# Analytics for Hospitals' Health-Care Data

A Project Report Submitted by

Padmacharan D SSNCE193002069
Prasanna K SSNCE193002073
Sidhesh R Allu SSNCE193002099
Tharun Arasu SK SSNCE193002111

Under the Guidance of

Industrial Mentor: Shivam Shivare

Faculty Mentor: Dr.Ramani B

Team ID: PNT2022TMID52975

## **Project Report Format**

## 1. INTRODUCTION

- 1.1. Project Overview
- 1.2. Purpose

## 2. LITERATURE SURVEY

- 2.1. Existing problem
- 2.2. References
- 2.3. Problem Statement Definition

## 3. IDEATION & PROPOSED SOLUTION

- 3.1. Empathy Map Canvas
- 3.2. Ideation & Brainstorming
- 3.3. Proposed Solution
- 3.4. Problem Solution fit

## 4. REQUIREMENT ANALYSIS

- 4.1. Functional requirement
- 4.2. Non-Functional requirements

## 5. PROJECT DESIGN

- 5.1. Data Flow Diagrams
- 5.2. Solution & Technical Architecture
- 5.3. User Stories

## 6. PROJECT PLANNING & SCHEDULING

- 6.1. Sprint Planning & Estimation
- 6.2. Sprint Delivery Schedule
- 6.3. Reports from JIRA

# 7. CODING & SOLUTIONING (Explain the features added in the project along with code)

- 7.1. Website with dashboard
- 7.2. Length of Stay (LOS) Prediction

## 8. **TESTING**

- 8.1. Test Cases
- 8.2. User Acceptance Testing

## 9. RESULTS

9.1. Performance Metrics

## 10. ADVANTAGES & DISADVANTAGES

## 11. **CONCLUSION**

## 12. FUTURE SCOPE

## 13. APPENDIX

Source Code

GitHub & Project Demo Link

## INTRODUCTION

## 1.1. Project Overview

**Project Title:** Analytics for Hospitals' Health-Care Data

**Proposed Solution:** Create dashboard for ease of use and understanding and to predict the length of stay of patients from past health care records.

**Problem:** Patients with high Length of Stay (LOS) have higher risks of spreading infection to other patients and hospital staff.

#### Goals:

- Know fundamental concepts and can work on IBM Cognos Analytics
- Gain a broad understanding of plotting different visualizations to provide the suitable solution.
- Able to create meaningful Visualizations and the Dashboard(s).

**Services Used:** IBM Cognos Analytics.

#### **Deliverables:**

- Create a website with dashboard that can visualise different patient data parameters
- Predict the length of stay of patients from other health records.

## 1.2. Purpose

The main purpose of this project is to accurately predict the Length of Stay for each patient on case-by-case basis so that different Hospitals and other Health Institutes can use this information for optimal resource allocation and better functioning. The length of stay is divided into 11 different classes ranging from 0-10 days to more than 100 days.

#### LITERATURE SURVEY

## 2.1. Existing Problems

Hospital Management face several challenges even after many years of its implementation They include

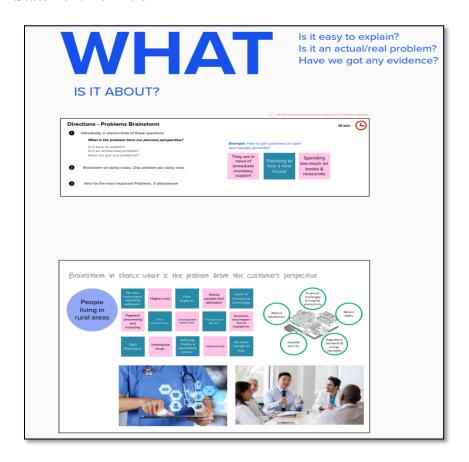
- There is a shortage of professional healthcare faculty with in-depth knowledge of HMS and other similar technologies.
- Poor acceptance of Hospital Management System Software.
- Lack of health informatics professionals capable of establishing and implementing the techniques.

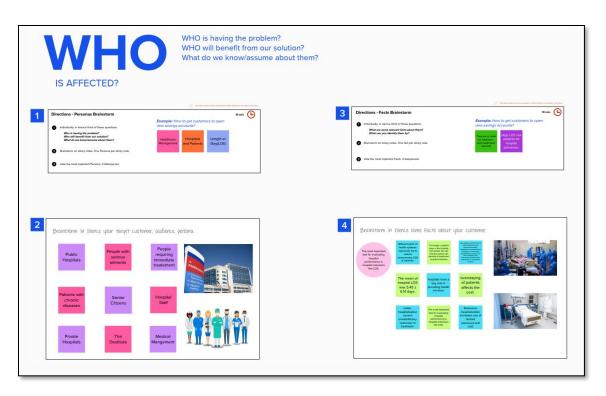
There also exists the problem of patients that stay at medical institutions for longer than required and also increase the risk of spreading the infection to other patients and hospital staff.

