

CLASSIFICATION OF ARRHYTHMIA BY USING DEEP LEARNING WITH 2-D ECG SPECTRAL IMAGE REPRESENTATION

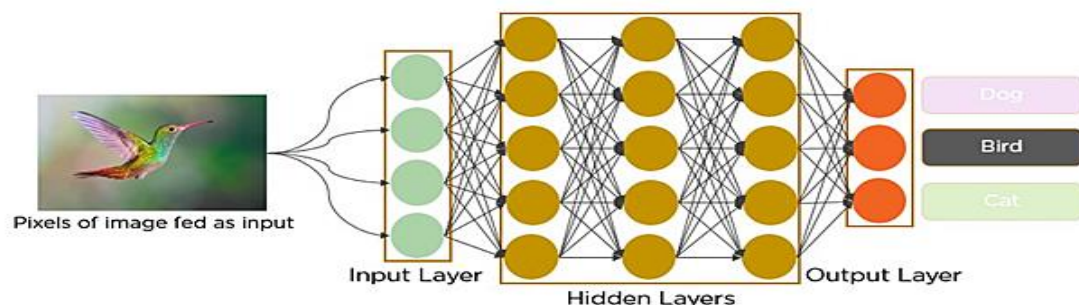
INTRODUCTION:

1.1 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.



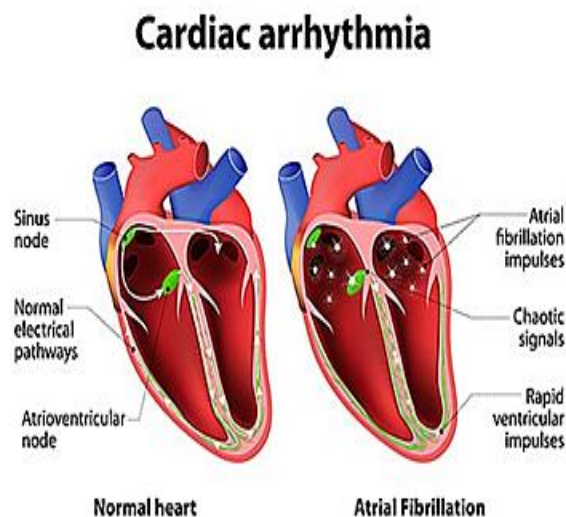
In deep learning, a convolutional neural network (CNN/ConvNet) is a class of deep neural networks, most commonly applied to analyze visual imagery. Now when we think of a neural

network we think about matrix multiplications but that is not the case with ConvNet. It uses a special technique called Convolution. Now in mathematics convolution is a mathematical operation on two functions that produces a third function that expresses how the shape of one is modified by the other.

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 PROPOSED SOLUTION:

An "ambulatory electrocardiogram" or an ECG) about the size of a postcard or digital camera that the patient will be using for 1 to 2 days, or up to 2 weeks. The test measures the movement of electrical signals or waves through the heart. These signals tell the heart to contract (squeeze) and pump blood. The patient will have electrodes taped to your skin. It's painless, although some people have mild skin irritation from the tape used to attach the electrodes to the chest. They can do everything but shower or bathe while wearing the electrodes. After the test period, patient will go back to see your doctor. They will be downloading the information.

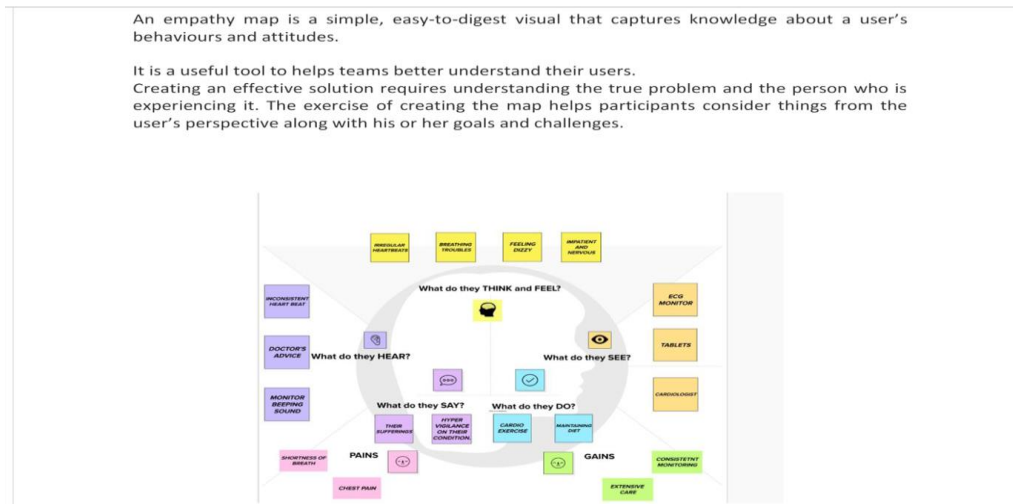
2.3 PROBLEM STATEMENT DEFINITION:

Problem Statement (PS)	I am (Customer)	I'm trying to	But	Because	Which makes me feel
------------------------	-----------------	---------------	-----	---------	---------------------

PS-1	A Lab Technician	Find a software to classify Arrhythmia beats based on ECG Output.	Accuracy in pattern detection is not too high.	Convolutional neural networks has some limitations.	Challenging.
PS-2	A cardiac patient	Find an ECG device to identify arrhythmia beats.	The device is not classifying the beats properly.	It takes lot of time to identify .	Scary.

IDEATION & PROPOSED SOLUTION:

3.1 Empathy Map Canvas:

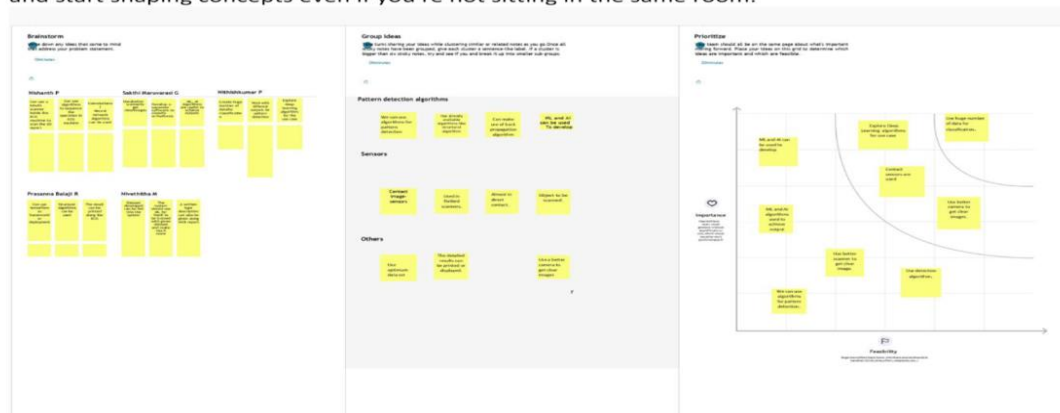


3.2 Ideation & Brainstorming:

Brainstorm & Idea Prioritization Template:

Brainstorming provides a free and open environment that encourages everyone within a team to participate in the creative thinking process that leads to problem solving. Prioritizing volume over value, out-of-the-box ideas are welcome and built upon, and all participants are encouraged to collaborate, helping each other develop a rich amount of creative solutions.

Use this template in your own brainstorming sessions so your team can unleash their imagination and start shaping concepts even if you're not sitting in the same room.



3.3 Proposed Solution:

Proposed Solution Template:

Project team shall fill the following information in proposed solution template.

