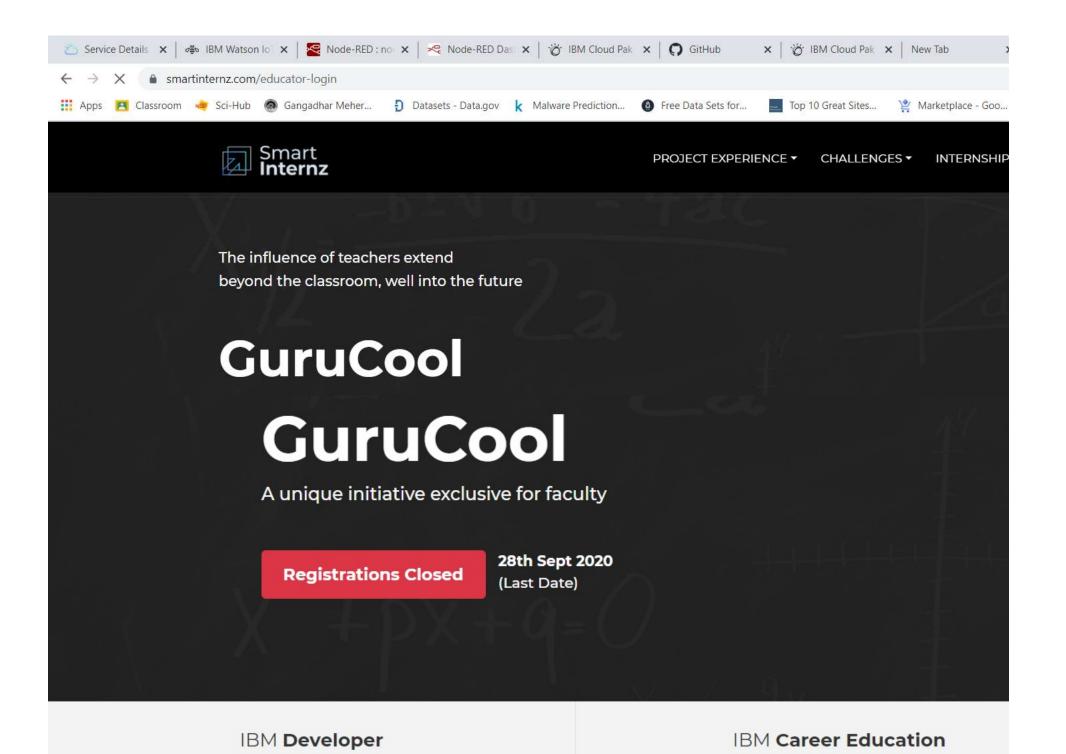
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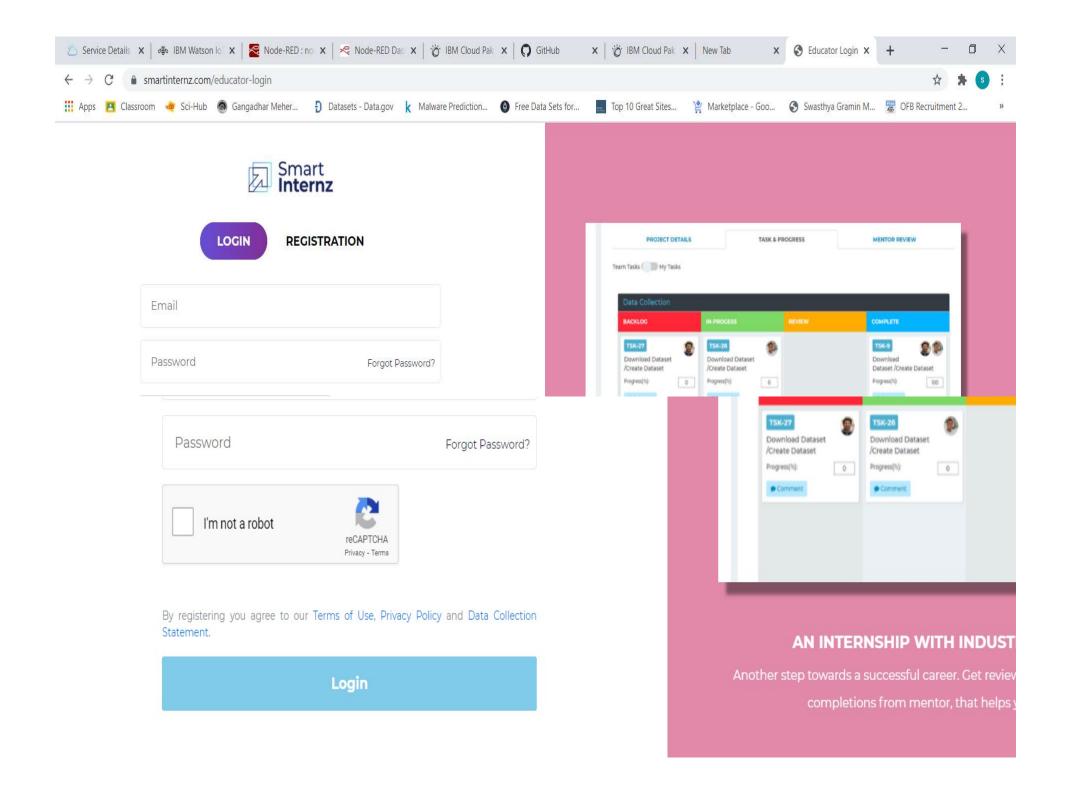
# **Sensor Input**

