

# **Arrhythmia Detection Using Deep Learning With 2-D spectrum image Representation**

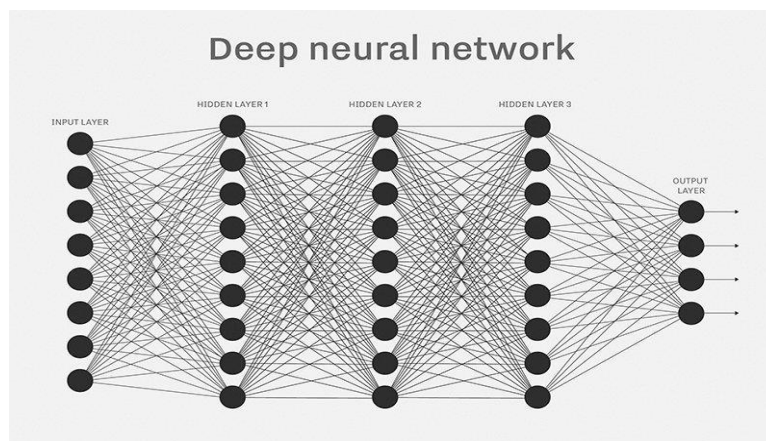
## **1. INTRODUCTION**

### **1.1 Project Overview:**

According to the World Health Organization (WHO), cardiovascular diseases (CVDs) are the number one cause of death today. Over 17.7 million people died from CVDs in the year 2017 all over the world which is about 31% of all deaths, and over 75% of these deaths occur in low and middle-income countries. Arrhythmia is a representative type of CVD that refers to any irregular change from the normal heart rhythms. There are several types of arrhythmia including atrial fibrillation, premature contraction, ventricular fibrillation, and tachycardia. Although a single arrhythmia heartbeat may not have a serious impact on life, continuous arrhythmia beats can result in fatal circumstances. In this project, we build an effective electrocardiogram (ECG) arrhythmia classification method using a convolutional neural network (CNN), in which we classify ECG into seven categories, one being normal and the other six being different types of arrhythmia using deep two-dimensional CNN with grayscale ECG images. We are creating a web application where the user selects the image which is to be classified. The image is fed into the model that is trained and the cited class will be displayed on the webpage.

### **1.2 Purpose**

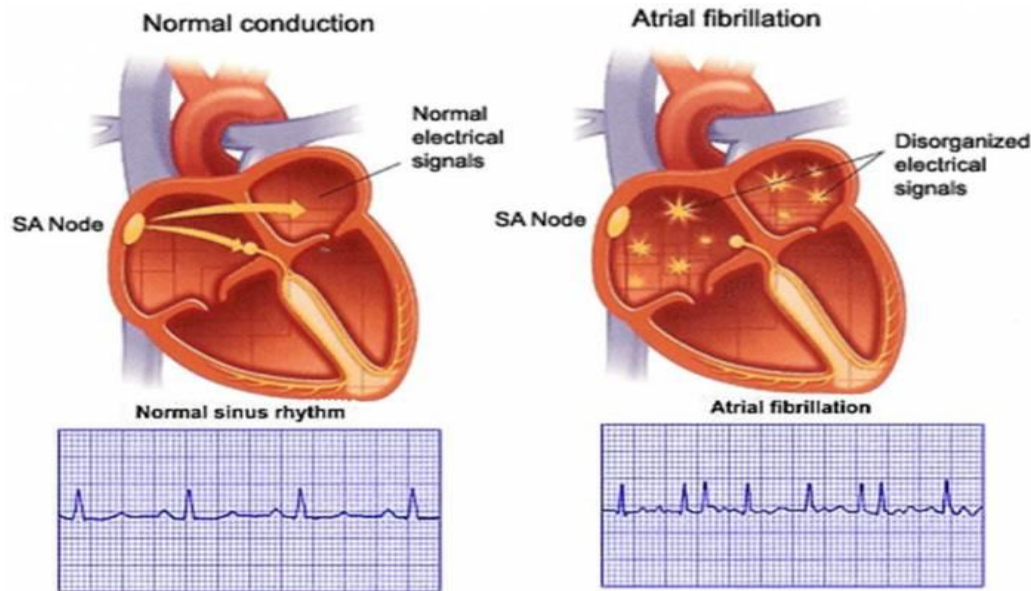
In the past few decades, Deep Learning has proved to be a compelling tool because of its ability to handle large amounts of data. The interest to use hidden layers has surpassed traditional techniques, especially in pattern recognition. One of the most popular deep neural networks is Convolutional Neural Networks.