S.No.	Parameter	Description
1.	Problem Statement (Problem to be solved)	ECG signals is crucial for precise diagnoses of patients' acute and chronic heart conditions. In this study, we propose a two-dimensional (2-D) convolutional neural network (CNN) model for the classification of ECG signals into eight classes; namely, 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 one-dimensional ECG time series signals are transformed into 2-D spectrograms through short-time Fourier transform. The 2-D CNN model consisting of four convolutional layers and four pooling layers is designed for extracting robust features from the input spectrograms. Our proposed methodology is evaluated on a publicly available MIT-BIH arrhythmia dataset. We achieved a state-of-the-art average classification accuracy of 99.11%, which is better than those of recently reported results in classifying similar types of arrhythmias. The performance is significant in other indices as well, including sensitivity and specificity, which indicates the success of the proposed method.
2.	Idea / Solution description	we propose a two-dimensional (2-D) convolutional neural network (CNN) model for the classification of ECG signals into eight classes; namely, 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 one-dimensional ECG time series signals are transformed into 2-D spectrograms through short-time Fourier transform. The 2-D CNN model consisting of four convolutional layers and four pooling layers is designed for extracting robust features from the input spectrograms. Our proposed methodology is evaluated on a publicly available MIT-BIH arrhythmia dataset. We achieved a state-of-the-art average classification accuracy of 99.11%, which is better than those of recently reported results in classifying similar types of arrhythmias. The performance is significant in other indices as well, including sensitivity and specificity, which indicates the success of the proposed method.
3.	Novelty / Uniqueness	We achieved a state-of-the-art average classification accuracy of 99.11%, which is better than those of recently reported results in classifying similar types of arrhythmias. The performance is significant in other indices as well, including sensitivity and specificity, which indicates the success of the proposed method.
4.	Social Impact / Customer Satisfaction	<ol style="list-style-type: none"> 1. Upgradeable Software 2. Works well with Unstructured Data 3. Better Self-Learning Capabilities 4. Supports Parallel and Distributed Algorithms. 5. Cost Effectiveness 6. Low cost maintenance
5.	Business Model (Revenue Model)	<p>Drivers of medical device growth -India</p> <ul style="list-style-type: none"> •Higher disposable incomes •Increase in public spend on healthcare •Increase in penetration of health insurance •Models of healthcare emerging •Many avenues for funding
6.	Scalability of the Solution	In this study, we proposed a 2-D CNN-based classification model for automatic classification of cardiac arrhythmias using ECG signals. An accurate taxonomy of ECG signals is extremely helpful in the prevention and diagnosis of CVDs. Deep CNN has proven useful in enhancing the accuracy of diagnosis algorithms in the fusion of medicine and modern machine learning technologies. Using 2-D images, can classify eight kinds of arrhythmia, namely, NOR, VFW, PVC, VEB, RBB, LBB, PAB, and APC, and it achieved 97.91% average sensitivity, 99.61% specificity, 99.11% average accuracy, and 98.59% positive predictive value (precision). These results indicate that the prediction and classification of arrhythmia with 2-D ECG representation as spectrograms and the CNN model is a reliable operative technique in the diagnosis of CVDs. The proposed scheme can help experts diagnose CVDs by referring to the automated classification of ECG signals. The present research uses only a single-lead ECG signal. The effect of multiple lead ECG data to further improve experimental cases will be studied in future work

3.4 Problem Solution fit:

Problem-Solution fit canvas 2.0 Purpose / Vision

Define CS, fit into CC	1. CUSTOMER SEGMENT(S) CS Who is your customer? i.e. working parents of 0-5 y.o. kids our main target customers are heart specialists(cardiologist),medical labs	6. CUSTOMER CONSTRAINTS CC 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. many cardiologist require vast experience to analyze the ECG reports and to identify the abnormal heartbeat.	5. AVAILABLE SOLUTIONS AS 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. usually experienced cardiologist look into the ECG scan pattern and identify the problem. recently computer aided diagnostics has unraveled a new arena of opportunities. different methods to classify types of arrhythmia using machine learning and deep learning exists. The problem is that these architectures are too deep and they take quite some to train and take up some space as well.	Explore AS, differentiate
	2. JOBS-TO-BE-DONE / PROBLEMS J&P Which jobs-to-be-done (or problems) do you address for your customers? There could be more than one, explore different sides. * classify different types of arrhythmia for diagnosis and treatment * Try to gain insight frm the available ECG data about certain specific characteristics related to the disease and its treatment.	9. PROBLEM ROOT CAUSE RC 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. The reports when analysed manually consumes more time sometimes even false negative outcome is produced. so this may not be helpful for the patient.	7. BEHAVIOUR BE 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) * To refer to experts in their fields * Research to learn and more about different types of arrhythmia	
3. TRIGGERS TR What triggers customers to act? i.e. seeing their neighbour installing solar panels, reading about a more efficient solution in the news. Increasing mortality rates due to untreated arrhythmia.	10. YOUR SOLUTION SL 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 behaviour. To address the problem of misclassification, we intend to use ai to assist different laboratories and doctors with the classification of different major types of arrhythmia. our solution involves the uses of deep learning and feature selection methods that help improve the current classification accuracy obtained by CNNs, and reduce the workload of doctors in diagnosis.	8. CHANNELS of BEHAVIOUR CH 8.1 ONLINE What kind of actions do customers take online? Extract online channels from #7 * To go online and research more about different types of arrhythmia. 8.2 OFFLINE What kind of actions do customers take offline? Extract offline channels from #7 and use them for customer development. * Refer experts in their fields and goes through books and papers to know about different types of arrhythmia patients.	Extract online & offline CH at BE	
4. EMOTIONS: BEFORE / AFTER EM How do customers feel when they face a problem or a job and afterwards? i.e. lost, insecure > confident, in control - use it in your communication strategy & design. * Apprehensive /much more confident * confused /clarified	Problem-Solution fit canvas is licensed under a Creative Commons Attribution-NonCommercial-NoDerivatives 4.0 license Created by Daria Nepriakhina / Amaltama.com			

AMALTAMA

REQUIREMENT ANALYSIS:-

4.1 Functional requirement:

Functional Requirements:

Following are the functional requirements of the proposed solution.

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)
FR-1	User Registration	Registration through gmail Registration through mobile number.
FR-2	User Confirmation	Confirmation via Email Confirmation via OTP.

4.2 Non-Functional requirements:

Non-functional Requirements:

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

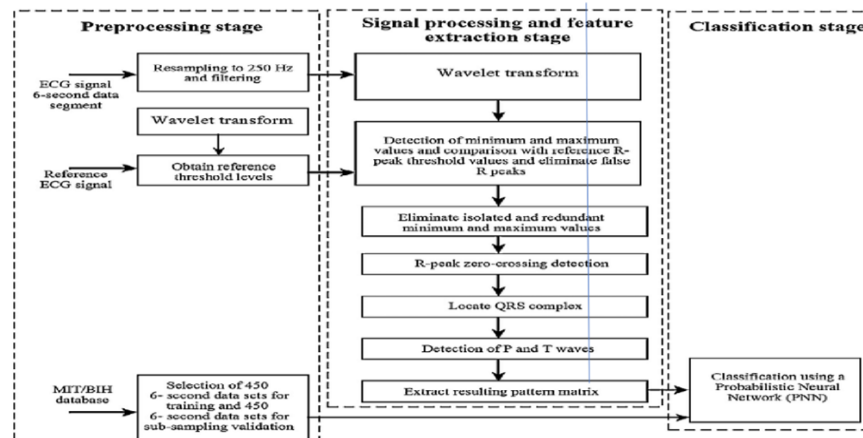
FR No.	Non-Functional Requirement	Description
NFR-1	Usability	Our main target are heart specialists(cardiologist), medical labs. Our software is used by our customers in an easy manner.
NFR-2	Security	The ECG images are encrypted. Our Application is secured using various layers of firewall. Only the Authorized person can able to access the images.
NFR-3	Reliability	We achieved a state-of-the-art average classification accuracy of 99.11%, which is better than those of recently reported results in classifying similar types of arrhythmias.
NFR-4	Performance	The performance is significant in other indices as well, including sensitivity and specificity, which indicates the success of the proposed method.
NFR-5	Availability	Our Software is available 24*7 for registered authentic users. The images are only available to the authorized Medical Specialists.
NFR-6	Scalability	We proposed a 2-D CNN-based classification model for automatic classification of cardiac arrhythmias using ECG signals. An accurate taxonomy of ECG signals is extremely helpful in the prevention and diagnosis of CVDs. Deep CNN has proven useful in enhancing the accuracy of diagnosis algorithms in the fusion of medicine and modern machine learning technologies. Using 2-D images, can classify eight kinds of arrhythmia,

PROJECT DESIGN

5.1 Data Flow Diagrams:

Data Flow Diagrams:

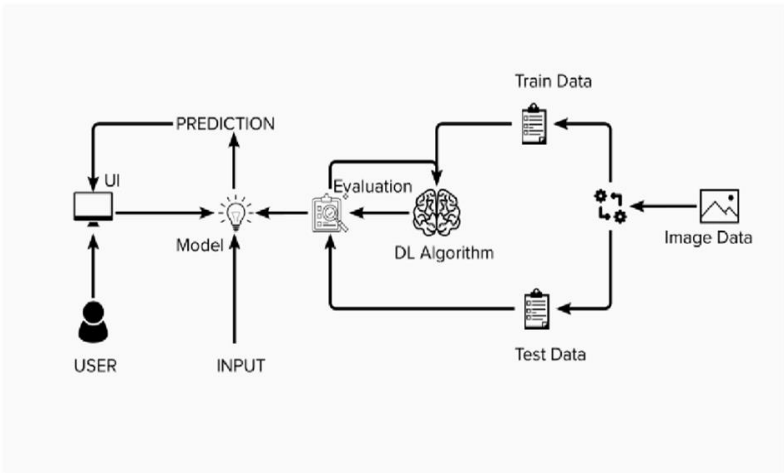
A Data Flow Diagram (DFD) is a traditional visual representation of the information flows within a system. A neat and clear DFD can depict the right amount of the system requirement graphically. It shows how data enters and leaves the system, what changes the information, and where data is stored.