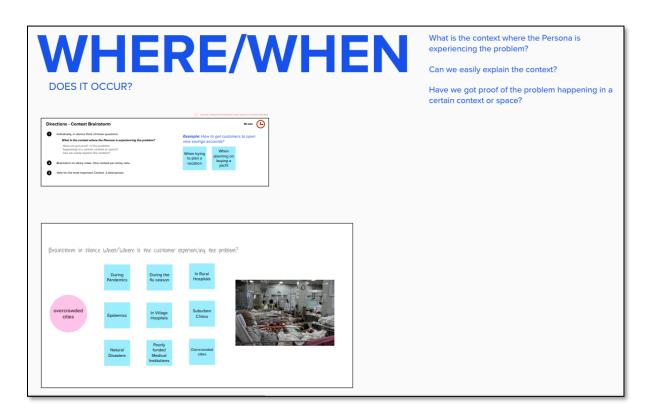
#### 2.2. References

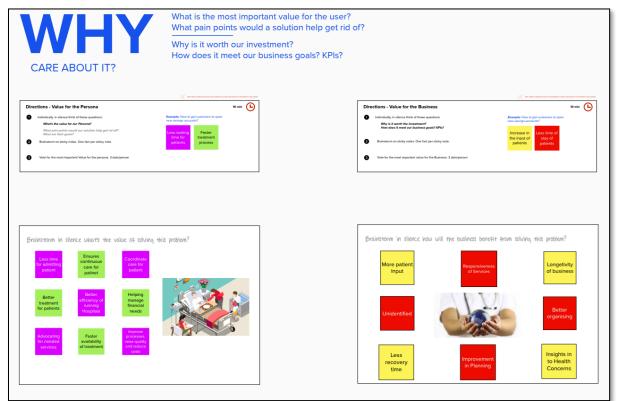
- [1] A. Menon, A. M. S, A. Maria Joykutty, A. Y. Av and A. Y. Av, "Data Visualization and Predictive Analysis for Smart Healthcare: Tool for a Hospital," *2021 IEEE Region 10 Symposium (TENSYMP)*, 2021, pp. 1-8, Doi: 10.1109/TENSYMP52854.2021.9550822.
- [2] Islam MS, Hasan MM, Wang X, Germack HD, Noor-E-Alam M. ASystematic Review on Healthcare Analytics: Application and Theoretical Perspective of Data Mining.Healthcare (Basel). 2018 May23;6(2):54. Doi: 10.3390/healthcare6020054. PMID: 29882866; PMCID: PMC6023432.
- [3] S. Crossfield, O. Johnson, and T. Fleming,"Large ScaleInfrastructure for Health Data Analytics," 2016 IEEE International Conference on Healthcare Informatics (ICHI), 2016, pp. 306-306, Doi:10.1109/ICHI.2016.48.
- [4] Z. Fu, X. Gu, J. Fu, M. Moattari and F. Zulkernine, "Predicting the Length of Stay of Patients in Hospitals," 2021 IEEE International Conference on Bioinformatics and Biomedicine (BIBM), 2021, pp. 3150-3156, Doi: 10.1109/BIBM52615.2021.9669527.