## **2. LITERATURE SURVEY**

## 2.1 Existing problem

Cardiovascular diseases (CVDs) are the number one cause of death today. Over 17.7 million people died from CVDs in the year 2017 all over the world which is about 31% of all deaths, and over 75% of these deaths occur in low and middle-income countries. Arrhythmia is a representative type of CVD that refers to any irregular change from the normal heart rhythms. There are several types of arrhythmia including atrial fibrillation, premature contraction, ventricular fibrillation, and tachycardia.



## 2.2 References

Topic	Authors	Published Year	Abstract
Classification of Arrhythmia by Using Deep Learning with 2-D ECG Spectral Image Representation	Amin Ullah, Syed Muhammad Anwar, Muhammad Billal, Raja Mehmood	2020	In this study they have proposed a two-dimensional(2D) convolutional neural network (CNN) model for the classification of ECG signals into the eight classes. The eight classes are the normal beat, premature ventricular contraction beat, paced beat, right bundle branch block beat, left bundle branch block beat, atrial premature contraction beat, ventricular flutter wave beat and ventricular escape beat. The proposed CNN classifier was implemented in Python with the open-source library Tensor Flow which was developed by Google for deep learning. This

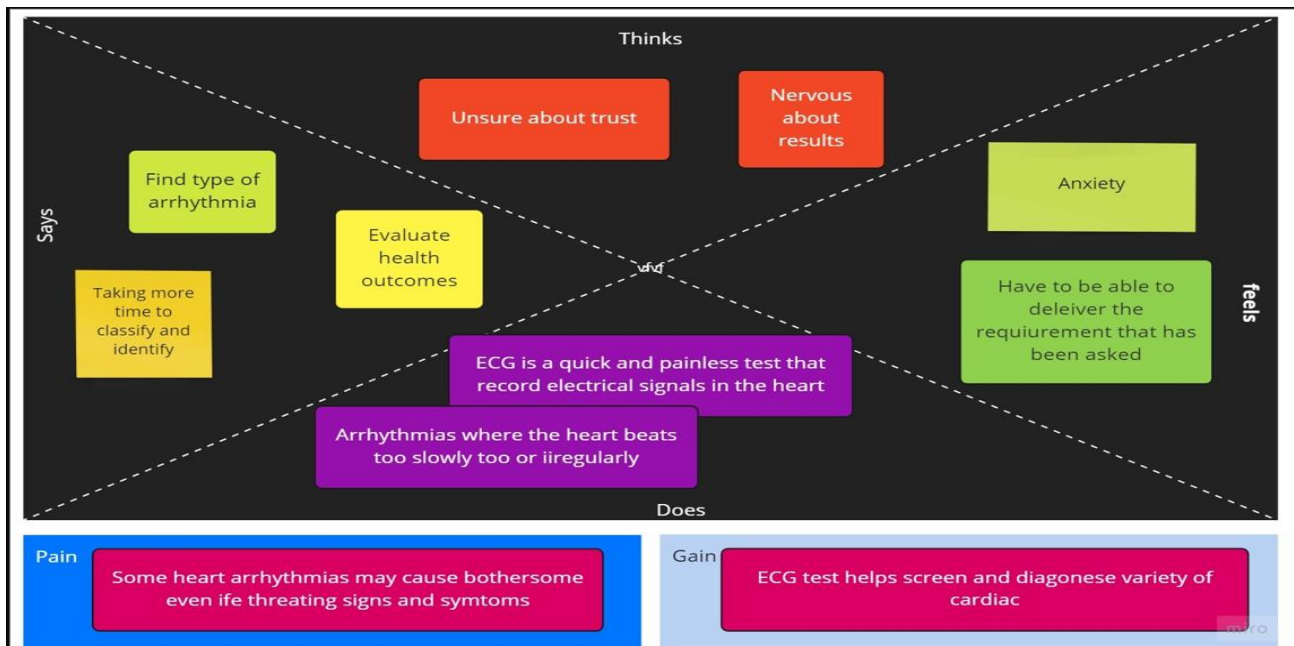
			research paper uses only a single-lead ECG signal.
Cardiac Arrhythmia Detection from 2D ECG Images by Using Deep Learning Technique	E. Izci, M. A. Ozdemir, M. Degirmenci and A. Akan	2019	In this research paper, they proposed a deep learning approach using convolutional neural network for classifying the five different types of arrhythmia disease. They segmented the heartbeats out of the ECG radio signals and all the images are converted into grayscale images. They implemented novel preprocessing, extraction and selection. They achieved an accuracy of 97.52%
Cyberbullying Detection Using Machine Learning	Aaminah Ali, Adeel M.Syed	2020	In this study they used numeric data for training, So the text was first converted into the numerical form using a label encoder. After that, the dataset divided into 80% training set, and 20% test and then classification algorithms were applied. The algorithms were used in classification are SVM, naïve Bayes, Random Forest and then an ensemble approach. The ensemble approach was a hybrid model, In this approach, a soft voting criterion was used. Which predicts the class label utilizing the maximum sum of the predicted probabilities.
Social Media Cyberbullying Detection using Machine Learning	John Hani , Mohamed Nashaat , Mostafa Ahmed	2019	In this research paper, they have proposed an approach to detect cyberbullying using machine learning techniques. And they evaluated their model on two classifiers SVM and neural network and they used TFIDF and sentiment analysis algorithms for features extraction. The dataset has been taken from Kaggle. Thus, a larger cyberbullying data is needed to improve the performance. Hence, deep learning techniques will be suitable in the larger data as they are proven to outperform machine learning approaches over larger size data.