5.2 Solution & Technical Architecture:

Technical Architecture:

The Deliverable shall include the architectural diagram as below and the information as per the table1 & table 2



5.3 User Stories:

User Stories

Use the below template to list all the user stories for the product.

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Individual patient	To know or verify the arrhythmia condition	USN-1	As a user, I can log in to the website using user name and password(credentials) or create one if am new, can proceed with filling out the details as individual and uploading the scanned copy of ECG report	I can access my account / dashboard	High	Sprint-1
		USN-2	As a user, I will receive the report through email or download it on the site itself	I can receive output in pdf format	Low	Sprint-1
		USN-3	As a user, I can see my previous reports of arrhythmia (stages) on my account	I can access it if its a authorised log in	High	Sprint-2
	Login	USN-4	As a user, I can log into the application by entering email & password		High	Sprint-1
Lab Technicians(Hospitals)	To print summary along with ECG if the patient is observed with arrhythmia	USN-5	As a user I need an application to run along with my ECG machine to classify the type of arrhythmia and provide a report with results	It can run along with its operation every time	High	Sprint-3

PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning & Estimation:

S.NO	MILESTONE	ACTIVITIES	DATEs
1.	Preparation Phase	Pre-requisites	1 Nov 2022
		Prior knowledge	1 Nov 2022
		Project Structure	3 Nov 2022

		Project Flow	5 Nov 2022
		Project Objectives	2 Nov 2022
		Registrations	
		Environment Set-up	
2.	Ideation Phase	Literature Survey	7 Oct 2022
		Empathy Map	7 Oct 2022
		Problem Statement	5 Oct 2022
		Ideation	7 Oct 2022
3.	Project Design Phase - I	Proposed Solution	13 Oct 2022
		Problem Solution Fit	13 Oct 2022
		Solution Architecture	13 Oct 2022

4.	Project Design Phase - II	Customer Journey	14 Oct 2022
		Requirement Analysis	14 Oct 2022
		Data Flow Diagrams	19 Oct 2022
		Technology Architecture	19 Oct 2022
5.	Project Planning Phase	Milestones & Tasks	5 Nov 2022
		Sprint Schedules	5 Nov 2022
6.	Project Development Phase	Sprint-1	
		Sprint-2	
		Sprint-3	
		Sprint-4	

6.2 Sprint Delivery Schedule

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	HomePage	USN-1	As a user, I can register for the application by entering my email, password, and confirming my password.	2	High	Prasanna Balaji R
Sprint-1		USN-2	As a user, I will receive confirmation email once I have registered for the application	1	High	Nithishkumar P
Sprint-2	Registrarion	USN-3	As a user, I can register for the application through Facebook	2	Low	Nivethitha M
Sprint-3	Uploading Image	USN-4	As a user, I could upload ECG Image	2	Medium	Sakthi Maruvarasi G
Sprint-4	Result	USN-5	The convolutional neural network identifies and classify arrythmia and gives out the result	1	High	Nishanth P

Project Tracker, Velocity & Burndown Chart: (4 Marks)

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	04 Nov 2022
Sprint-2	20	6 Days	31 Oct 2022	05 Nov 2022	20	08 Nov 2022
Sprint-3	20	6 Days	07 Nov 2022	12 Nov 2022	20	13 Nov 2022
Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022	20	16 Nov 2022

CODING & SOLUTIONING

7.1 Feature 1:

7.1.1 APP.PY:-

```
import os
import numpy as np #used for numerical analysis
from flask import Flask,request,render_template
# Flask-It is our framework which we are going to use to run/serve our application.
```

```

#request-for accessing file which was uploaded by the user on our application.
#render_template- used for rendering the html pages
from tensorflow.keras.models import load_model#to load our trained model
from tensorflow.keras.preprocessing import image

app=Flask(__name__)#our flask app
model=load_model('ECG.h5')#loading the model

@app.route("/") #default route
def about():
    return render_template("home.html")#rendering html page

@app.route("/home") #default route
def home():
    return render_template("home.html")#rendering html page

@app.route("/info") #default route
def information():
    return render_template("info.html")#rendering html page

@app.route("/upload") #default route
def test():
    return render_template("predict.html")#rendering html page

@app.route("/predict",methods=["POST","GET"]) #route for our prediction
def upload():

    if request.method=='POST':

        f=request.files['file'] #requesting the file
        basepath=os.path.dirname('__file__')#storing the file directory
        filepath=os.path.join(basepath,"uploads",f.filename)#storing the file in uploads folder
        f.save(filepath)#saving the file
        print("file save")

        img=image.load_img(filepath,target_size=(64,64)) #load and reshaping the image
        x=image.img_to_array(img)#converting image to array
        x=np.expand_dims(x,axis=0)#changing the dimensions of the image

        ##    pred=model.predict(x)#predicting classes
        ##    y_pred = np.argmax(pred)
        ##    print("prediction",y_pred)#printing the prediction

```

```

preds=model.predict(x)#predicting classes
pred=np.argmax(preds,axis=1)#predicting classes
print("prediction",pred)#printing the prediction

index=['Left Bundle Branch Block','Normal','Premature Atrial Contraction','Premature Ventricular Contractions', 'Right Bundle Branch Block','Ventricular Fibrillation']
##    result=str(index[y_pred])
    result=str(index[pred[0]])

    return result#resturing the result
return None

#port = int(os.getenv("PORT"))
if __name__=="__main__":

    app.run(debug=False)#running our app
    #app.run(host='0.0.0.0', port=8000)

```

7.2 Feature 2:

7.2.1 HOME.HTML:

```

<!DOCTYPE
PE html>

<html>
<head>
    <meta charset="UTF-8">
    <meta http-equiv="X-UA-Compatible" content="IE=edge">
    <meta name="viewport" content="width=device-width, initial-scale=1.0">

    <title>Home</title>
    <style>
        body {

            margin: 0px;
            padding: 0px;
            font-family: sans-serif;
        }

        /* .pd {

```

```
padding-bottom: 100%;  
} */
```

```
.navbar {  
padding: 20px 0px 40px;  
;  
background-color: #222;
```

```
font-size: 25px;  
text-align: center;
```

```
}
```

```
.navbar a {  
color: #eee;  
float: right;  
text-decoration: none;  
font-style: normal;  
font-family: sans-serif;  
padding-right: 10px;  
}
```

```
.navbar a:hover {  
background-color: rgb(0, 0, 0);  
color: rgb(17, 194, 238);  
border-radius: 5px;  
padding: 5px;  
}
```

```
.content{  
background-image: url("https://thumbs.gfycat.com/ChiefHeftyBasil-small.gif");  
background-size: cover;  
background-repeat: no-repeat;  
height: 87vh;  
margin-top: -21px;  
}
```

```
.dic p {  
color: white;  
text-align: center;  
font-family: sans-serif;  
font-size: 30px;  
}
```

```
    footer{
        display: flex;
        justify-content: center;
        background-color: #222;
margin-top: -10px;
color: white;
```

```
    }
</style>
</head>
```

```
<body>
    <div class="navbar">
        <a href="/upload">Predict</a>
        <a href="/info">Info</a>
        <a href="/home">Home</a>

    </div>
```

```
    <div class="content">
```

```
        <h2 style="display: flex;justify-content: center; color:white;size:15;font-family:comic Sans MS">ECG
arrhythmia classification using CNN</h2>
```

```
    <div class="dic">
```

```
        <p>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 single arrhythmia heartbeat may not have a serious impact on life, continuous
            arrhythmia beats can result in fatal circumstances.Electrocardiogram (ECG) is a non-invasive medical
            tool that displays the rhythm and status
            of the heart. Therefore, automatic detection of irregular heart rhythms from ECG signals is a
            significant task in the field of cardiology.
```