## 2.3. Problem Statement Definition

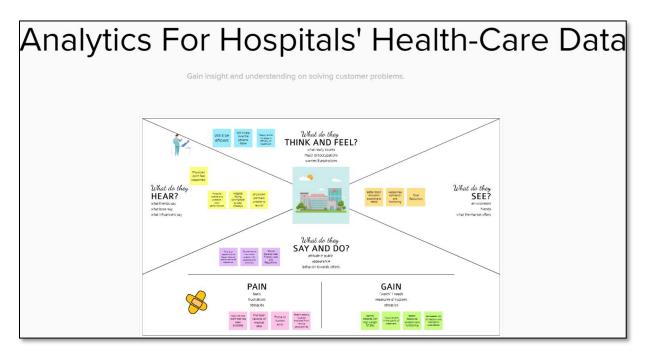




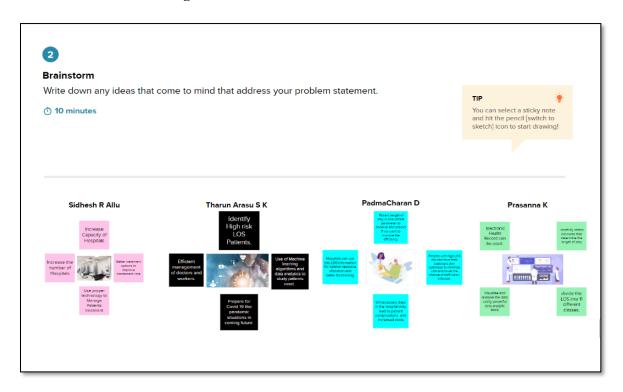


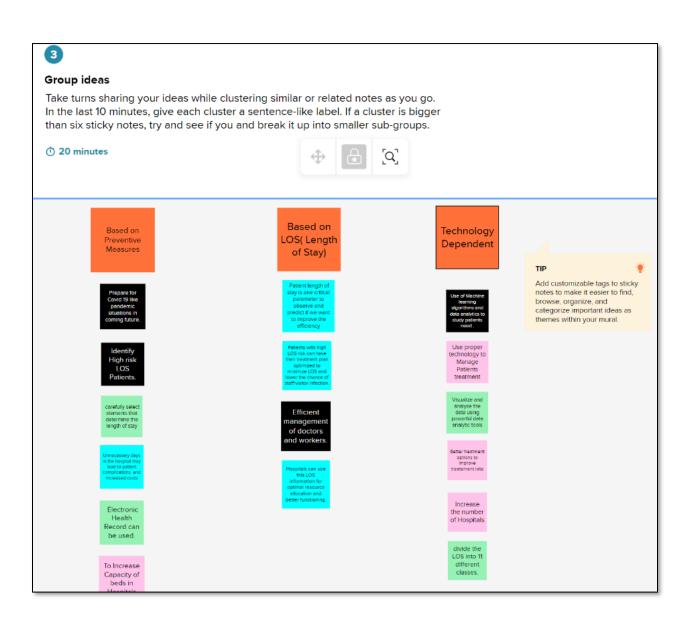


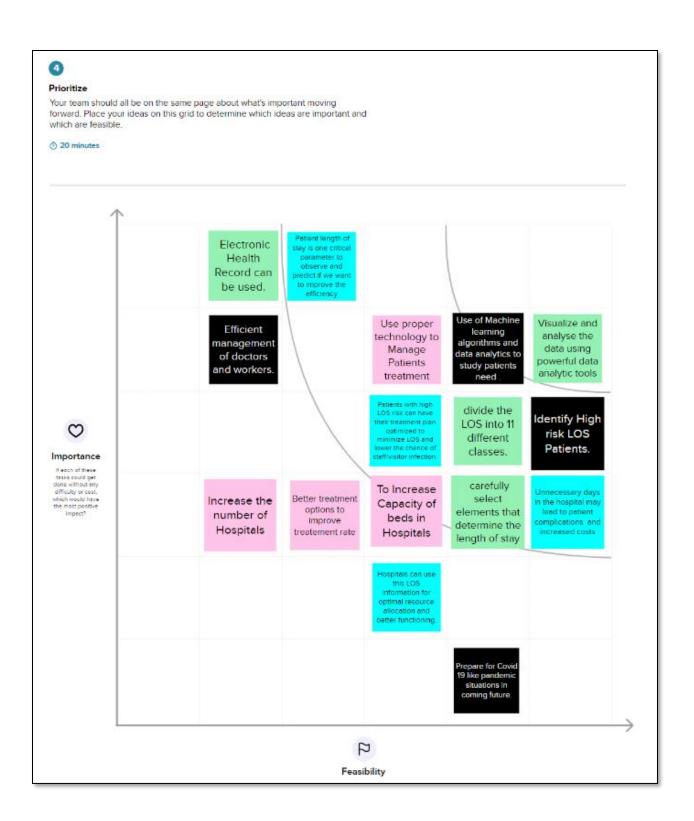
## 3.1. Empathy Map Canvas



## 3.2. Ideation & Brainstorming







# 3.3. Proposed Solution

S. No.	Parameter	Description
1.	Problem Statement (Problem to besolved)	Patients with high LOS (Length of stay) need to have their treatment plan optimized to reduce the chances of spreading the infection to the staffand visitors.
2.	Idea / Solution description	Allocation of hospital resources and staff based on the LOS (Length of Stay) that hasbeendetermined using DataAnalytics and Data visualisation basedon the previoushealth records of patients.
3.	Novelty / Uniqueness	The solution makes use of the patients LOS(Length of stay), which can be classified into 11 different classes. Also relies heavilyon the use of data analytics and data visualisation tools.
4.	Social Impact / CustomerSatisfaction	Efficient management of patients in hospitals leading to faster treatment rates. Hospitals startgrowing, incorporating morespecialities and treatment facilities for the patients with moreLOS (Length of stay).  Judicious use of resources to treat patientsbased on LOS.
5.	Business Model (Revenue Model)	Profitable for healthcare industries due toefficient management in treatment time.
6.	Scalability of the Solution	High Scalability. Proposed solution will help improve the workflow of the hospital management enhancing all treatment processes without hindering the daily operations.

#### 3.4. Problem Solution fit

#### 5. AVAILABLE SOLUTIONS AS 1. CUSTOMER SEGMENT(S) 6. CUSTOMER $\mathbf{CC}$ CONSTRAINTS The available solutions are, All types of health care institutes i. Budget of medical institutions ii. Availability of dataiii. Technological constraint ranging from small clinics to i. Providing correct input ii. Avoid human errors iii. Network stability large hospitals. iv. Consistent data storage v. Improving medical technology vi. Proper hospital management 2. JOBS-TO-BE-DONE / PROBLEMS J&P 9. PROBLEM ROOT CAUSE RC 7. BEHAVIOUR BE The jobs to be done are, i. Upload the patient dataset ii. Prepare Dataset i. Improper treatment plans ii. No proper management in hospitals Improper treatment plans The behavior include, i. Can easily visualize changes iii. Exploring the data iii. Huge number of in data ii. Easy to usc iii. Customizable according to iv. Perform metrics and rules patients. iv. Inadequate number of v. Visualising the data doctors and nurses. v.No proper determination of LOS (length of stay) users preference The Problems are, i. Wrong input ii. Data latency iii. Poor network standard 3. TRIGGER 10. YOUR SOLUTION 8. CHANNELS of BEHAVIOUR The Triggers of the solution Our proposed solution consists of, The channels that support behaviors are i. Using LOS(length of stay) of i. Proper visualization of data ii. Choosing appropriate data i. It takes longer time patient iii. Simple color scheme and smart ii. Inefficient treatment ii. Classify LOS into 11 classes design elements options iii. Provide treatment plan based on iv. Proper marketing and advertising 4. EMOTIONS: BEFORE / AFTER of product iv. Include essential metrics based EM on medical history of patient Before: Expected proper treatment on due After: Happy with on time treatment.