## 2.3 Problem Statement Definition

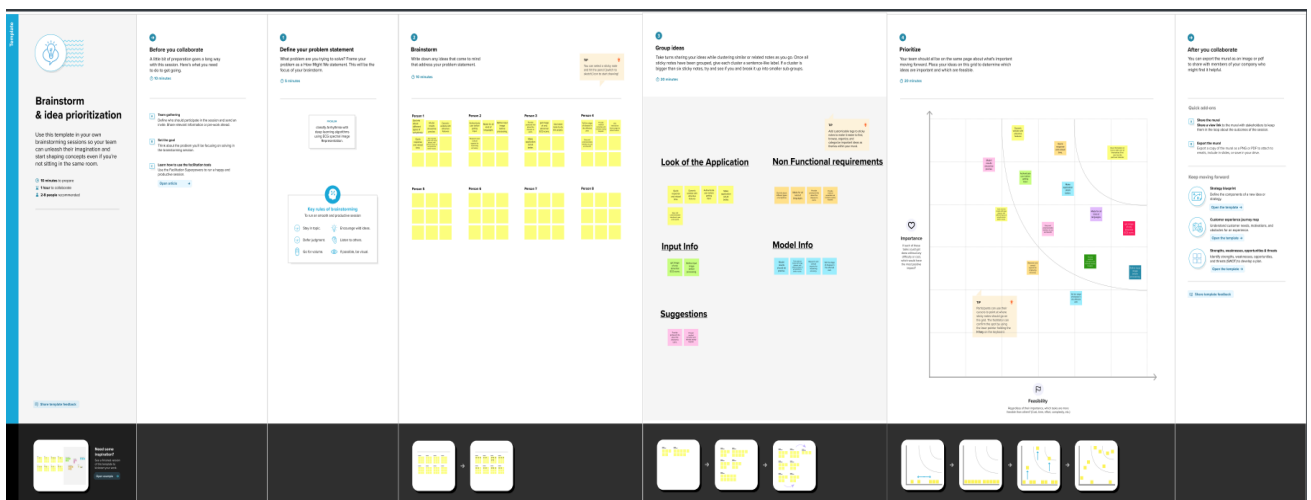
The presented problem wants us to identify and classify the types of Arrhythmias provided in the dataset using spectral images of the Electrocardiogram that are employed for prediction of cardiovascular diseases