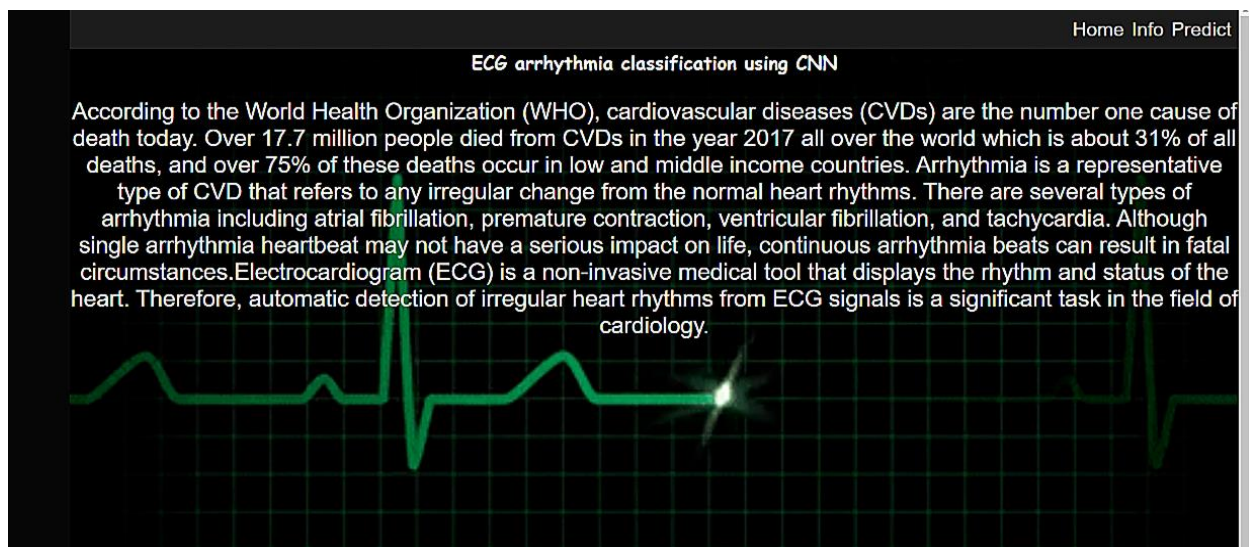
```
        </p>
    </div>
</div>
```



```
<footer>
    <h4>@All Rights Reserved</h4>
</footer>
</body>

</html>
```

OUTPUT:



7.2.2 INFO.HTML:

```
1  <!DOCTYPE html>
2  <html lang="en">
3
4  <head>
5      <meta charset="UTF-8">
6      <meta http-equiv="X-UA-Compatible" content="IE=edge">
7      <meta name="viewport" content="width=device-width, initial-scale=1.0">
8      <title>Info</title>
9
10     <style>
11         body {
12
13             margin: 0px;
14             padding: 0px;
15         }
16
17
18
19         .navbar {
20
21             padding: 20px 0px 40px;
22             ;
23             background-color: #222;
24
25             font-size: 25px;
26             text-align: center;
27
28         }
29     }
30
31     .navbar a {
32         color: #eee;
33         float: right;
34         text-decoration: none;
35         font-style: normal;
36         font-family: sans-serif;
37
```

```
37
38     padding-right: 10px;
39 }
40
41 .navbar a:hover {
42     background-color: rgb(0, 0, 0);
43     color: rgb(17, 194, 238);
44     border-radius: 5px;
45     padding: 5px;
46 }
47
48 .content{
49
50     display: grid;
51     grid-template-columns: 1fr 1fr;
52     justify-content: space-between;
53     gap: 20px;
54     background-color: #222;
55
56 }
57 .content .info{
58     display: grid;
59     grid-template-columns: 1fr 1fr;
60     margin: 10px;
61     background-color: #eee;
62     border-radius: 5px;
63     height: auto;
64 }
65
66 .info img{
67     margin-top: 20px;    }
68     footer{
69         display: flex;
70         justify-content: center;
71         background-color: #222;
72     margin-top: -10px;
73     color: white;
74 }
```

```

74
75
76     }
77
78
79     </style>
80 </head>
81
82 <body>
83     <header class="navbar">
84         <a href="/upload">Predict</a>
85         <a href="/info">Info</a>
86         <a href="/home">Home</a>
87     </header>
88     <div>
89         <h1 style="text-align: center; text-decoration: underline;">ECG</h1>
90         <div class="content">
91             <div class="info">
92                 <span>
93                     <h4>Normal</h4>
94                     If the test is normal, it should show that your heart is beating at an even rate of 60 to 100 beats per minute. Many different heart conditions can show up
95                 </span>
96
97                 
100                 <span>
101                     <h2>Ventricular Fibrillation: </h2> Ventricular fibrillation (VF) is due to multiple wavelet reentrant electrical activity and is manifested on electrocardi
102                 </span>
103                 
104             </div>
105             <div class="info">
106                 <span>
107
108                     <h2>Premature atrial contractions:</h2> Premature atrial contractions (PACs) are extra heartbeats that begin in one of your heart's two upper chambers (atr
109
110

```

```
111         </span>
112         
115         <span>
116             <h3> Premature ventricular contractions: </h3> ventricular contractions (PVCs) are extra heartbeats that begin in one of the heart's two lower pumping chamb
117
118
119         </span>
120         
123         <span>
124             <h3> Right bundle branch block:</h3> Right bundle branch block (RBB) is an electrocardiogram finding resulting in a widened QRS and electrocardiographic vec
125         </span>
126         
130         <span>
131             <h3> Left bundle branch block:</h3> Left bundle branch block (LBBB)occurs when something blocks or disrupts the electrical impulse that causes your heart t
132
133
134         </span>
135         
3
4 <head>
5     <meta charset="UTF-8">
6     <meta http-equiv="X-UA-Compatible" content="IE=edge">
7     <meta name="viewport" content="width=device-width, initial-scale=1.0">
8     <title>Predict</title>
9 </head>
10
11 <style>
12     body {
13
14         margin: 0px;
15         padding: 0px;
16     }
17
18
19
20     .navbar {
21
22         padding: 20px 0px 40px;
23         ;
24         background-color: #222;
25
26         font-size: 25px;
27         text-align: center;
28
29     }
30
31
32     .navbar a {
33         color: #eee;
34         float: right;
35         text-decoration: none;
36         font-style: normal;
37         font-family: sans-serif;
38     }
```



```
38
39     padding-right: 10px;
40 }
41
42 .navbar a:hover {
43     background-color: rgb(0, 0, 0);
44     color: rgb(17, 194, 238);
45     border-radius: 5px;
46     padding: 5px;
47 }
48
49 .content{
50
51     background-image: url("https://www.news-medical.net/image.axd?picture=2021%2F1%2Fshutterstock_1576424071.jpg");
52     height: 90vh;
53     background-repeat: no-repeat;
54     background-size: cover;
55     width: 100%;
56     position: relative;
57 }
58
59
60
61 .wrapper {
62     display: grid;
63     position: absolute;
64     justify-content: center;
65     align-items: center;
66
67     width: 40%;
68     margin: 10% auto;
69     color: #eee;
70     margin-left: 30%;
71
72
73 }
74
75
```

```

75     input[type=button]{
76         padding: 10px;
77         background-color: rgb(170, 89, 89);
78         border: 0;
79         border-radius: 5px;
80         color: white;
81     }
82
83     input[type=button]:hover{
84         padding: 12px;
85         background-color: rgb(7, 7, 7);
86         color: white;
87
88     }
89
90     footer{
91         display: flex;
92         justify-content: center;
93         background-color: #222;
94         margin-top: -10px;
95         color: white;
96
97     }
98 }
99 </style>
100
101 <body>
102     <header class="navbar">
103         <a href="/upload">Predict</a>
104         <a href="/info">Info</a>
105         <a href="/home">Home</a>
106     </header>
107     <div class="content">
108         <div class="wrapper">
109             <form method="post" id="upload-file" enctype="multipart/form-data">
110
111                 <div>