# REQUIREMENT ANALYSIS

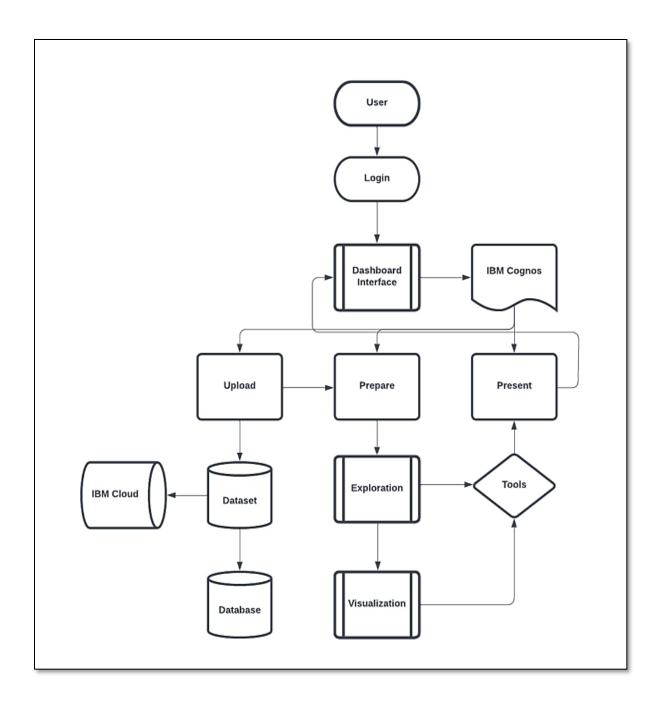
## **4.1. Functional Requirement**

FR No.	Functional Requirement (Epic)	Sub Requirement (Story/ Sub-Task)
FR-1	User Registration	Registration through Google Forms
1 IX-1	User Registration	Registration through Gmail.
FR-2	User Confirmation Confirmation via Email.	
FR-3	A	Dashboard helps predict the patient's Health risks basedon
FK-3	Accuracy	LOS (Length of Stay).
		Dashboard helps to share the patient's information
FR-4	Interoperability	which is compatible with various other hospital
		management systems.
FR-5	Compliance	The dashboard is interactive in real time and is easy to
111-3	Compliance	operate.
FR-6	Concise	The dashboard has a clear understandable display. It is
rk-0	Concise	customizable and is interactive.

## **4.2. Non-Functional Requirement**

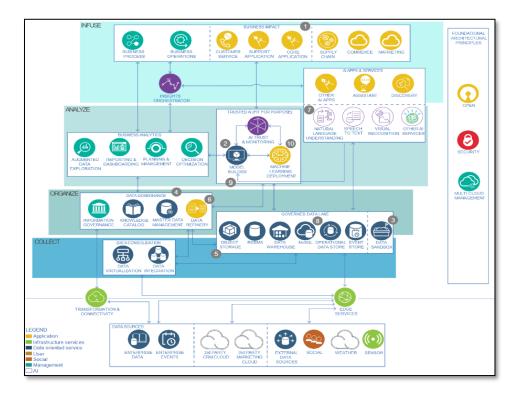
FR No.	Non-Functional Requirement	Description
NFR-1	Usability	The dashboard will display the best treatment plan based on the LOS (Length of Stay) through the use of data visualization tools like pie charts and graphs.
NFR-2	Security	The Dashboard has inbuilt security feature to preventdata breach and loss of data. It also has the added feature of indicating authentication errors.
NFR-3	Reliability	The Dashboard will be consistent and reliable to the users and help the user to in an effective, efficient and reliable manner.
NFR-4	Performance	The Dashboard has a feedback system in which the customers can use and will be used to furtherimprove the functionality.
NFR-5	Availability	The Dashboard is available to meet the users demands in a timely manner and it also helps to provide necessary information to the users dataset.
NFR-6	Scalability	Other Hospital Management facilities can be added with the Dashboard to make a common Hospital Management System

# 5.1. Data Flow Diagram

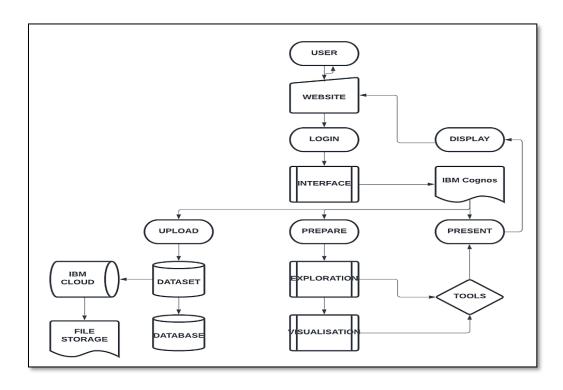


## 5.2. Solution & Technical Architecture

## **5.2.1 Solution Architecture**



## **5.2.2. Technical Architecture**



## **5.3. User Stories**

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer (Mobile user)	Registration	USN-1	As a user, I can register for the dashboard by entering my email, and password, and confirmingmy password.	I can access my account in the dashboard	High	Sprint-1
		USN-2	As a user, I will receive a confirmation email once I have registered for the dashboard	I can receive a confirmationemail& click confirm	High	Sprint-1
		USN-3	As a user, I can register for the dashboardthrough social media	I can register &access the dashboard with Social Media Login	Low	Sprint-2
		USN-4	As a user, I can register for the dashboard through Gmail	I can register and access dashboard with Gmail	Medium	Sprint-2
	Login	USN-5	As a user, I can log into the application by entering email& password	I can login to the accounting my email login.	High	Sprint-2
	Dashboard	USN-6	As a user, I can use my account in my dashboard for uploading dataset.	I can login to the account for uploading dataset.	Medium	Sprint-3
Customer (Webuser)	Website	USN-7	As a user, I can use my dashboard in website	I can login into the dashboard by visiting website.	Medium	Sprint-3
Customer Care Executive		USN-8	As a user, I can contact Customer care Executive for my login.	I can contact customer executive for my login.	High	Sprint-4
Administrator		USN-9	As a user, I can contact administrator for my queries.	I can contact administrator for solving my queries.	High	Sprint-4
Exploration	Dashboard	USN-10	As a user, I can prepare data by using Exploration Techniques.	I can prepare data by using Exploration Techniques.	High	Sprint-3
Presentation	Dashboard	USN-11	As a user, I can Present data in mydashboard.	I can present data byusing my account in dashboard.	High	Sprint-4
Visualization	Dashboard	USN-12	As a user, I can Prepare Data by using Visualization Techniques.	I can prepare data byusing Visualization Techniques.	High	Sprint-3

