A Minor Project Report On ANALYTICS FOR HOSPITALS HEALTH-CARE DATA

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INTRODUCTION

1.1 PROJECT OVERVIEW

Today's healthcare industries are moving from volume-based business into value based business, which requires an overwork from doctors and nurses to be more productive and efficient. This will improve healthcare practice, changing individual life style and driving them into longer life, prevent diseases, illnesses and infections. In order, it's also going to develop a healthcare data analytic technique that will promise for a better medical practice and healthcare data predictive analytics based on filling gaps of traditional healthcare data analytics techniques and overcoming its problems. Moreover, it's also aims to promise professionals of a better quality of medical results, as well as reduce time needed to analyze healthcare data by keeping systems up to-date and sorting medical data in a logical structure along with accessing and retrieving patient's historical data fast and smoothly.

1.2 PURPOSE

In the area of healthcare data analytics, hospitals and clinics are looking for a new data mining techniques that will suite evolution of information technology and analyze a huge amount of complex data. The proposed technique is recommended rather than offered, since it will facilitate and enhance healthcare practice, by enabling systems to use data and analyze it efficiently and smoothly, because it will fill the gaps of previous techniques used in the hospitals, handle big data issue and avoid data loss, which will lead to improve care, assist diseases prediction and prevention systems and reduce cost. This technique is promising a better results and more benefits if it's applied correctly and properly.

LITERATURE SURVEY

2.1 EXISTING PROBLEM

Implementation of Big Data in Hospital using Cluster Analytics. In the existing system technology is needed to develop rapidly, including in the health sector. Any industry and any countries, big data technology has become a significant database where the data or information to generate can be used in medicine, one of which is applied in hospitals, clinics and the private sector. Big data is evolutionary and software in a suitable environment is being developed again. Health care data drives big data, for example data on patients with heart failure and with this data, authors can use data analytic tools to detect this so that it can be anticipated so that it can reduce the death rate due to this. The problem is that many hospitals have not applied technology to detect their health, and seeing from the times the technological era is increasingly advanced, including in the field of hospitals. To get this solution, use clustering analytics with method K-means using rapid miner studio. Hopefully this analytics that can help research in medical field to do a prediction for heart failure symptoms.

2.2 REFERENCES

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Castro, D 2007, Improving Health Care: Why a Dose of IT May Be Just What the DoctorOrdered, The information technology and innovation foundation, U.S.

2.3 PROBLEM STATEMENT DEFINITION

Recent Covid-19 Pandemic has raised alarms over one of the most overlooked areas to focus is Healthcare Management. While healthcare management has various use cases for using data science, patient length of stay is one critical parameter to observe and predict if one wants to improve the efficiency of the healthcare management in a hospital. This parameter helps hospitals to identify patients of high LOS-risk (patients who will stay longer) at the time of admission. Once identified, patients with high LOS risk can have their treatment plan optimized to minimize LOS and lower the chance of staff/visitor infection. Also, prior knowledge of LOS can aid in logistics such as room and bed allocation planning. Suppose you have been hired as Data Scientist of Health Man – a not for profit organization dedicated to manage the functioning of Hospitals in a professional and optimal manner.

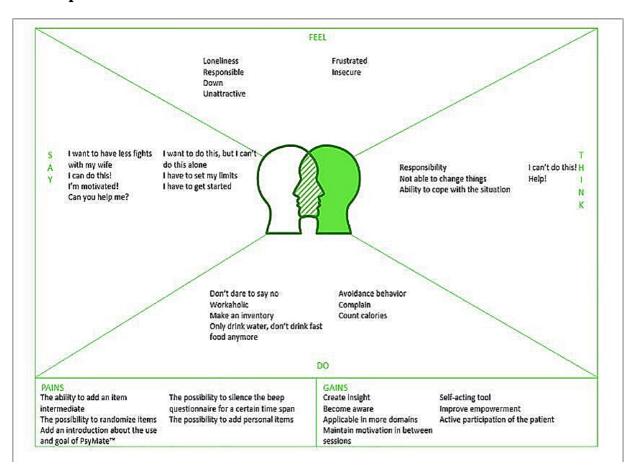
I am	The person to analyze the health care data using data analytics.	
I'm trying to	Use the recent technologies to predict the Analytics for	
	Hospitals' Health-Care Data.	
But	I am unaware of the existing technology that can help me a lot to analyze the health care data and I don't know to use the correct technology.	
Because	I don't want to waste the cost and time	
Which makes me feel	I want a best accuracy which can analyze the health care data so that the people can move with their necessary treatments.	