```

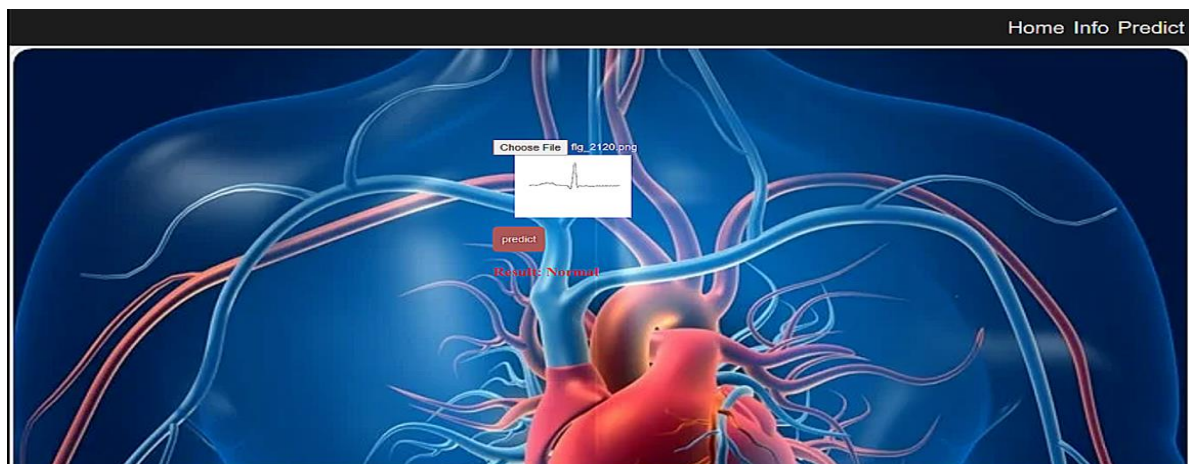
```
111         <div>
112
113             <input type="file" name="file" id="imageUpload" required="true">
114         </div>
115         <div class="image-section" style="display:none;">
116             <div class="holder" style="margin-left: 10%">
117                 
118             </div>
119         </div>
120         <div style="margin-top:10px; ">
121             <input type="button" value="predict" id="btn-predict" >
122         </div>
123
124         <div class="loader" style="display:none;"></div>
125
126         <h3 style="color:rgb(211, 23, 23)" id="result">
127             <span> </span>
128     </h3>
129
130
131     </form>
132 </div>
133 </div>
134
135 <div>
136     <div >
137
138     </div>
139 </div>
140 <footer>
141     <h4>@All Rights Reserved</h4>
142 </footer>
143 <script src="https://ajax.googleapis.com/ajax/libs/jquery/1.9.1/jquery.min.js">
144 </script>
145
146 <script>
147     $(document).ready(() => {
148         $("#imageUpload").change(function () {
```

```

136     <div >
137
138     </div>
139 </div>
140 <footer>
141     <h4>@All Rights Reserved</h4>
142 </footer>
143 <script src="https://ajax.googleapis.com/ajax/libs/jquery/1.9.1/jquery.min.js">
144 </script>
145
146 <script>
147     $(document).ready(() => {
148         $("#imageUpload").change(function () {
149             const file = this.files[0];
150             if (file) {
151                 let reader = new FileReader();
152                 reader.onload = function (event) {
153                     $("#imgPreview")
154                         .attr("src", event.target.result);
155                 };
156                 reader.readAsDataURL(file);
157             }
158         });
159     });
160 </script>
161 <script src="{{ url_for('static', filename='js/main.js') }}" type="text/javascript"></script>
162
163 </body>
164
165 </html>

```

OUTPUT:



TESTING

8.1 Test Cases:

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

Section	Total Cases	Not Tested	Fail	Pas s
Home page	3	0	3	3
Information page	6	1	1	5

Predict page	2	0	0	2
Final Report Output	4	0	0	4
Version Control	2	0	0	2

8.2 User Acceptance Testing:

				Date	15-Nov-22				
				Team ID	PNT2022TMD19938				
				Project Name	Classification of Arrhythmia by Using Deep Lea				
				Maximum Marks	4 marks				
Test case ID	Feature Type	Component	Test Scenario	Pre-Requisite	Steps To Execute	Test Data	Expected Result	Actual Result	Status
Navigation	Functional	Home Page	Validate all the tabs in the navigator		1.Enter URL and click go		All the three tabs should be visible	Working as	pass
Home	Functional	Home page	Verify the visibility of the video		1.Enter URL and click go		User should able to view the video	Working as	Pass
Home pageTC_002	Functional	Home Page	Validate the description of the image		1.Enter URL and click go		Description should be visible on	Working as	Pass
Home pageTC_003	Functional	Home Page	Verify the user is able to navigate to		1.Enter the URL and click go		It should redirect the user to the	Working as	pass
Information page T	Functional	Introductio	Verify the user is in the introduction		1.Enter the URL and click go		User should be in the introduction	Working as	pass
Information	Functional	Introductio	Verify the page title and information		1. Enter the URL and click go		User should able to view the	Working	pass
Predict page	Functional	predict	Verify the working of predict page		1.Enter the URL and click go		User should be able to visit the	Working	pass
Predict page	Functional	predict	Verify the upload image option		1.Enter the URL and click go		Make sure the option works	Working	Pass
Predict	Functional	predict	Verify the choose button is enabled		1.Enter the URL and click go		The choose button option should	Working as	Pass
Predict	Functional	predict	Verify the user is able to access		1.Enter the URL and click go		Image should be uploaded	Working as	pass
Predict	Functional	predict	Verify the selected image is same		1.Enter the URL and click go		Selected image should be an ECG 2D	Working as	pass
Predict	Functional	predict	Verify the working condition of the		1.Enter the URL and click go		The type of arrhythmia should be	Working as	pass

FLOW CHART & RESULTS WITH SCREENSHOTS:

9.1 FLOW CHART & RESULTS BY TRAINING MODEL IN LOCAL MACHINE:-

A. DATASET COLLECTION:

The dataset contains six classes:

1. Left Bundle Branch Block
2. Normal
3. Premature Atrial Contraction
4. Premature Ventricular Contractions
5. Right Bundle Branch Block
6. Ventricular Fibrillation

B. IMAGE PREPROCESSING:

Image Pre-processing includes the following main tasks

- **Import ImageDataGenerator Library:**

Image data augmentation is a technique that can be used to artificially expand the size of a training dataset by creating modified versions of images in the dataset.

The Keras deep learning neural network library provides the capability to fit models using image data augmentation via the ImageDataGenerator class.

```
In [3]: from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Dense
        from tensorflow.keras.layers import Convolution2D
        from tensorflow.keras.layers import MaxPooling2D
        from tensorflow.keras.layers import Flatten
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

- **Configure ImageDataGenerator Class:**

There are five main types of data augmentation techniques for image data; specifically:

1. Image shifts via the width_shift_range and height_shift_range arguments.
2. Image flips via the horizontal_flip and vertical_flip arguments.
3. Image rotates via the rotation_range argument
4. Image brightness via the brightness_range argument.
5. Image zooms via the zoom_range argument.

```
In [4]: train_datagen = ImageDataGenerator(rescale = 1./255, shear_range = 0.2, zoom_range = 0.2, horizontal_flip = True)
```



```
In [5]: test_datagen = ImageDataGenerator(rescale = 1./255)
```

- **Applying ImageDataGenerator functionality to the trainset and test set:**

We will apply ImageDataGenerator functionality to Trainset and Testset by using the following code

This function will return batches of images from the subdirectories Left Bundle Branch Block, Normal, Premature Atrial Contraction, Premature Ventricular Contractions, Right Bundle Branch Block and Ventricular Fibrillation, together with labels 0 to 5{'Left Bundle Branch Block': 0, 'Normal': 1, 'Premature Atrial Contraction': 2, 'Premature Ventricular Contractions': 3, 'Right Bundle Branch Block': 4, 'Ventricular Fibrillation': 5}

```
In [6]: x_train = train_datagen.flow_from_directory("/content/data/train",target_size = (64,64),batch_size = 32,class_mode = "categorical")
```

Found 15341 images belonging to 6 classes.

```
In [7]: x_test = test_datagen.flow_from_directory("/content/data/test",target_size = (64,64),batch_size = 32,class_mode = "categorical")
```

Found 6825 images belonging to 6 classes.