## PROJECT PLANNING & SCHEDULING

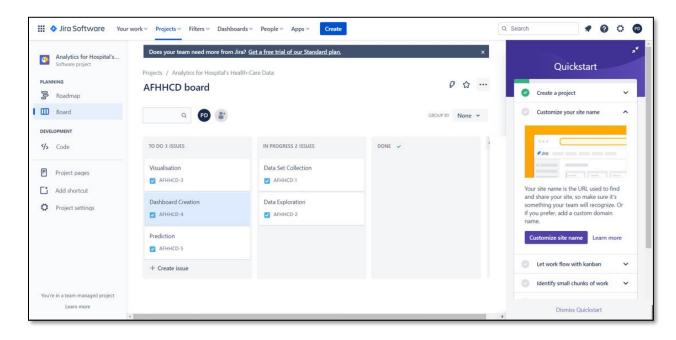
# **6.1. Sprint Planning & Estimation**

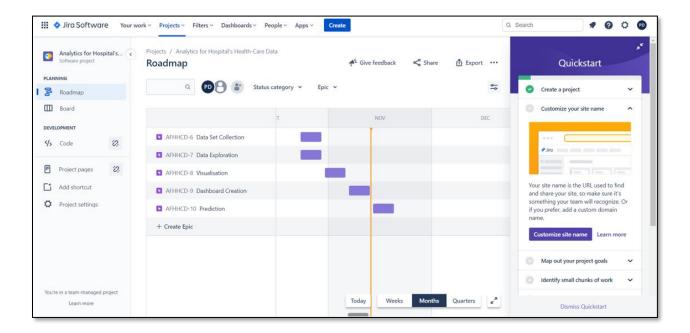
Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Data Set Collection	USN-1	The user requires past health care data about the patient for further analysis.	5	Medium	Padmacharan D, Tharun Arasu SK
Sprint-1	Data Exploration	USN-2	Exploring the data set and cleaning the data if required.	5	High	Padmacharan D, Tharun Arasu SK
Sprint-2	Visualisation	USN-3	User can create various visualisations for better understanding of different parameters.	10	High	Prasanna K, Sidhesh R Allu
Sprint-3	Dashboard Creation	USN-4	For better user experience and viewing of visualisations dashboard can be used.	10	Medium	Padmacharan D, Tharun Arasu SK, Prasanna K, Sidhesh R Allu
Sprint-4	Prediction	USN-5	The user can predict the length of stay (LOS) of patient from past health care data.	10	High	Padmacharan D, Tharun Arasu S.K, Prasanna K, Sidhesh R Allu

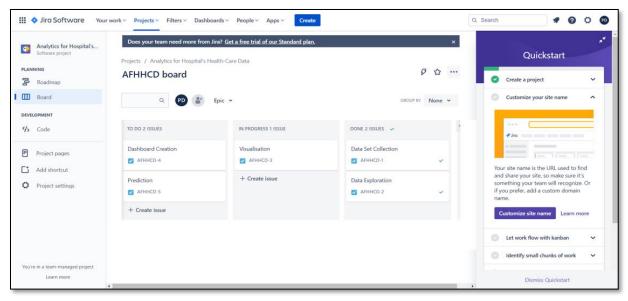
# **6.2. Sprint Delivery Schedule**

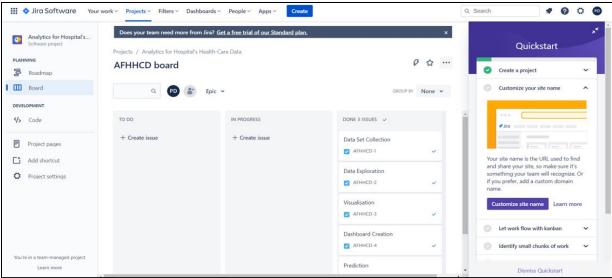
Sprint	Total Story Points	Duratio n	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	20	6 Days	24 Oct 2022	29 Oct 2022	10	29 Oct 2022
Sprint-2	20	6 Days	31 Oct 2022	05 Nov 2022	10	05 Nov 2022
Sprint-3	20	6 Days	07 Nov 2022	12 Nov 2022	10	12 Nov 2022
Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022	10	19 Nov 2022