3. IDEATION & PROPOSED SOLUTION

3.1 EMPATHY MAP CANVAS

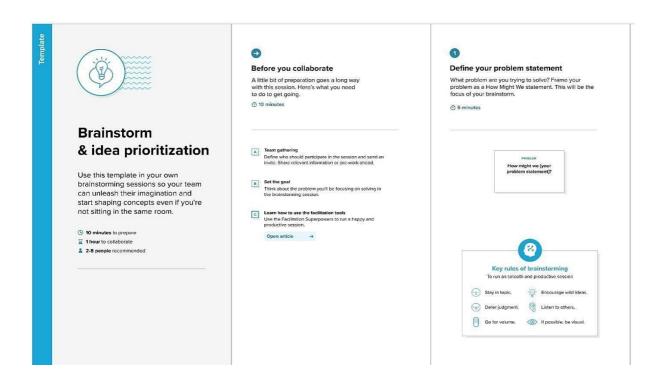
An empathy map is a simple, easy-to-digest visual that captures knowledge about a user's behaviours and attitudes. It is a useful tool to helps teams better understand their users. Creating an effective solution requires understanding the true problem and the person who is experiencing it. The exercise of creating the map helps participants consider things from the user's perspective along with his or her goals and challenges.

Example:

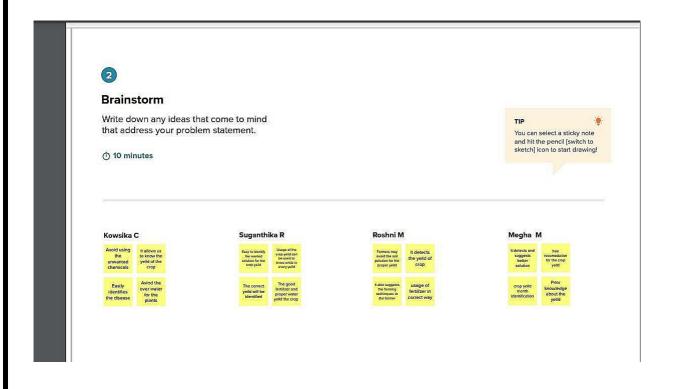


3.2 Ideation & Brainstorming

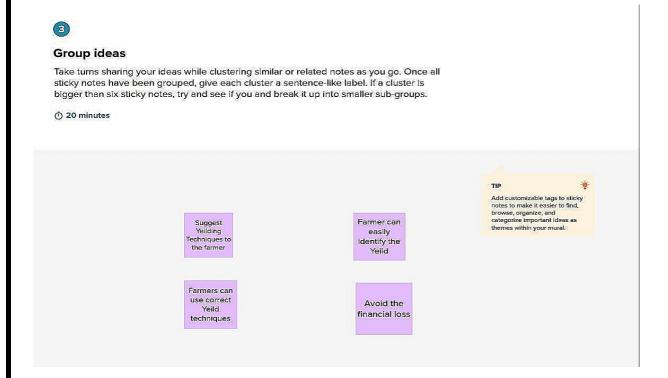
Step 1 : Team Gathering, Collaboration and select the problem statement



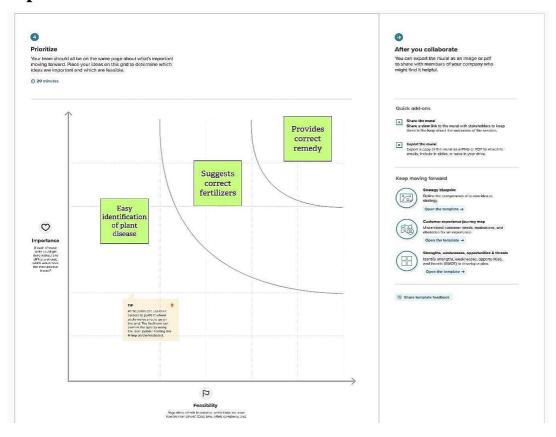
Step 2:Brainstorm, Idea listing and Grouping



Step 3: Group ideas



Step 4:Idea Prioritization



3.3 Proposed Solution

S.No	Parameter	Description
1.	Problem statement	During the covid-19 pandemic, we have
	(problem tobe solved)	faced one of the difficult times of our
		life. Everyone seeks to survive from the
		great disaster. At the time of pandemic,
		no one get to know about which hospital
		has vacant beds (free beds) to admit
		themselves or others infected by covid.
		This situation made the death rate
		higher.
2.	Idea/solution description	Predictive analytics can create patient
		journey dashboards and disease
		trajectories that helps us to know about
		the patient's period of stay. It improves
		effective allocation of beds and other
		resources, treatment delivery, improves
		efficiencies, and so on.
3.	Novelty / uniqueness	Healthcare data frequently resides in
		several locations. The Collected data
		should be stored in central system(like
		centralized storage).Uniqueness of our
		project is that we can able to use data for
		different things such as which medicine is
		more effective and for understanding
		behavioural pattern of particular disease.