```
In [8]: x_train.class_indices
```

```
Out[8]: {'Left Bundle Branch Block': 0,
        'Normal': 1,
        'Premature Atrial Contraction': 2,
        'Premature Ventricular Contractions': 3,
        'Right Bundle Branch Block': 4,
        'Ventricular Fibrillation': 5}
```

We can see that for training there are 15341 images belonging to 6 classes and for testing there are 6825 images belonging to 6 classes.

C. MODEL BUILDING

We are ready with the augmented and pre-processed image data, we will begin our build our model by following the below steps:

- **Import the model building Libraries:**

```
In [9]: from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Dense, Convolution2D, MaxPooling2D, Flatten
```

- **Initializing the model:**

Keras has 2 ways to define a neural network:

1. Sequential
2. Function API

The Sequential class is used to define linear initializations of network layers which then, collectively, constitute a model. In our example below, we will use the Sequential constructor to create a model, which will then have layers added to it using the add () method. Now, will initialize our model.

- **Adding CNN Layers:**

We are adding a convolution layer with an activation function as “relu” and with a small filter size (3,3) and a number of filters as (32) followed by a max-pooling layer. The Max pool layer is used to downsample the input. The flatten layer flattens the input.

Initialize the model

```
In [10]: model=Sequential()
```

Adding CNN layers

```
In [11]: model.add(Convolution2D(32,(3,3),activation="relu",strides=(1,1),input_shape=(64,64,3)))
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Flatten())
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 62, 62, 32)	896
max_pooling2d (MaxPooling2D)	(None, 31, 31, 32)	0
flatten (Flatten)	(None, 30752)	0
=====		
Total params: 896		
Trainable params: 896		
Non-trainable params: 0		

- **Adding Hidden Layers:**

Dense layer is deeply connected neural network layer. It is most common and frequently used layer.

```
In [12]: model.add(Dense(500,activation="relu"))
model.add(Dense(500,activation="relu"))
```

- **Adding Output Layer:**

```
In [13]: model.add(Dense(6,activation="softmax"))
```

Understanding the model is very important phase to properly use it for training and prediction purposes. Keras provides a simple method, summary to get the full information about the model and its layers.

- **Configure the Learning Process:**

- The compilation is the final step in creating a model. Once the compilation is done, we can move on to the training phase. The loss function is used to find error or deviation in the learning process. Keras requires loss function during the model compilation process.
- Optimization is an important process that optimizes the input weights by comparing the prediction and the loss function. Here we are using adam optimizer

- Metrics is used to evaluate the performance of your model. It is similar to loss function, but not used in the training process.

```
In [14]: model.compile(loss="categorical_crossentropy",optimizer="adam",metrics=['accuracy'])
          len(x_train)
```

- **Training the model:**

We will train our model with our image dataset. `fit_generator` functions used to train a deep learning neural network.

```
In [15]: model.fit(x_train,epochs=5,steps_per_epoch=len(x_train),validation_data=x_test,validation_steps=len(x_test))

Epoch 1/5
480/480 [=====] - 140s 289ms/step - loss: 1.0137 - accuracy: 0.6721 - val_loss: 0.6701 - val_accuracy: 0.7538
Epoch 2/5
480/480 [=====] - 142s 296ms/step - loss: 0.3082 - accuracy: 0.9022 - val_loss: 0.5529 - val_accuracy: 0.8101
Epoch 3/5
480/480 [=====] - 140s 291ms/step - loss: 0.1732 - accuracy: 0.9475 - val_loss: 0.4951 - val_accuracy: 0.8379
Epoch 4/5
480/480 [=====] - 137s 286ms/step - loss: 0.1226 - accuracy: 0.9627 - val_loss: 0.5920 - val_accuracy: 0.8503
Epoch 5/5
480/480 [=====] - 139s 290ms/step - loss: 0.0892 - accuracy: 0.9727 - val_loss: 0.4156 - val_accuracy: 0.8752
```

- **Saving the model:**

The model is saved with .h5 extension as follows

An H5 file is a data file saved in the Hierarchical Data Format (HDF).It contains multidimensional arrays of scientific data.

```
In [16]: model.save('ECG.h5')
```

- **Testing the model:**

Load necessary libraries and load the saved model using `load_model`

Taking an image as input and checking the results

Note: The target size should for the image that is should be the same as the target size that you have used for training.

```
In [17]: import numpy as np
from tensorflow.keras.models import load_model
from tensorflow.keras.preprocessing import image
model=load_model('ECG.h5')
img=image.load_img("/content/data/test/Right Bundle Branch Block/fig_101.png",target_size=(64,64))
img
```

Out[17]: 

```
In [18]: x=image.img_to_array(img)
x
```

```
In [22]: pred=model.predict(x)

1/1 [=====] - 0s 79ms/step
```

```
In [23]: pred
```

```
Out[23]: array([[0., 0., 0., 0., 1., 0.]], dtype=float32)
```

```
In [24]: index=['Left Bundle Branch Block',
'Normal',
'Premature Atrial Contraction',
'Premature Ventricular Contractions',
'Right Bundle Branch Block',
'Ventricular Fibrillation']
index[np.argmax(pred)]
```

```
Out[24]: 'Right Bundle Branch Block'
```

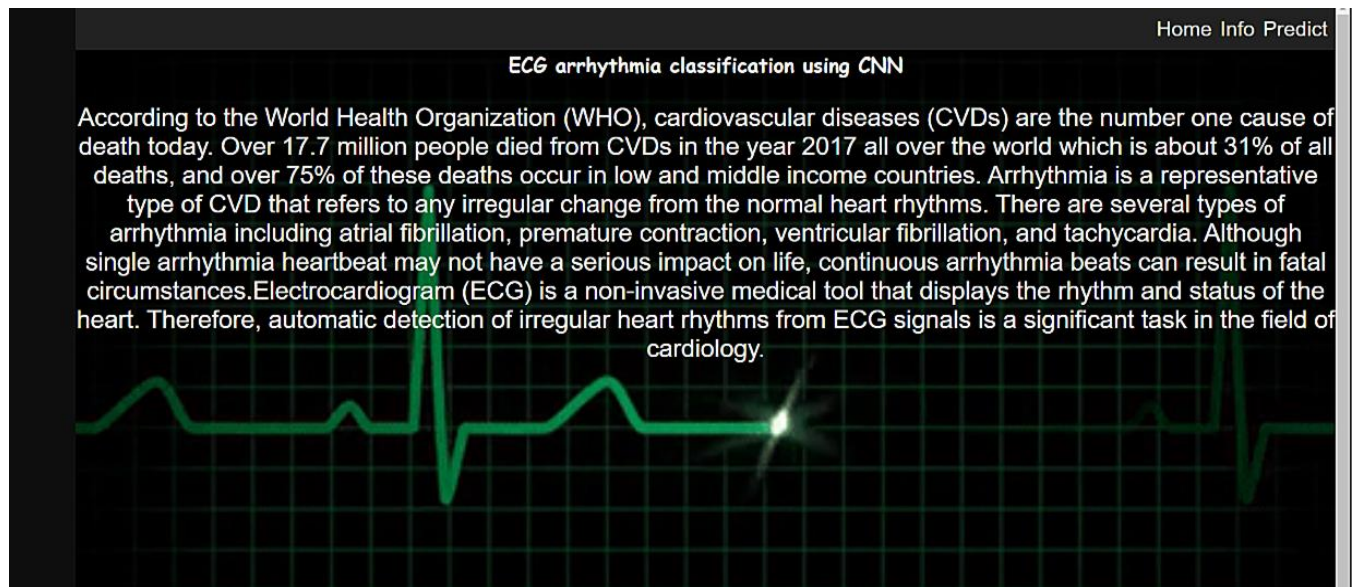
D. APPLICATION BUILDING:

In this section, we will be building a web application that is integrated into the model we built. A UI is provided for the uses where he has uploaded an image. The uploaded image is given to the saved model and prediction is showcased on the UI.