## 3. IDEATION & PROPOSED SOLUTION

### 3.1 Empathy Map Canvas



### 3.2 Ideation & Brainstorming



### 3.3 Proposed Solution

S.No.	Parameter	Description
1.	Problem Statement (Problem to be solved)	The presented problem wants us to identify and classify the types of Arrhythmias provided in the dataset using spectral images of the Electrocardiogram that are employed for prediction of cardiovascular diseases.
2.	Idea / Solution description	The given dataset for this problem initially undergoes various data pre-processing steps to identify various forms of noise in the dataset and denoise them to make the data suitable for training a deep learning model. We will employ 2-Dimensional Convolutional Neural Network Model to carry out this classification.
3.	Novelty / Uniqueness	<ul style="list-style-type: none"><li>• Deploys the model to a mobile application by assigning all heavy pre-processing to the cloud through an API.</li><li>• Usage of cloud-based ML-training services.</li><li>• Proposed solution will be deployed in the cloud making it easily accessible over the internet to people across the globe.</li><li>• Provide well detailed instructions or point the user to well-equipped hospitals to get good treatment.</li></ul>

4.	Social Impact / Customer Satisfaction	<ul style="list-style-type: none"> <li>• Promotes Simplicity.</li> <li>• Promotes Self-Diagnosis.</li> <li>• Requires minimal effort and time.</li> <li>• Proposed solution abides by privacy laws and no private information of user is stored.</li> <li>• Delivers highly accurate results(classification of arrhythmia) in a short span of time.</li> </ul>
5.	Business Model (Revenue Model)	<ul style="list-style-type: none"> <li>• Our business model primarily covers the expense we incur by deploying the service in cloud platforms.</li> <li>• Primary consumers of our proposed service are hospitals who seek immediate consultation or use our service as a reference.</li> <li>• Our service can be used by anybody who has access to internet services.</li> <li>• Most of the competing products do not offer a Revenue is generated from Corporate Editions which has a monthly subscription, whereas the Community Edition is free for individuals.</li> <li>• Users who would like to not travel to hospitals to get an ECG or get a selfdiagnosis can rent/buy an ECG Machine through our service which will be the secondary source of income.</li> </ul>
6.	Scalability of the Solution	<ul style="list-style-type: none"> <li>• Increasing the dataset used for model training will in turn increase the application's scalability.</li> <li>• Making the model more reliable will lead to hospitals using this application, this also eliminates human error.</li> <li>• More powerful Cloud Instances for Concurrent use of the application.</li> <li>• Periodically expanding the dataset and updating the model to increase scalability and reliability.</li> </ul>

- Cloud services guarantee high availability so there is very little probability that the service will face serious down times.

### 3.4 Problem Solution fit

Define CS, fit into CC	<b>1. CUSTOMER SEGMENT(S)</b> Who is your customer? i.e. working parents of 0-5 y.o. kids	<b>6. CUSTOMER CONSTRAINTS</b> What constraints prevent your customers from taking action or limit their choices of solutions? i.e. spending power, budget, no cash, network connection, available devices.	<b>5. AVAILABLE SOLUTIONS</b> Which solutions are available to the customers when they face the problem or need to get the job done? What have they tried in the past? What pros & cons do these solutions have? i.e. pen and paper is an alternative to digital notetaking.	Explore AS, differentiate
	1. People who are suffering from arrhythmia 2. Medical professionals 3. Elderly 4. People with severe heart problems	1. Budget constrains 2. Doubt on accuracy 3. Fear on getting started with technology 4. Malfunction of device	1. Direct diagnosis with doctor and waiting for results for much longer time 2. ECG/Holter monitor- portable ECG device 3. Echocardiogram 4. Implantable loop recorder - used in cases of infrequent conditions	
Focus on J&P, tap into BE, understand RC	<b>2. JOBS-TO-BE-DONE / PROBLEMS</b> Which jobs-to-be-done (or problems) do you address for your customers? There could be more than one; explore different sides.	<b>9. PROBLEM ROOT CAUSE</b> What is the real reason that this problem exists? What is the back story behind the need to do this job? i.e. customers have to do it because of the change in regulations.	<b>7. BEHAVIOUR</b> What does your customer do to address the problem and get the job done? i.e. directly related: find the right solar panel installer, calculate usage and benefits; indirectly associated: customers spend free time on volunteering work (i.e. Greenpeace)	Focus on J&P, tap into BE, understand RC
	1. Difficulty in visiting the hospital often 2. Unable to monitor heart asynchronous heart rhythm 3. Inability to know the heart rate	Users are not available at all the time to visit the doctors due to personal commitments, health issues, longer distance between their place and hospitals and unavailability of medical appointments	1. Users expect heart pulse to be perfectly analyzed 2. Accurate prediction of the condition 3. Expectation of accurate results 4. Ease of informing immediate relatives and medical professionals during arrhythmic attacks	
<b>3. TRIGGERS</b> What triggers customers to act? i.e. seeing their neighbour installing solar panels, reading about a more efficient solution in the news.		<b>10. YOUR SOLUTION</b> If you are working on an existing business, write down your current solution first, fill in the canvas, and check how much it fits reality. If you are working on a new business proposition, then keep it blank until you fill in the canvas and come up with a solution that fits within customer limitations, solves a problem and matches customer behaviours.	<b>8. CHANNELS of BEHAVIOUR</b> <b>8.1 ONLINE</b> What kind of actions do customers take online? Extract online channels from #7. <b>8.2 OFFLINE</b> What kind of actions do customers take offline? Extract offline channels from #7 and use them for customer development.	