4.	Social	effective use of resource, Enhanced
	impact/customer	diagnosis ,Improved Treatment
	satisfaction	enhancing the overall ,quality of
		treatment and life of patients.
5.	Business model	With the gathered data, redirecting the
	(revenue model)	patients to particular hospital based on the
		vacancy, leading retailers used methods
		like market-basket analysis to discover
		insights about consumer purchase
		behaviour and used these insights to
		optimize the physical store experience,
		target relevant ads and streamline the
		supply chain, among other strategic
		initiatives.
6.	Scalability of the solution	A variety of institutions must store,
		evaluate, and take action on the massive
		amounts of data being produced by the
		health care sector as it expands quickly.
		India is a vast, culturally varied nation
		with a sizable population that is
		increasingly able to access centralised
		healthcare services

3.4 Proposed Solution Fit

N

2. CUSTOMER SEGMENT(S)

Patients Hospital Management

6.CUSTOMER STATE

Inadequate information about availability of required resource

5. AVAILABLE SOLUTIONS

Tableau cloud Text Mining Information Retrieval

2. PROBLEMS / PAINS

Effective Resource allocation Reduce Waiting time for patients in Hospitals

9.ROOT / CAUSE

of every problem

No proper system or less efficient Prediction System

7.BEHAVIOR

Tracking the information with the available Technologies

3.TRIGGERS TO ACT

Covid Pandemic Emergency Situations

4.EMOTIONS

BEFORE: Feeling bad & Frustrated AFTER: Feeling better &Relaxed

10. YOUR SOLUTION

Existing: ratio of discharges in given period of time to no. of beds in hospital during the time period Proposing: Using predictive analysis powered by AI

8. CHANNELS OF BEHAVIOR

ONLINE: Use of data from all regions (data Exploration)

OFFLINE: Use of data Collect from nearby facilities

REQUIREMENT ANALYSIS

4.1 Functional requirement .

FR	Functional Requirement (Epic)	Sub Requirement (Story / Sub-
No		Task)
FR-1		Dashboard through User can view the
	Dashboard	details
FR-2		User have different type of view
	User view	that information presented in dashboard.
FR-3	Interoperability	Dashboard helps to share the patient's information interoperable to the hospitals in timely manner
FR-4	Accuracy	The compliance of a dashboard is like to use very interactively in real time by the hospitals.
FR-5	Compliance	These dashboards are clear, intuitive, and customizable and interactive in manner.

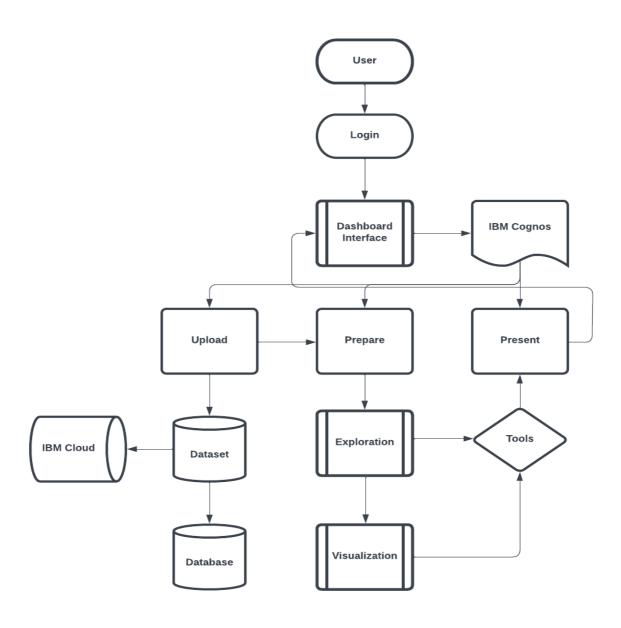
4.2 Non-Functional requirements

FR No.	Non-Functional Requirement	Description
NFR-1	Usability	This Dashboards are designed to offer a comprehensive overview of patient's LOS, and do so through the use of data visualization tools like charts and graphs

NFR-2	Security	The Dashboard helps to indicate the current threatlevel to the Hospitals; an indication of events and incidents that have occurred; a record of authentication errors; unauthorized access
NFR-3	Reliability	This dashboard will be consistent and reliable to the users and helps the user to use in effective, efficientand reliable manner.
NFR-4	Performance	This dashboard can scan the backend users and analyzing the frequency in which they visit the dashboard helps understand how useful and helpful the data displayed is for tasks.
NFR-5	Availability	The dashboard can available to meet user's demand in timely manner and it is also helps to provide necessary information to the user's dataset
NFR-6	Scalability	The layers used in the dashboard are a hosted feature layer, feature layer view, or hosted tile layer.