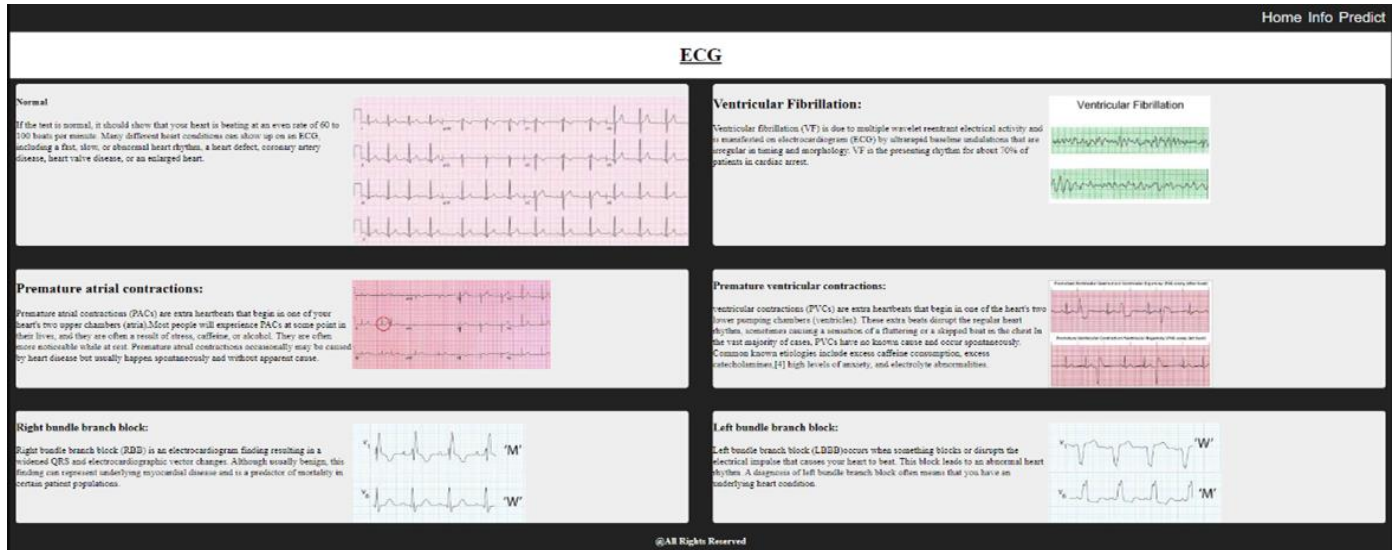
This section has the following tasks

- **Building HTML Pages:**
 - We use HTML to create the front end part of the web page.
 - Here, we created 4 html pages- home.html, predict_base.html, predict.html, information.

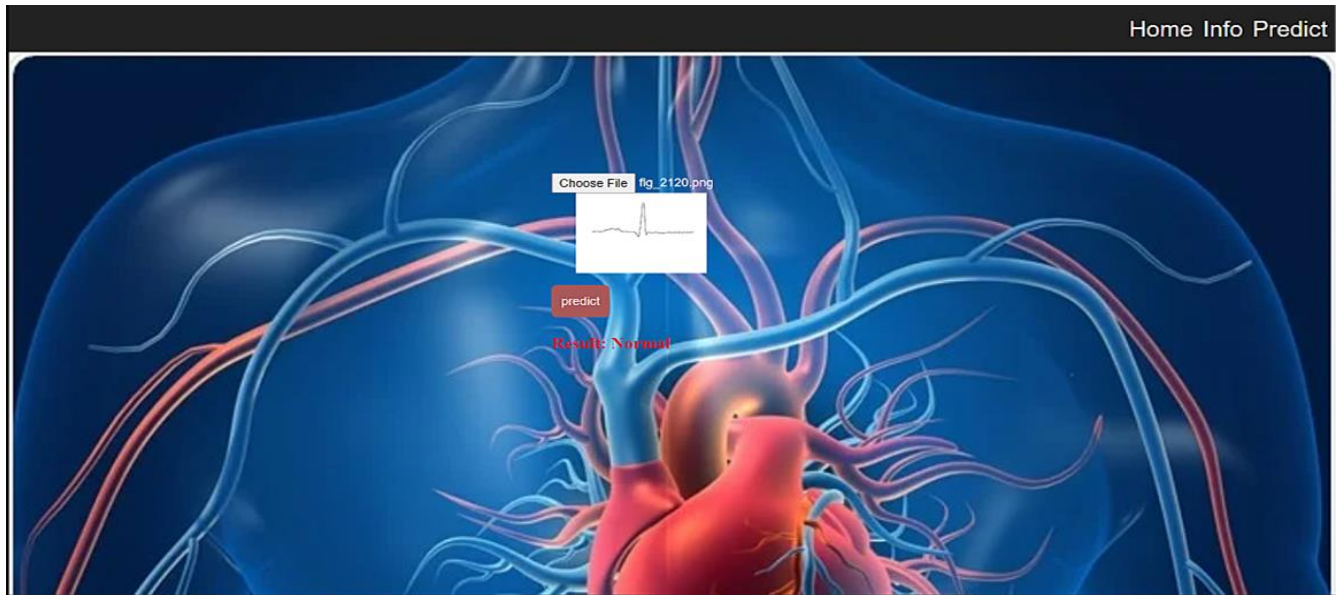
- [home.html](#) displays the home page.



- [information.html](#) displays all important details to be known about ECG.



- Predict-base.html and predict.html accept input from the user and predicts the values.



- **Building server-side script:**

We will build the flask file ‘app.py’ which is a web framework written in python for server-side scripting.

- The app starts running when the “__name__” constructor is called in main.
- render_template is used to return HTML file.
- “GET” method is used to take input from the user.
- “POST” method is used to display the output to the user.

```
Microsoft Windows [Version 10.0.19044.2130]
(c) Microsoft Corporation. All rights reserved.

C:\Users\lap\Desktop>Flask\python -m flask run
2022-11-15 16:50:44.652196: I tensorflow/stream_executor/platform/default/dso_loader.cc:64] Could not load dynamic library 'cudart64_110.dll'; dlerror: cudart64_110.dll not found
2022-11-15 16:50:44.652339: I tensorflow/stream_executor/cuda/cudart_stub.cc:29] Ignore above cudart dlerror if you do not have a GPU set up on your machine.
2022-11-15 16:51:20.441380: W tensorflow/stream_executor/platform/default/dso_loader.cc:64] Could not load dynamic library 'nvcuda.dll'; dlerror: nvcuda.dll not found
2022-11-15 16:51:20.441897: W tensorflow/stream_executor/cuda/cuda_driver.cc:263] failed call to cuInit: UNKNOWN ERROR (303)
2022-11-15 16:51:20.610442: I tensorflow/stream_executor/cuda/cuda_diagnostics.cc:169] retrieving CUDA diagnostic information for host: DESKTOP-4RDSMH6
2022-11-15 16:51:20.611759: I tensorflow/stream_executor/cuda/cuda_diagnostics.cc:176] hostname: DESKTOP-4RDSMH6
2022-11-15 16:51:20.664627: I tensorflow/core/platform/cpu_feature_guard.cc:193] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations: AVX AVX2
To enable them in other operations, rebuild tensorflow with the appropriate compiler flags.
* Debug mode: off
WARNING: This is a development server. Do not use it in a production deployment. Use a production WSGI server instead.
* Running on http://127.0.0.1:5000
Press CTRL+C to quit
127.0.0.1 - - [15/Nov/2022 16:51:48] "GET / HTTP/1.1" 200 -
127.0.0.1 - - [15/Nov/2022 16:52:14] "GET /info HTTP/1.1" 200 -
127.0.0.1 - - [15/Nov/2022 16:52:22] "GET /upload HTTP/1.1" 200 -
127.0.0.1 - - [15/Nov/2022 16:52:22] "GET /upload HTTP/1.1" 200 -
127.0.0.1 - - [15/Nov/2022 16:52:22] "GET /static/js/main.js HTTP/1.1" 304 -
File save
1/1 [=====] - 4s 4s/step
1/1 [=====] - 4s 4s/step
prediction prediction [0]
127.0.0.1 - - [15/Nov/2022 16:53:15] "POST /predict HTTP/1.1" 200 -
127.0.0.1 - - [15/Nov/2022 16:53:15] "POST /predict HTTP/1.1" 200 -
File save
1/1 [=====] - 0s 59ms/step
prediction [1]
127.0.0.1 - - [15/Nov/2022 16:53:43] "POST /predict HTTP/1.1" 200 -
```

- **Running The App:**

```
C:\Users\lap\Desktop\Flask>python -m flask run
```

```
To enable them in other operations, rebuild tensorflow with the appropriate compiler flags.  
* Debug mode: off  
WARNING: This is a development server. Do not use it in a production deployment. Use a production WSGI server instead.  
* Running on http://127.0.0.1:5000
```