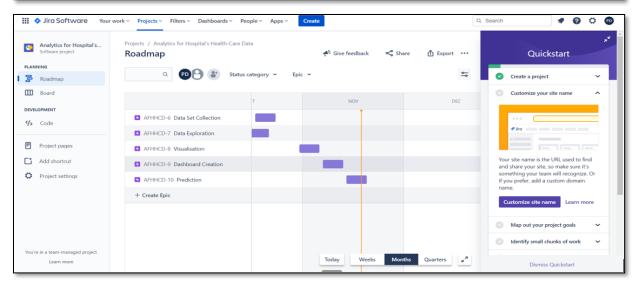
## 6.3. Reports from JIRA







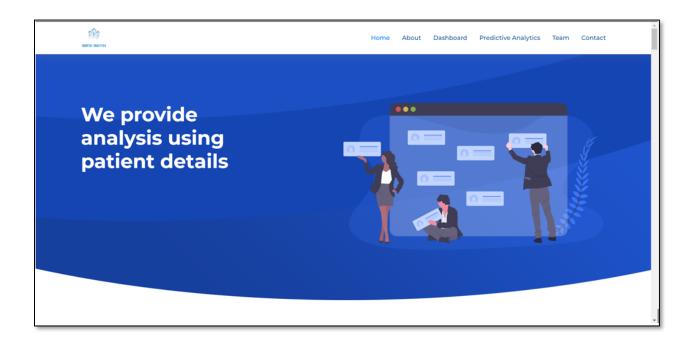


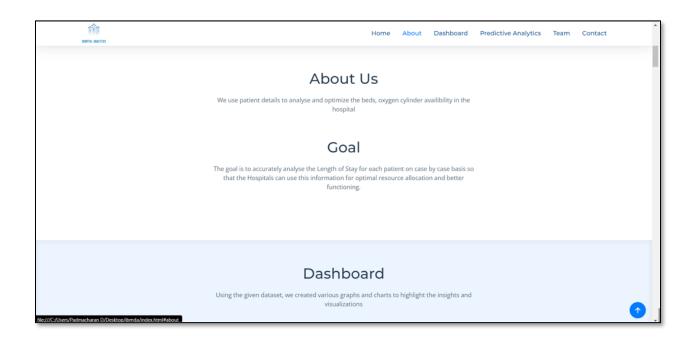


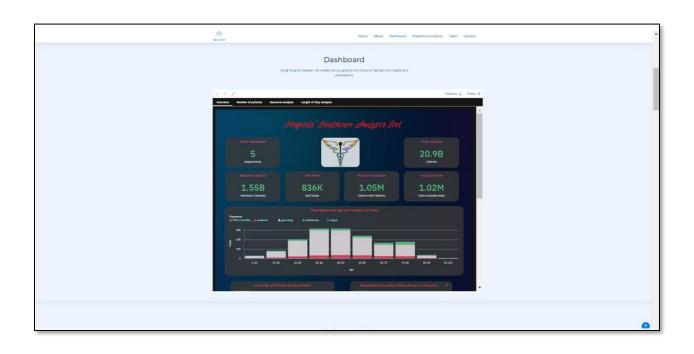
## **CODING & SOLUTIONING**

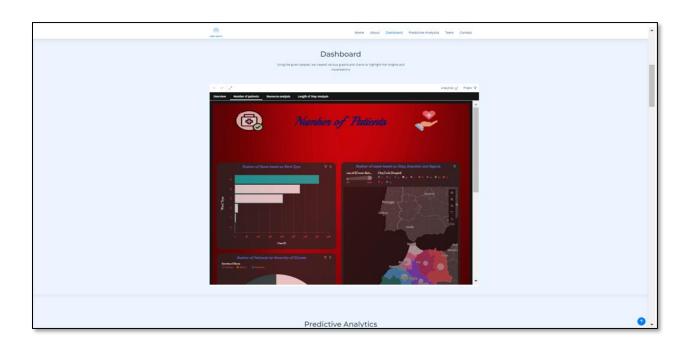
## 7.1. Website with dashboard

A website was created to display different visualisations between different patient data collected from previous health records.

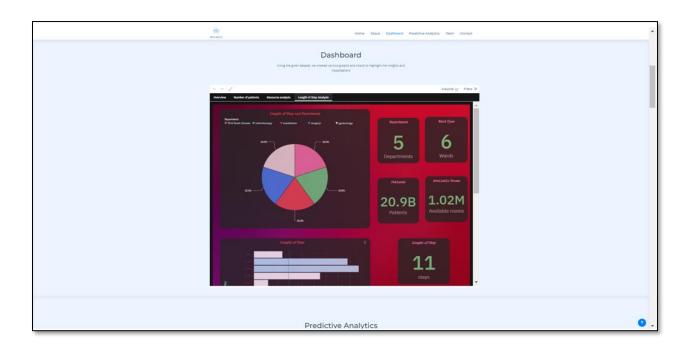












# 7.2. Length of Stay (LOS) Prediction

# **Dataset Description:**

	Column	Description
0	case_id	Case_ID registered in Hospital
1	Hospital_code	Unique code for the Hospital
2	Hospital_type_code	Unique code for the type of Hospital
3	City_Code_Hospital	City Code of the Hospital
4	Hospital_region_code	Region Code of the Hospital
5	Available Extra Rooms in Hospital	Number of Extra rooms available in the Hospital
6	Department	Department overlooking the case
7	Ward_Type	Code for the Ward type
8	Ward_Facility_Code	Code for the Ward Facility
9	Bed Grade	Condition of Bed in the Ward
10	patientid	Unique Patient Id
11	City_Code_Patient	City Code for the patient
12	Type of Admission	Admission Type registered by the Hospital
13	Severity of Illness	Severity of the illness recorded at the time o
14	Visitors with Patient	Number of Visitors with the patient
15	Age	Age of the patient
16	Admission_Deposit	Deposit at the Admission Time
17	Stay	Stay Days by the patient