PROJECT DESIGN

5.1 Data Flow Diagrams



5.2 Solution & Technical Architecture

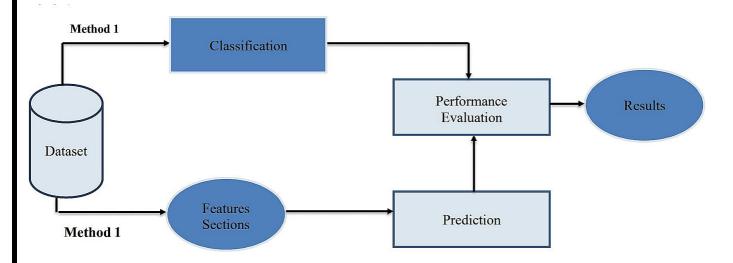


Table:1 Components and Technology

S.No	Component	Description	Technology	
1.	Importing data	Data Import lets you upload data from external sources and combine it with data you collect via Analytics	Python, numpy, pandas.	
2.	Data Cleaning	Data cleaning is a process by which inaccurate, poorly formatted, or otherwise messy data is organized and corrected	Python, numpy, pandas	
3.	Data Preprocessing	Data preprocessing, a component of data preparation, describes any type of processing performed on raw data to prepare it for another data processing procedure	Python, numpy, scipy, pandas	
4.	Training data	Training data is the subset of original data that is used to train the machine learning model,	Numpy, scipy, pandas	
5.	Testing data	Test data is data which has been specifically identified for use in tests, typically of a computer program.	Numpy, scipy, pandas	
6.	Machine learning model	A machine learning model is a file that has been trained to recognize certain types of patterns. You train a model over a set of data, providing it an algorithm that it can use to reason over and learn from those data	Numpy, scipy, pandas, sklearn	
7.	Improve model performance	Accuracy is one metric for evaluating classification models. Informally, accuracy is the fraction of predictions our model got right.	sklearn	
8.	Checking accuracy	A data accuracy check, sometimes called a data sanity check, is a set of quality validations that takeplace before using data.	Sklearn	

5.2 User Stories

User	Functional	User Story	User Story / Task	Priority	Release
Type	Requirement	Number			
	(Epic)				
Customer (Mobile user)	Dashboard	User Number-1	As a user, I access the data in visualize mode	Medium	Sprint-1
		Numbar_7	As a user, I can access the data from Queries, graph ,pie chart	High	Sprint-1

PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning & Estimation

Sprint planning is an event in scrum that kicks off the sprint. The purpose of sprint planning is to define what can be delivered in the sprint and how that work will be achieved. Sprint planning is done in collaboration with the whole scrum team. In scrum, the sprint is a set period of time where all the work is done. However, before you can leap into action you have to set up the sprint. You need to decide on how long the time box is going to be, the sprint goal, and where you're going to start. The sprint planning session kicks off the sprint by setting the agenda and focus. If done correctly, it also creates an environment where the team is motivated, challenged, and can be successful. Bad sprint plans can derail the team by setting unrealistic expectations. The following is the sprint planning and estimation for the project