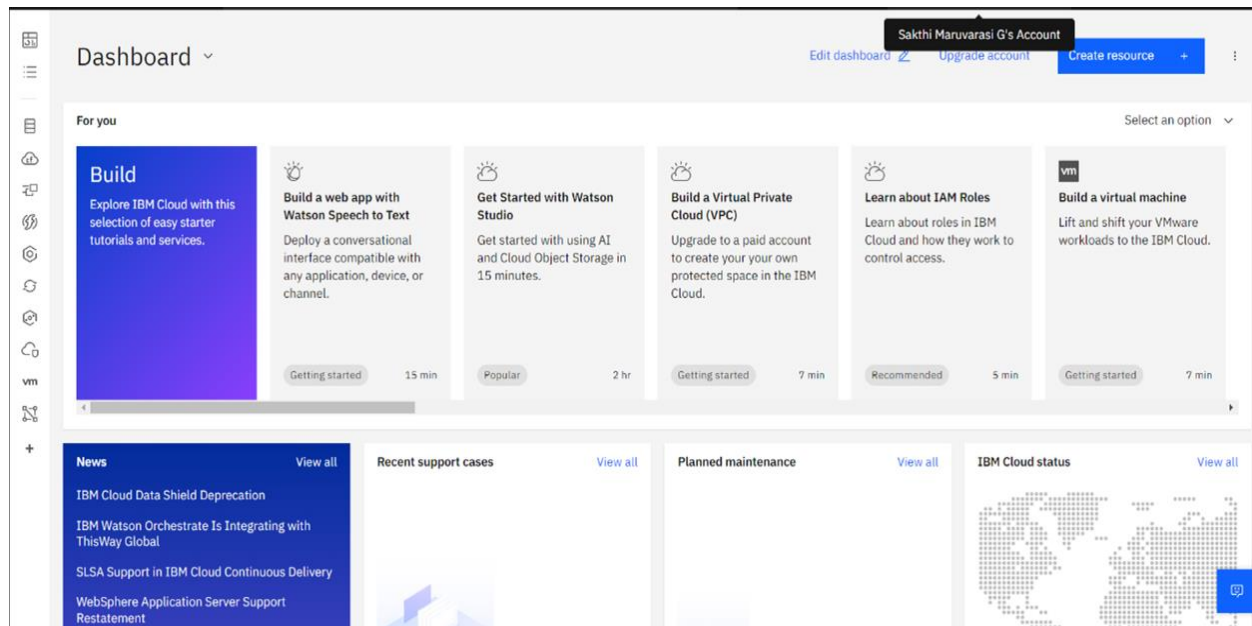
Navigate to the localhost (<http://127.0.0.1:5000/>) where you can view your web page.

9.2 Flow Chart & Results by training model in IBM WATSON STUDIO:

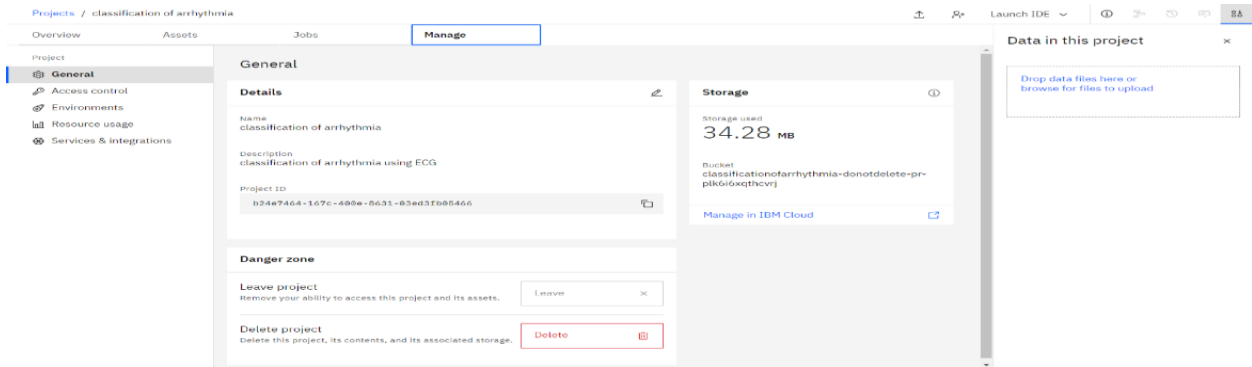
A. Creating IBM cloud account:-

We have to create an IBM Cloud Account and should log in.

B. Register for IBM Cloud account:-



C. Deployment space in the watson studio:-



D. Apply CNN algorithm and save the model and deploy it using API key generated:-

```
In [120... model.save('ECG.h5')
```

```
In [121... !tar -zcvf ECG-model_new.tgz ECG.h5
```

ECG.h5

```
In [134... ls -l
```

```
data/  
ECG.h5  
ECG-model_new.tgz  
my_model1.tar1.gz
```

```
In [95]: from ibm_watson_machine_learning import APIClient
wml_credentials = {
    "url": "https://us-south.ml.cloud.ibm.com",
    "apikey": "m6NbiyMRA29yQcTssg5Wi0HkkZHyK-gib4hSGE3ppyVy"
}
client = APIClient(wml_credentials)
```

```
In [124... client = APIClient(wml_credentials)
```

```
In [125... client.spaces.list()
```

Note: 'limit' is not provided. Only first 50 records will be displayed if the number of records exceed 50

ID	NAME	CREATED
d68afc27-adb2-4919-a9d0-cd6acb8700f3	Model building	2022-11-11T06:11:54.400Z

```
In [126... def guid_from_space_name(client, space_name):
    space = client.spaces.get_details()
    #print(space)
    return(next(item for item in space['resources'] if item['entity']['name'] == space_name)['metadata']['id'])
```

```
In [127... space_uid = guid_from_space_name(client, 'Model building')
print("Space UID = " + space_uid)
```

Space UID = d68afc27-adb2-4919-a9d0-cd6acb8700f3

```
In [128... client.set.default_space(space_uid)
```

Out[128... 'SUCCESS'

```
In [130... software_spec_uid = client.software_specifications.get_uid_by_name("tensorflow_rt22.1-py3.9")
software_spec_uid
```

Out[130... 'acd9c798-6974-5d2f-a657-ce06e986df4d'

```
In [131... model_details = client.repository.store_model(model='ECG-model_new.tgz', meta_props={
    client.repository.ModelMetaNames.NAME: "ECG_Model",
    client.repository.ModelMetaNames.TYPE: "tensorflow_2.7",
    client.repository.ModelMetaNames.SOFTWARE_SPEC_UID: software_spec_uid})
model_id=client.repository.get_model_uid(model_details)
```

This method is deprecated, please use get_model_id()

/opt/conda/envs/Python-3.9/lib/python3.9/site-packages/ibm_watson_machine_learning/repository.py:1453: UserWarning: This method is deprecated, please use get_model_id()
warn("This method is deprecated, please use get_model_id()")

```
In [132... model_id
```

Out[132... '7bc41052-6999-4af0-ba61-aebca5eaa72e'

```
In [133... client.repository.download(model_id, 'my_model1.tar1.gz')
```

E. For downloading the model we have to run the last part of the above code in the local jupyter notebook:

```
In [133... client.repository.download(model_id, 'my_model1.tar.gz')
```

Hence we trained the model using IBM Watson.

ADVANTAGES & DISADVANTAGES:

6.1 ADVANTAGES:

- i. The proposed model predicts Arrhythmia in images with a high accuracy rate of nearly 96%
- ii. The early detection of Arrhythmia gives better understanding of disease causes, initiates therapeutic interventions and enables developing appropriate treatments.

6.2 DISADVANTAGES:

- i. Not useful for identifying the different stages of Arrhythmia disease.
- ii. Not useful in monitoring motor symptoms

APPLICATIONS :

- It is useful for identifying the arrhythmia disease at an early stage.
- It is useful in detecting cardiovascular disorders

CONCLUSION:

- Cardiovascular disease is a major health problem in today's world. The early diagnosis of cardiac arrhythmia highly relies on the ECG.
- Unfortunately, the expert level of medical resources is rare, visually identify the ECG signal is challenging and time-consuming.
- The advantages of the proposed CNN network have been put to evidence.
- It is endowed with an ability to effectively process the non-filtered dataset with its potential anti-noise features. Besides that, ten-fold cross-validation is implemented in this work to further demonstrate the robustness of the network.

FUTURE SCOPE:

For future work, it would be interesting to explore the use of optimization techniques to find a feasible design and solution. The limitation of our study is that we have yet to apply any optimization techniques to optimize the model parameters and we believe that with the implementation of the optimization, it will be able to further elevate the performance of the proposed solution to the next level.

REFERENCES:

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- <https://www.analyticsvidhya.com/blog/2021/05/convolutional-neural-networks-cnn/>
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