## 4. REQUIREMENT ANALYSIS

### 4.1 Functional requirement

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)

FR-1	User registration	Registration through website through email for verification.
FR-2	User Confirmation	Verification of the OTP on the website.
FR-3	User data	Uploading the image of ECG used for arrhythmia diagnosis.
FR-4	User diagnosis results (Single and multiple input)	Results of the arrhythmia prediction and classification displayed in the website.
FR-5	User diagnosis results (multiple input)	Results of the arrhythmia prediction and classification sent to registered email.

Following are the functional requirements of the proposed solution.

#### 4.2 Non-Functional requirements

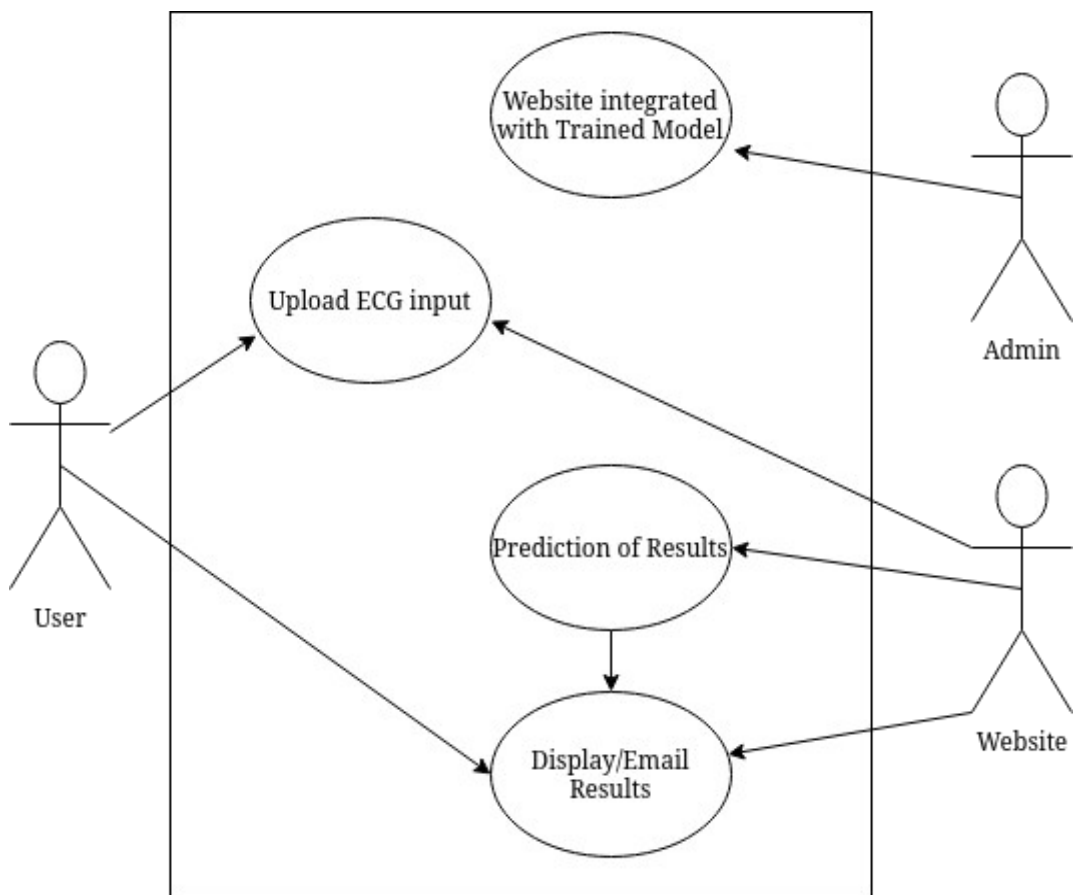
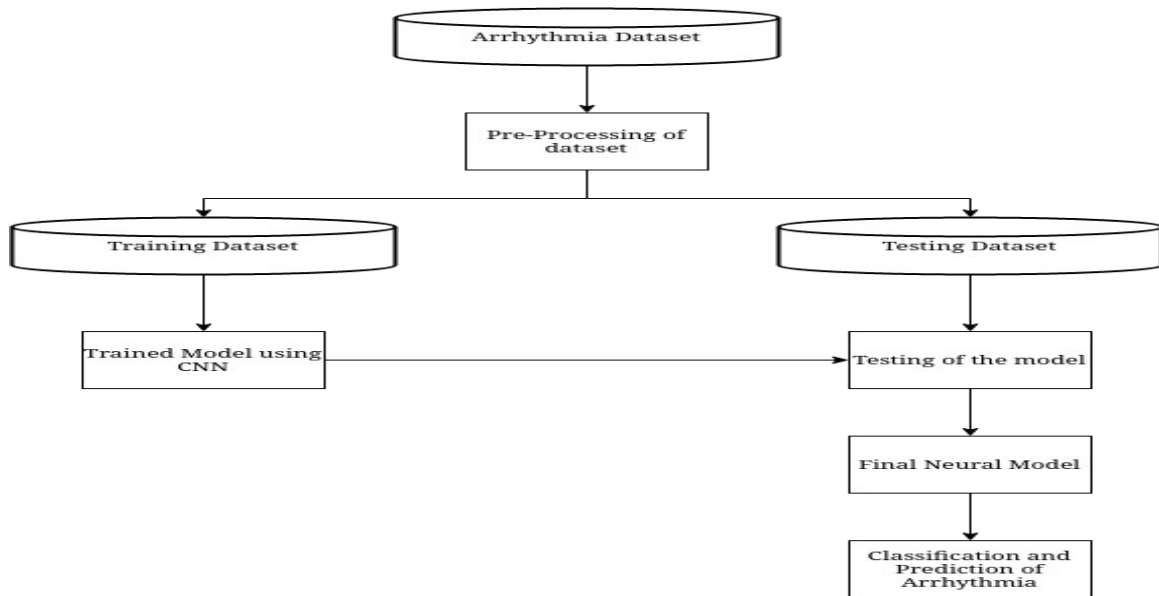
Following are the non-functional requirements of the proposed solution.