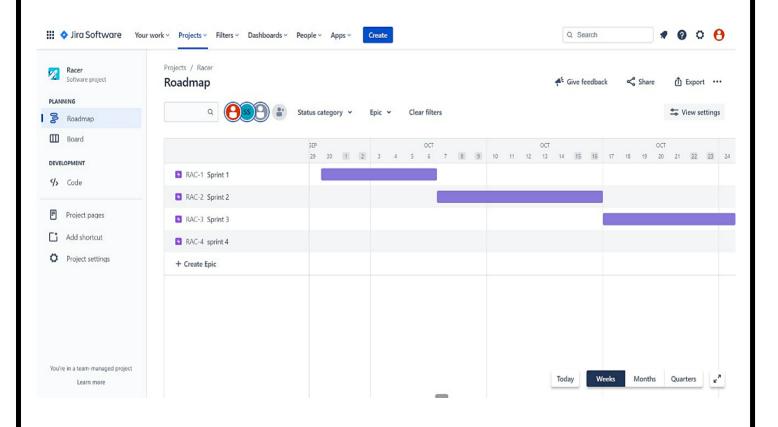
Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Registration	USN-1	As a health care provider I can create account in IBM cloud and the data are collected.	20	High	Sweetha S Swathi S Sruthi V G Sanjika S
Sprint-2	Analyze	USN-2	As a health care provider all the data that are collected is cleaned and uploaded inthe database or IBM cloud.	20	Medium	
Sprint-3	Dashboard	USN-3	As a health care provider I can use my account in my dashboard for uploading dataset.	10	Medium	Sweetha S Sanjika S
Sprint-3	Visualization	USN-4	As a health care provider I can prepare data for Visualization.	10	High	Swathi S Sruthi V G
Sprint-4	Visualization	USN-5	As a health care provider I can present data in my dashboard.	10	High	Sweetha S Sanjika S
Sprint-4	Prediction	USN-6	As a health care provider I can predict the length of stay	10	High	Sweetha S Swathi S

6.2 Sprint Delivery Schedule

Since sprints take place over a fixed period of time, it's critical to avoid wasting time during planning and development. And this is precisely where sprint scheduling enters the equation. In case you're unfamiliar, a sprint schedule is a document that outlines sprint planning from end to end. It's one of the first steps in the agile sprint planning process—and something that requires adequate research, planning, and communication. Sprint Delivery Schedule for the proposed solution.

Sprint	Total Story Points	Duration	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	20	29 Oct 2022
Sprint-2	20	6 Days	31 Oct 2022	05 Nov 2022	20	05 Nov 2022
Sprint-3	20	6 Days	07 Nov 2022	12 Nov 2022	20	12 Nov 2022
Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022	20	19 Nov 2022

6.3 Reports from JIRA



CODING & SOLUTIONING

7.1 Feature

Excel worksheets come with a standard limit of 1,048,576 rows. While performance in Excel will slow well before the said row limit, it's a common requirement for users to analyze datasets over one million rows in size. Congo's Analytics compresses your data so you can extract insights from large datasets. With a well- built data model, Congo's Analytics can help you analyze datasets containing over 100 million rows.

import pandas as pd import numpy as np import matplotlib.pyplot as plt from matplotlib.colors import ListedColormap import seaborn as sns from warningsimport filterwarnings filterwarnings('ignore') pd.options.display.max_columns = None pd.options.display.max_rows = None pd.options.display.float_format = '{:.6f}'.format from sklearn.model_selection import train_test_split import statsmodels import statsmodels.api as sm from sklearn.preprocessing import StandardScaler from sklearn import metrics from sklearn.linear_model import LogisticRegression from sklearn.metrics import classification_report from sklearn.metrics import cohen_kappa_score from sklearn.metrics import confusion_matrix

from sklearn.metrics import roc_curve

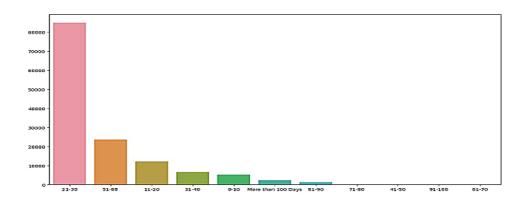
```
from sklearn.metrics import accuracy_score
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn import tree
from sklearn.model_selection import GridSearchCV
from sklearn.ensemble import
AdaBoostClassifier,GradientBoostingClassifier
from catboost import CatBoostClassifier
from sklearn.feature_selection import RFE
plt.rcParams['figure.figsize'] = [15,8]
```