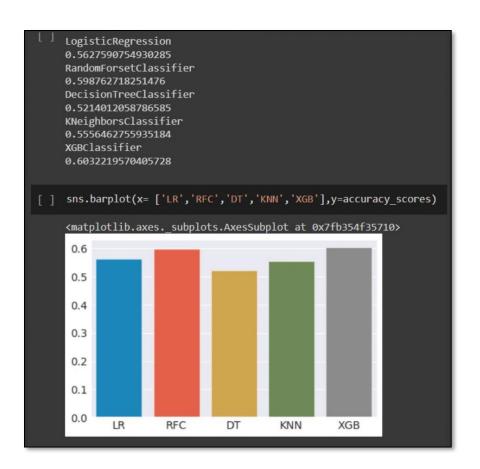
train.Stay.value_counts()				
21-30	87491			
11-20	78139			
31-40	55159			
51-60	35018			
0-10	23604			
41-50	11743			
71-80	10254			
More than 100 Days	6683			
81-90	4838			
91-100	2765			
61-70	2744			
Name: Stay, dtype:	int64			

## **Building Model:**

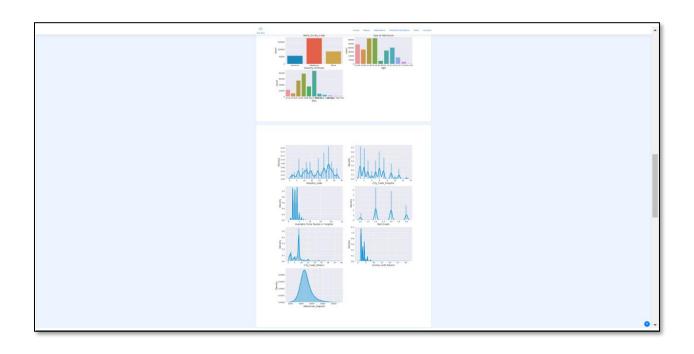
The ML model makes use of Logistic Regression, Random Forest Classifier, Decision Tree Classifier, KNN Classifier and XGB Classifier.

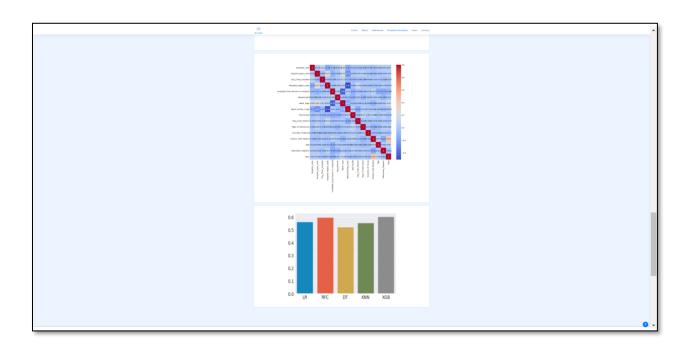
```
Building Model
[ ] from sklearn.linear_model import LogisticRegression
     from sklearn.ensemble import RandomForestClassifier
     from sklearn.naive_bayes import MultinomialNB
     from sklearn.neighbors import KNeighborsClassifier
     from xgboost import XGBClassifier
     from \ sklearn. \verb|metrics| import| accuracy\_score, \ confusion\_matrix, \ classification\_report|
     from sklearn.model_selection import RandomizedSearchCV
[ ] value= [LogisticRegression(), RandomForestClassifier(), DecisionTreeClassifier(), KNeighborsClassifier(), XGBClassifier()]
     key= ['LogisticRegression', 'RandomForsetClassifier', 'DecisionTreeClassifier', 'KNeighborsClassifier', 'XGBClassifier']
     models= dict(zip(key,value))
     accuracy_scores=[]
     for key, value in models.items():
         value.fit(X_train,y_train)
         y_pred= value.predict(X_test)
         accuracy= accuracy_score(y_test, y_pred)
accuracy_scores.append(accuracy)
         print(key)
         print(accuracy)
```

From the results we can infer that XGB Classifier gives the best accuracy for the given dataset. Thus, XGB Classifier is used to predict the Length of Stay of patients.





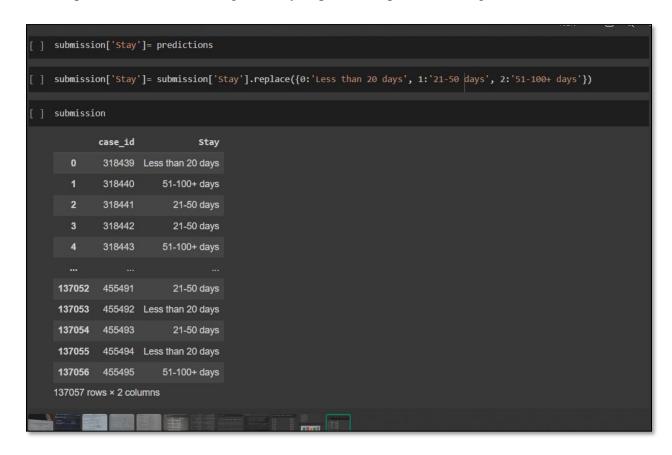




#### **TESTING**

#### 8.1. Test Cases

The ML model was trained using the given train data set. The XGB Classifier's hyperparameters are tuned according to test data set and the predictions have been done using sample sub dataset. The Length of stay of patients is predicted using this model.