FR No.	Non-Functional Requirement	Description
NFR-1	<b>Usability</b>	Usability or ease of use refers to the many factors that constitute what users often describe as user-friendliness.
NFR-2	<b>Security</b>	It assures all data inside the system or its part will be protected against malware attacks or unauthorized access.
NFR-3	<b>Reliability</b>	It is the probability and percentage of the software performing without failure for a specific number of users or amount of time.
NFR-4	<b>Performance</b>	It defines how fast a software system or a particular piece of it responds to certain users actions under a certain workload.
NFR-5	<b>Availability</b>	It describes how likely the system is accessible to a user at a given point in time.
NFR-6	<b>Scalability</b>	It is the ability to handle an increase in workload without performance degradation, or its ability to quickly enlarge.

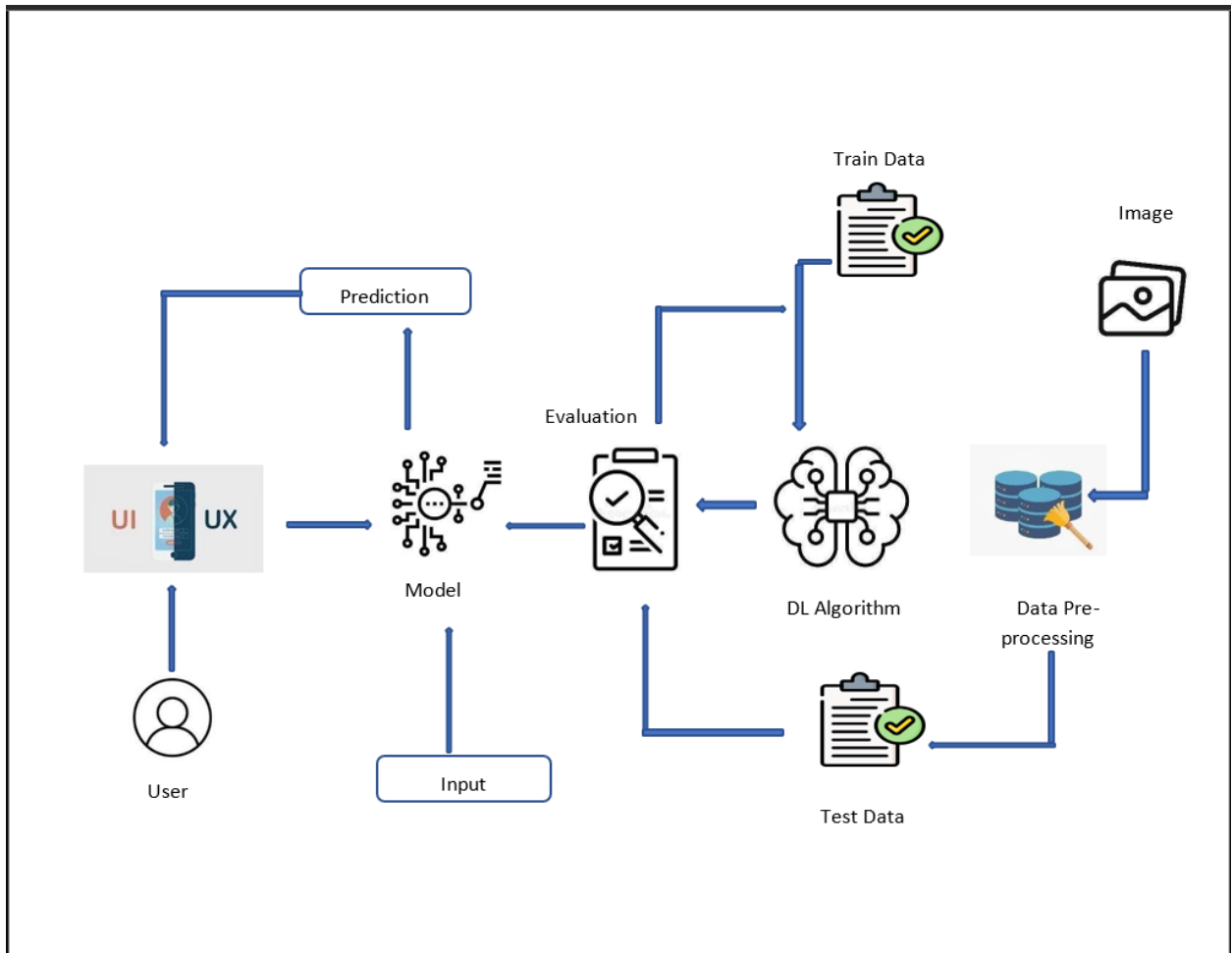


## 5. PROJECT DESIGN

### 5.1 Data Flow Diagrams



## 5.2 Solution & Technical Architecture



## 5.3 User Stories

User Type	Functional Requirement (Epic)	User Story Number	User Story/Task	Acceptance Criteria	Priority	Release
Customer (Web User)	Registration	USN-1	As a user I can register an account on the website	I can access classification page	High	Sprint-1
	Login	USN-2	As a user I can login into the website using credentials	I can access classification page	High	Sprint-1

	Arrhythmia Classification page	USN-3	User uploads the data	Input for the model	High	Sprint-1
	Info page	USN-4	Info page displays information about Arrhythmia	User can learn about Arrhythmia	Medium	Sprint-2

## 6. 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	Download The Dataset	USN-1	We will download the Dataset contains Six classes	2	Low	Harish
Sprint-1	Import The Image Data Generator Library	USN-2	We will import Image Data Generator	2	Low	Harish
Sprint-1	Configure Image Data Generator class	USN-3	We will configure the Image Data Generator class	6	Medium	Harish
Sprint-1	Apply the Image Data Generator functionality to Train Dataset	USN-4	We will apply Image Data Generator to train dataset	10	High	Harish, Manimaran

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-2	Import Libraries	USN-5	We will import required Libraries	1	Low	Harish
Sprint-2	Initialize the Model	USN-6	Initializing the Image recognition model	1	Medium	Harish
Sprint-2	Adding CNN layer	USN-7	We will add Convolutional Neural Network (CNN) used for image/object recognition and classification	4	High	Harish