Feature 2

Always the user can't come to the Coding part and comes to get the results because the user don't know about the coding applications so the User builds need interactive website which user can easily catch the results in the application.

```
for i in [ 'Stay', 'Department', 'Available Extra Rooms in Hospital',
'Ward_Type', 'Ward_Facility_Code', 'Age', 'Type of Admission',
'Severity of Illness', 'Bed Grade', 'Hospital_region_code',
'Hospital_type_code', 'City_Code_Hospital', 'Hospital_code',
'City_Code_Patient', 'Visitors with Patient']:
Count = train[i].value_counts()
plt.figure(figsize=(10,5))
sns.barplot(x=count.index.values, y=count.values,data=train)
plt.xlabel(i)
plt.ylabel('No of occurences')
Total = train.isnull().sum().sort_values(ascending=False)
Percent = (train.isnull().sum()*100/train.isnull().count()).sort_values(ascendin g=False)
missing_data = pd.concat([Total, Percent], axis = 1, keys = ['Total', 'Percentage of Missing Values'])
print('X_train', X_train.shape)
print('y_train', y_train.shape)
print('X_test', X_test.shape)
print('y_test', y_test.shape)
X_train (219655, 16)
y_train (219655,)
```

```
X_test (94138, 16)
y_test (94138,)
from sklearn.model_selection import KFold,cross_val_score
kfold=KFold(n_splits=10, shuffle=True, random_state=10)
LR = LogisticRegression()
LR.fit(X_train,y_train
) y_pred_LR=LR.predict(X_test)
accuracy_score(y_test,y_pred_LR)*100 37.94217000573626
rf classification = RandomForestClassifier(random state = 10)
rf_model = rf_classification.fit(X_train, y_train)
y_pred_RF = rf_model.predict(X_test)
accuracy_score(y_test,y_pred_RF)*100 38.315026875438186
print(classification_report(y_test,y_pred_RF))
rf_classification_tuned = RandomForestClassifier(criterion = 'gini',
n_estimators = 47, random_state = 10)
rf_model_tuned = rf_classification_tuned.fit(X_train, y_train)
y_pred_RF_tuned = rf_model_tuned.predict(X_test)
accuracy_score(y_test,y_pred_RF_tuned)*100 37.94323227602031
print(classification_report(y_test,y_pred_RF_tuned)
ls = df_full_test.columns.tolist()
in_data = df_full_test[ls]
out_data = cb.predict (in_data)
test = pd.read_csv("D:/HealthCare/test_data.csv") submit = pd.DataFrame()
submit ['case_id'] = test['case_id'] submit ['Stay'] = out_data stay_decode = { 1 : '0-10', 2 : '11-20', 3 : '21-30', 4 :
'31-40', 5: '41-50', 6: '51-60', 7: '61-70',8: '71-80', 9: '81-90', 10: '91-100', 11: 'More than 100 Days' }
submit ['Stay'] = submit ['Stay'].map(stay_decode)
submit.head(15)
```



8.TESTING

8.1 Test Case

		Date	19/11/2022						
		Team ID	PNT2022TMID16345						
		Project Name	Analytics for Hospitals' Health-Care Da	1		v			100000000000000000000000000000000000000
Test case ID	Feature Type	Test case ID	Feature Type	Test case ID	Expected Result	Actual Result	Statu	Commnets	TC for Automation(Y/N
Dashboard	Functional	LoginPage_TC_004	Functional	oard&pathRef=.public_f olders%2FHealth_Care%2FD epartment%2BPatient&c loseWindowOnLastView=tru	Dashboard A dashboard helps you to monitor events or activites at a glance by providing key insight and analysis about your data on one or more page or screens	Working as expected	Pass	NIL	Y
Story	Verify whether the story is functioned on the analytics dashboard	IBM COGONOS WITH WATSON and IBM cloud	Enter the Analytics Home page Lelick the story option View the ibm embedded story about the Hospital data in html page and after click the back to top button, go to the top of the main page.	.com/bi/?pathRef=.public_fo		Working as expected	Pass	NIL	Y
Report	Verify user is able to view and run the reports	IBM COGONOS WITH WATSON and IBM cloud	Enter the Analytics Home page Lelick the report option View the ibm embedded report about the Hospital data in html page and after click the back to top button, go to the top of the main page.	https://us3.ca.analytics.ibm. com/bi/?perspective=story& amp;pathRef=.public_folders	page,multiple-query report	Working as expected	Pass	NIL	Y

8.2 ACCEPTANCE TESTING

The purpose of this document is to briefly explain the test coverage and open issues of the project at the time of the release to User Acceptance Testing(UAT)

Defect Analysis

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

Resolution	Severity1	Severity2	Severity3	Severity4	Subtotal
By Design	5	5	3	0	13
Duplicate	0	0	0	0	0
External	7	5	1	0	13
Fixed	11	8	7	5	31
Not Reproduced	1	0	0	0	1
Skipped	0	0	0	0	0
Won'tFix	0	0	0	0	0
Totals	24	18	11	5	58

Model Performance Testing

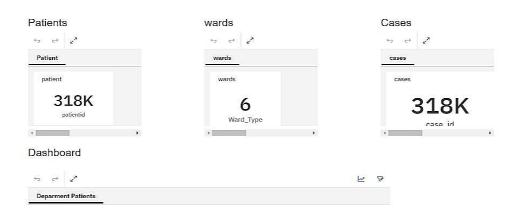
Project team shall fill the following information in model performance testing template.