## **8.2.** User Acceptances Testing

## 8.2.1. Defect Analysis

This report shows the number of resolved or closed bugs at each severity level, and they were resolved

Resolution	Severit y 1	Severity 2	Severity 3	Severity 4	Subtotal
By Design	6	2	2	0	10
Duplicate	1	0	0	1	2
External	1	4	1	2	8
Fixed	5	0	6	6	17
Not Reproduced	1	1	0	1	3
Skipped	1	1	0	0	2
Won't Fix	0	1	2	1	4
Totals	15	9	11	11	46

## 8.2.2. Test Case Analysis

This report shows the number of test cases that have passed, failed, and untested

Section	Total Cases	Not Tested	Fai l	Pass
Design	8	2	1	5
Dashboard	15	4	2	9
Responsiveness	10	2	0	8
Exception Reporting	17	5	2	10
Final Report Output	13	3	1	9

#### 9.1. Performance Metrics

An ML model was trained to predict the Length of Stay of patients. It makes use of XGB Classifier Model. The model predicts the Length of Stay with an accuracy of 64%. The Length of Stay of 1,37,057 patients has been predicted based on 11 different parameters.

```
y pred= grid.best estimator .predict(X test)
print(accuracy score(y test,y pred))
print(confusion matrix(y_test,y_pred))
print(classification report(y test,y pred))
0.6439203617635976
[[10095 9637 617]
  5507 22391 2981]
   636 3300 8524]]
             precision
                          recall f1-score
                                              support
          0
                  0.62
                             0.50
                                       0.55
                                                20349
                                       0.68
                  0.63
                             0.73
                                                30879
                  0.70
                             0.68
                                       0.69
                                                12460
                                       0.64
                                                63688
    accuracy
                  0.65
                             0.64
                                       0.64
                                                63688
   macro avg
weighted avg
                  0.64
                             0.64
                                       0.64
                                                63688
```

The final website showcases the data visualisations between various health data that has been provided in the data set. The ML model has been used to predict the Length of Stay of patients which makes it easy for the Hospital Management to give the best treatment option to patients which further makes hospitals very efficient.

The users found the deployed website very easy to use and were comfortable using the dashboard which is embedded on the website.

# **9.2. Model Performance Testing**

S. No.	Parameter	Screenshot / Values
1.	Dashboard design	No of Visualizations – 13 Visualizations
2.	Data Responsiveness	Yes, the dashboard visualizations change according to selected parameter.
3.	Amount Data to Rendered (DB2 Metrics)	0KB
4.	Utilization of Data Filters	Data Filters were used which include ascending, descending and format.
5.	Effective User Story	No of Scene Added – 8
6.	Descriptive Reports	No of Visualizations / Graphs - 11 Summary Values and 13 Visualizations.

#### ADVANTAGES AND DISADVANTAGES

## **Advantages:**

- The length of stay can be determined based on which the patient can be given a personalised treatment plan.
- The patient is given a better treatment plan to reduce length of stay there by reducing the chances of spreading infection to other patients and staff.
- The visualisations on the dashboard can help patients understand the connection of different health data.
- The dashboard is customizable and can be personalised for each patient.
- Visualising the data helps increase efficiency of hospitals form the health care management viewpoint as they can allocate proper resources for patients based on the length of stay.

## **Disadvantages:**

- Huge amount of patient data is required for accurate results. Data Collection is a tedious and time-consuming process.
- Proper encryption should be provided to secure patient's health records.

#### **CONCLUSION**

The visualisations on the dashboard help us understand the relation between different health parameters. An ML model was trained to predict the length of stay of a patient based on past medical data.

The following can be inferred from visualising and analysing past health care records

- The most common values of Age are 41-50 (20 %) and 31-40 (20 %), together occurring over 127 thousand times, which is 40 % of the total.
- The most common value of Department is gynaecology, occurring over 249 thousand times, which is 78.3 % of the total.
- 40% of patients are people within the age limit 31-50.
- Gynaecology Department is the department where patients count is very high.
- In the department of Gynaecology, available extra rooms are very high.
- Even in gynaecology department the ward types R, Q, and S have very high number of patients
- Overall, in a hospital number of cases in R type ward is very high.
- Most of the patients are moderately ill.
- In gynaecology department most of the admissions are emergency.
- In Morocco, the hospitals with city code 1 and 2 are visited by more number of patients.
- The admission deposits are high in the department of surgery.
- In the gynaecology department, the bed grade 3 is highly used.
- Ward\_Type\_CAT40451 is the more commonly used ward type in the hospital.
- The number of stays is high in gynaecology department.
- More number of patients are stay in the hospital for 11 30 days

## **FUTURE SCOPE**

The project can be further extended to help coordinate resources among all Hospitals in a particular location. Patient Health Records can be used to further predict and analyse the best type of treatment plan for a particular ailment. Data regarding different medical procedure machines can also be mentioned and their usage can be displayed on the dashboard so that hospitals which require the machine can be allocated with the appropriate resources as required.

Further resource allocation will help reduce the cost of various treatment plans and will help hospital management by making the health industry very efficient.

## **APPENDIX**

## **Source Code:**

## **Website Code:**

https://github.com/IBM-EPBL/IBM-Project-19798-1659706601/blob/main/Final%20Deliverables/Webpage%20Code.html

## **ML Code:**

 $\frac{https://colab.research.google.com/drive/12TYMppgGEx6kD221q4mGAo96Cjtmj1A0?us}{p=sharing}$ 

## **GitHub Link:**

https://github.com/IBM-EPBL/IBM-Project-19798-1659706601

## **Project Demo Link:**

 $\underline{https://drive.google.com/file/d/1BwtzA2InJPyFPNfM1cBXcHMQXjbUpID0/view?usp=share\_link}$