Sprint-2	Adding Dense Layer	USN-8	We will add Dense Layer in which each neuron receives input from all the neurons of previous layer	4	High	Harish
Sprint-2	Configure The Learning Process	USN-9	We will configure The Learning process which is a method, mathematical logic or algorithm that improves the network's performance and/or training time.	2	Medium	Harish
Sprint-2	Train the Model	USN-10	We will train our model with our image dataset. Fit generator functions used to train a deep learning neural network	4	High	Harish
Sprint-2	Save the Model	USN-11	We will save The model with .h5 extension	2	Medium	Harish
Sprint-2	Test the model	USN-12	We will Test the model through Loaded necessary libraries, the saved model	2	Medium	Harish

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-3	Create Html files	USN-13	We use HTML to create the front end part of the web page.	8	High	Manimaran, Gowthaman
Sprint-3	Build Python code	USN-14	We build the flask file 'app.py' which is a web framework written in python for server-side scripting.	8	High	Bharath Veerakumar
Sprint-3	Run the App	USN-15	We can run the App	4	Medium	Gowthaman
Sprint-4	Register IBM Cloud	USN-16	We can register IBM Cloud	6	Medium	Gowthaman, Manimaran
Sprint-4	Train the model on IBM	USN-17	We can Train Out model on IBM	14	High	Gowthaman, Manimaran

## 6.2 Reports from JIRA



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

### 7.1 Application Building

We created a website using HTML, CSS which helps the user to classify the type of arrhythmia by inputting their ECG image into the website. Flask was also used to render our HTML files and also to integrate our model to the website which helps in classification of the disease.


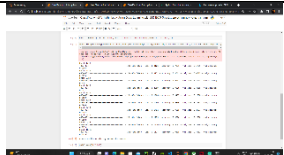
### 7.2 Model Building

Model Building is one of the most important parts of the application which helps to classify the arrhythmia disease. We imported various packages and libraries such as TensorFlow, Pandas, NumPy, etc. for data manipulation and pre-processing. Dataset was taken from Kaggle and we trained and tested using ImageDataGenerator.

We used the Sequential CNN (Convolution Neural Network) model as our classifying model which had 2 hidden layers. Model was made to run a total of 10 epochs for higher accuracy and we attained an accuracy of 92% which is a very high accuracy for disease prediction.

## 8. RESULTS

### 8.1 Performance Metrics

S.No.	Parameter	Values	Screenshot
1.	Model Summary	<pre>Model: "sequential"  __ Layer (type)                 Output Shape # ===== == conv2d (Conv2D)              (None, 62, 62, 32) max_pooling2d (MaxPooling2D) (None, 31, 31, 32)  conv2d_1 (Conv2D)            (None, 29, 29, 32) max_pooling2d_1 (MaxPooling2D) (None, 14, 14, 32)  flatten (Flatten)            (None, 6272) dense (Dense)                (None, 32) dense_1 (Dense)              (None, 6)  ===== == Total params: 211,078 Trainable params: 211,078 Non-trainable params: 0</pre>	
2.	Accuracy	<pre>Training Accuracy - 0.9655  Validation Accuracy - 0.9136</pre>	

## 9. ADVANTAGES & DISADVANTAGES

### ADVANTAGES:

- Very fast and time-efficient
- Easy to access and communicate
- Cost-efficient

### DISADVANTAGES:

- Low chance of error in classification
- No emotion in conveying the result
- Requires good internet access for accessing the website which may not be available in an urgent situation

## **10. CONCLUSION**

Application was built using various web technologies such as HTML, CSS, Flask, etc and model was also build using Machine Learning Techniques such as Data Preprocessing and model building using TensorFlow and other packages to classify arrhythmia accurately and time-efficiently.

## **11. FUTURE SCOPE**

This application can help other classification of diseases much more easier as it is nearly equivalent to this application and can further be researched to improve accuracy and other metrics.

## **12. APPENDIX**

### **Source Code**

Github Link: <https://github.com/IBM-EPBL/IBM-Project-10429-1659179513>

DemoLink:

[https://drive.google.com/file/d/1VCRzynzMdLN1Lgk1sHF32rl27JXqTgzA/view?usp=share\\_link](https://drive.google.com/file/d/1VCRzynzMdLN1Lgk1sHF32rl27JXqTgzA/view?usp=share_link)