Parameter	Screenshot / Values
Dashboard design	17/5
Data Responsiveness	The final output from IBM Cognos With Watson further converted into PDF or Story file ,So it can be viewed by all devices.
Amount Data to Rendered (DB2 Metrics)	OKB.
Utilization of Data Filters	The Utilization of data Filters like Ascending ,Descending ,Format and so on.
Effective User Story	12
Descriptive Reports	17/5
	Dashboard design Data Responsiveness Amount Data to Rendered (DB2 Metrics) Utilization of Data Filters Effective User Story

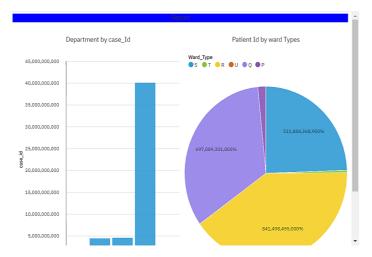
9.RESULTS

9.1 Performance Metrics

Analytics for health care-Data

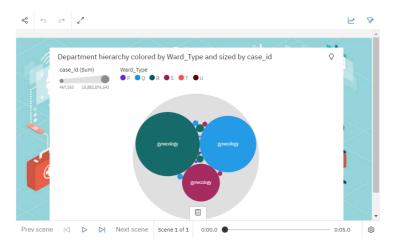


Report



Story

Story



ADVANTAGES & DISADVANTAGES

ADVANTAGES

- By digitizing, combining and effectively using big data, healthcare organizations ranging from single-physician offices and multi-provider groups to large hospital networks and accountable care organizations stand to realize significant benefits.
- Potential benefits include detecting diseases at earlier stages when they can be treated more easily and effectively; managing specific individual and population health and detecting health care fraud more quickly and efficiently.
- Numerous questions can be addressed with big data analytics. Certain developments
 or outcomes may be predicted and/or estimated based on vast amounts of historical
 data, such as length of stay (LOS)
- McKinsey estimates that big data analytics can enable more than \$300 billion in savings per year in U.S. healthcare, two thirds of that through reductions of approximately 8% in national healthcare expenditures.

DISADVANTAGE

- This technology needs to be extensively trained with curated data sets in order to perform as expected. However, the lag between data collection and processing has to be addressed
- Since the healthcare industry is crucial for patient care, the medical community
 need proof that the important managerial issues of ownership, governance and
 standards have to be considered and woven through these issues are those of
 continuous data acquisition and data cleansing
- Invasion of privacy is one of the major concerns of this technology. With companies
 delving into <u>customer</u> <u>data</u> until its very depths, maintaining privacy is next to
 impossible for the general public.

CONCLUSION

In this project, different variables were analyzed that correlate with Length of Stay by using patient-level and hospital-level data. By predicting a patient's length of stay at the time of admission helps hospitals to locate resources more efficiently and manage their patients more effectively. Identifying factors that associate with LOS to predict and manage the number of days patients stay, could help hospitals in managing resources and in the development of new treatment plans. Effective use of hospital resources and reducing the length of stay can reduce overall national medical expenses.

Big data analytics in healthcare is evolving into a promising field for providing insight from very large data sets and improving outcomes while reducing costs. Its potential is great; however there remain challenges to overcome. Big data analytics has the potential to transform the way healthcare providers use sophisticated technologies to gain insight from their clinical and other data repositories and make informed decisions. In the future we'll see the rapid, widespread implementation and use of big data analytics across the healthcare organization and the healthcare industry. To that end, the several challenges highlighted above, must be addressed. As big data analytics becomes more mainstream, issues such as guaranteeing privacy, safeguarding security, establishing standards and governance, and continually improving the tools and technologies will garner attention.

FUTUREN SCOPE

- Smart Staffing & Personnel Management: having a large volume of quality data helps health care professionals in allocating resources efficiently. Healthcare professionals can analyze the outcomes of checkups among individuals in various demographic groups and determine what factors prevent individuals from seekingtreatment.
- **Advanced Risk & Disease Management:** Healthcare institutions can offer accurate, preventive care. Effectively decreasing hospital admissions by digging into insights such as drug type, conditions, and the duration of patient visits, among many others.
- **Real-time Alerting:** Clinical Decision Support (CDS): applications in hospitals analyzes patient evidence on the spot, delivering recommendations to health professionals when they make prescriptive choices. However, to prevent unnecessary in-house procedures, physicians prefer people to stay away from hospitals
- **Enhancing Patient Engagement:** Every step they take, heart rates, sleeping habits, can be tracked for potential patients (who use smart wearables). All this information can be correlated with other trackable data to identify potential healthrisks.