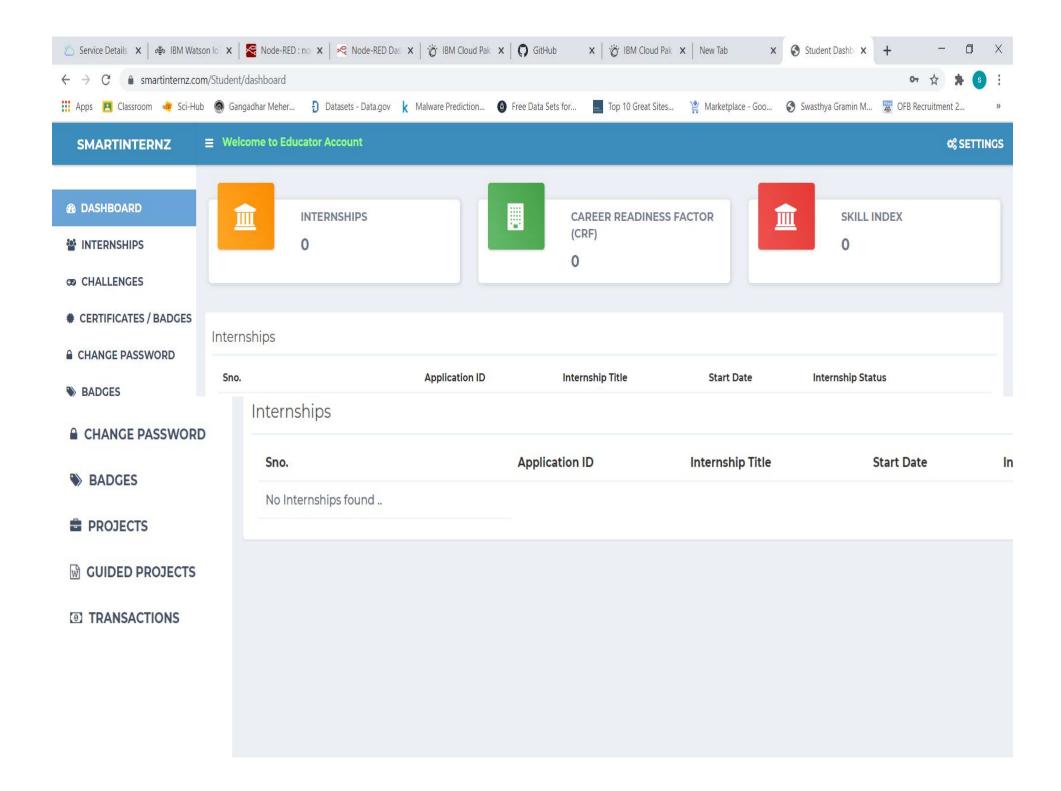


# Output









# **Advantages**

- Quick access by means of the mobile devices.
- Avoid contagious diseases, since this platform avoids rush.
- Daily or hourly status of a patient can be maintained for observation. This facility can be helpful to keep track of the patients who are in critical condition.
- Utilization of sensors decreases the possibility of human mistake.
- Ensures better care and treatment.
- Reduces medical expenses by reducing the travelling frequency.
- Lessens the involved space of the room and improves overall performance

# Disadvantages

- Some of the diseases require privacy and confidentiality. If the database got exposure to the public, then it violates the right to privacy of a person.
- We cannot replicate the healthcare system, no matter how efficient this system is, sometimes the person has to consult the health care professional.

# **Applications**

- handhold devices
- wearable
- Web based monitoring

### **Conclusion and future work**

- The health care system has to be improvised with the evolution and revolution of engineering and technology.
- This will create a positive environment to use the ICT in health care system.
- In this system the overall accuracy 95.9%, which holds good in the prediction but has to be improved.
- This can hamper the right to privacy in some extent which has to be taken care.

### References used

- Ananda Mohon Ghosh, Debashish Halder, SK Alamgir Hossain, "Remote Health Monitoring System through IoT", 5th International Conference on Informatics, Electronics and Vision (ICIEV), pp 921-926, 2016
- Dahlia Sam, S.Srinidhi, V. R. Niveditha, S.Amudha, D. Usha, "Progressed IOT Based Remote Health Monitoring System", International Journal of Control and Automation, Vol. 13, No. 2s, pp. 268-273, 2020
- Ngo Manh Khoi, Saguna Saguna, Karan Mitra, Christer Ahlund, "IReHMo: An Efficient IoT-Based Remote Health Monitoring System for Smart Regions", <a href="https://www.diva-portal.org/smash/get/diva2:1005647/FULLTEXT01.pdf">https://www.diva-portal.org/smash/get/diva2:1005647/FULLTEXT01.pdf</a>
- Mohd. Hamim, Sumit Paul, Syed Iqramul Hoque, Md. Nafiur Rahman, Ifat-Al-Baqee, "IoT Based Remote Health Monitoring System for Patients and Elderly People", International Conference on Robotics, Electrical and Signal Processing Techniques (ICREST), 978-1-5386-8014-8/19, IEEE, 2019

# Thank You...