CHAPTER -13 APPENDIX

13.1 SOURCE CODE

```
<!DOCTYPE html>
<html lang="en">
<head>
 <title>Bootstrap Example</title>
 <meta charset="utf-8">
 <meta name="viewport" content="width=device-width, initial-scale=1">
 <link rel="stylesheet"</pre>
href="https://maxcdn.bootstrapcdn.com/bootstrap/3.4.1/css/bootstrap.min.css">
 <script src="https://ajax.googleapis.com/ajax/libs/jquery/3.6.0/jquery.min.js"></script>
 <script
src="https://maxcdn.bootstrapcdn.com/bootstrap/3.4.1/js/bootstrap.min.js"></script>
</head>
<body>
<div class="jumbotron text-center">
 <h1>Analytics for health care-Data</h1>
</div>
<div class="container">
 <div class="row">
  <div class="col-sm-4">
   <h3>Patients</h3>
   <iframe
src="https://us3.ca.analytics.ibm.com/bi/?perspective=dashboard&pathRef=.public_fold
ers%2FHealth_Care%2Fpatients&closeWindowOnLastView=true&ui_appbar=fals
e&ui_navbar=false&shareMode=embedded&action=view&mode=dashb
oard&subView=model000001848c07e3e4_00000002" width="250" height="250"
frameborder="0" gesture="media" allow="encrypted-media" allowfullscreen=""></iframe>
</div>
```

```
<div class="col-sm-4">
   <h3>wards</h3>
   <iframe
src="https://us3.ca.analytics.ibm.com/bi/?perspective=dashboard&pathRef=.public_fold
ers%2FHealth_Care%2Fwards&closeWindowOnLastView=true&ui_appbar=false
&ui_navbar=false&shareMode=embedded&action=view&mode=dashbo
ard&subView=model000001848c040509_00000002" width="250" height="250"
frameborder="0" gesture="media" allow="encrypted-media" allowfullscreen=""></iframe>
</div>
  <div class="col-sm-4">
   <h3>Cases</h3>
   <iframe
src="https://us3.ca.analytics.ibm.com/bi/?perspective=dashboard&pathRef=.public_fold
ers%2FHealth_Care%2Fcases&closeWindowOnLastView=true&ui_appbar=false
&ui_navbar=false&shareMode=embedded&action=view&mode=dashbo
ard&subView=model000001848c019a6e_00000000" width="250" height="250"
frameborder="0" gesture="media" allow="encrypted-media" allowfullscreen=""></iframe>
  </div>
 </div>
 <h3>Dashboard</h3>
 <br>
 >
  <iframe
src="https://us3.ca.analytics.ibm.com/bi/?perspective=dashboard&pathRef=.public_fold
ers%2FHealth Care%2FDepartment%2BPatient&closeWindowOnLastView=true&amp
;ui_appbar=false&ui_navbar=false&shareMode=embedded&action=view&a
mp;mode=dashboard&subView=model000001848b4421a7_00000002" width="800"
height="500" frameborder="0" gesture="media" allow="encrypted-media"
allowfullscreen=""></iframe>
 <br>
 <h3>Report</h3>
 <br/>br>
 >
  <iframe
```

```
src="https://us3.ca.analytics.ibm.com/bi/?pathRef=.public_folders%2FHealth_Care%2FRep
ort%2B1&closeWindowOnLastView=true&ui_appbar=false&ui_navbar=fals
e&shareMode=embedded&action=run&format=HTML&prompt=false"
width="800" height="500" frameborder="0" gesture="media" allow="encrypted-media"
allowfullscreen=""></iframe> 
 <hr>
 <h3>Story</h3>
 <hr>
 >
  <iframe
src="https://us3.ca.analytics.ibm.com/bi/?perspective=story&pathRef=.public_folders%
2FHealth Care%2Fstory&closeWindowOnLastView=true&ui appbar=false&
ui_navbar=false&shareMode=embedded&action=view&sceneId=model0000
01848d38ab03_00000001&sceneTime=0" width="800" height="500" frameborder="0"
gesture="media" allow="encrypted-media" allowfullscreen=""></iframe>
 </div>
</body>
</html>
13.2 GITHUB & PROJECT DEMO LINK
```

GITHUB ID: https://github.com/IBM-EPBL/IBM-Project-18976-1659691797

PROJECT DEMO LINK

 $https://drive.google.com/file/d/16XBVuBT2o82Dt9L7YsC1Jo6RTfitma_i/view?usp=sharing$