Project On

Intelligent Vechile Damage Assessment & Cost Estimator for Insurance Companies

powered By IBM India

Submitted By

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Project ID: PNT2022TMIDO7174

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1. INTRODUCTION

1.1. Project Overview

The project "Intelligent Vechile Damage Assessment and Cost Estimator for Insurance Companies" is a responsive web application powered by aritifical Intelligence and IBM Watson Cloud. Deep Learning model is trained with the various damaged car images in various views and the VGG16 from the TensorFlow library is used for the better Deep Learning model architecture. An attractive front end can be developed using HTML and CSS. The pages such as Index.html , login.html, logout.html, register.html and prediction.html are created and embedded with the IBM cloud databse using python framework called flask. The web application takes the image input and estimate the cost for the insurance companies based on the damages in the car.

1.2. Purpose

The project is based on the domain of Artificial Intelligence and powered by the IBM watson cloud. A responsive web application can be developed using the HTML and CSS which is connected to waston cloud. In the cloud, a database service by availing the service Instance of the IBM cloud and the database API key is collected and connected with the front-end using flash which is an python framework for desiging the backend. Pages such as index.html, login.html, logout.html and prediction.html are used to interact with the web application. The user can register and the data of the user is saved in the databse of the IBM cloud, during the time of login, the login ID is compared with the ID in the databse and allow the user to the next page. The Deep Learning model is build using the VGG16 which is present in the keras library and the model is trained with the images of mulitple car with various level cum types of damages. The model is deployed in the back-end using the flask and the prediction.html page is setted to collect the image from the user. The prediction algorithm is used treat the image and estimated the cost for the user. The project is based on the various components which helps to handle the back - end and Front - end. Then front - end is build using html and css which is connectedback - end which is build using the python and IBM cloud. The project is powered by the IBM Watson cloud and is based in the artificial intelligence field. With the use of HTML and CSS and the Waston Cloud, a responsive web application may be created. The database API key is gathered and connected with the front-end using flash, which is a python framework for designing the backend, in the cloud when a database service is used. To communicate with the web application, utilise pages like prediction.html, login.html, and logout.html.

2. LITERATURE SURVEY

2.1. Existing Problem

The problem is defined as the optimzed way to estimate insurance cost based on the mulitple damages in the various areas in the vechile for the insurance companies. As the existing methods for estimating the cost takes lot of time and energy in the way of inspecting the vechile.

2.2. References

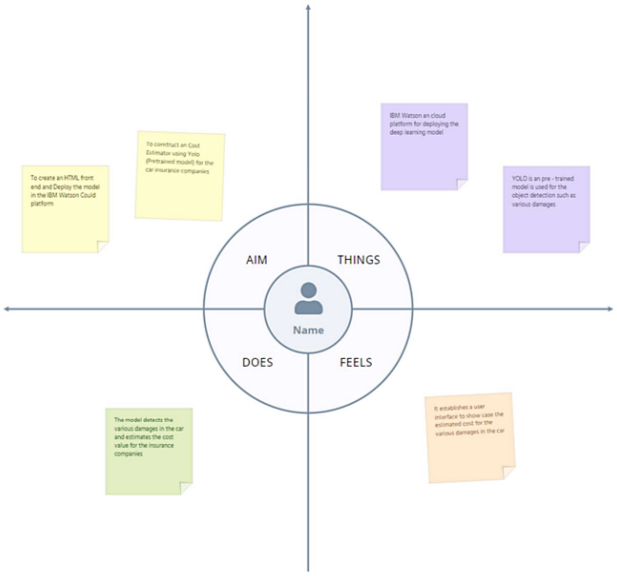
At present, under the guidance of the new generation of information technology, the rapid accumulation of data, the continuous improvement of computing power, the continuous optimization of algorithm models, and the rapid rise of multi-scene applications have made profound changes in the development environment of artificial intelligence. In this paper, based on the demand of automobile insurance claims and intelligent transportation, combined with abundant basic data and advanced machine vision algorithm, an intelligent damage determination system of 'Artificial Intelligence + Vehicle Insurance' is constructed.

S.No	Paper Name	Year	Author Name
1.	Research on Intelligent Vehicle Damage Assessment System Based on Computer Vision	2020	Zhu Qianqian, Guo Weiming, Shen Ying and Zhao Zihao
2.	Car Damage Assessment Based on VGG Models	2021	Phyu Mar Kyu, Kuntpong Woraratpanya
3.	Assessing Car Damage with Convolutional Neural Networks	2021	Harit Bandi

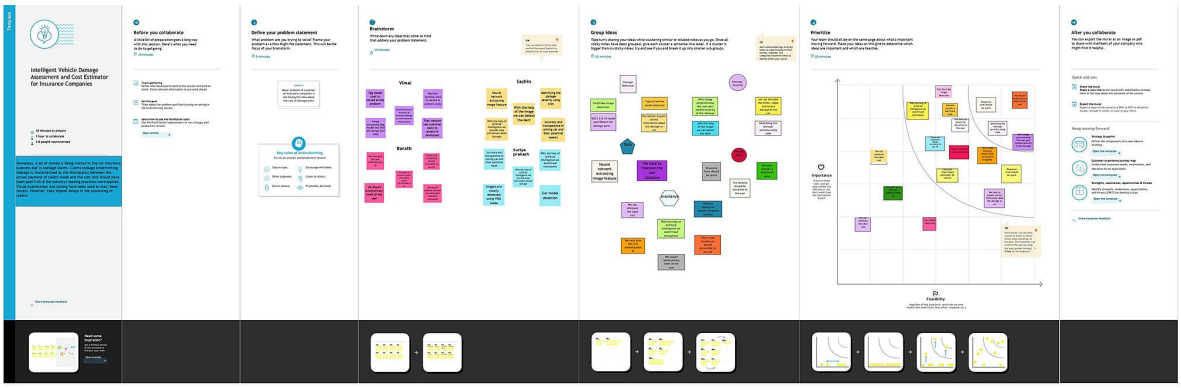
2.3. Problem Statement Defination

3. IDEATION & PROPOSED SOLUTION

3.1. Empathy Map



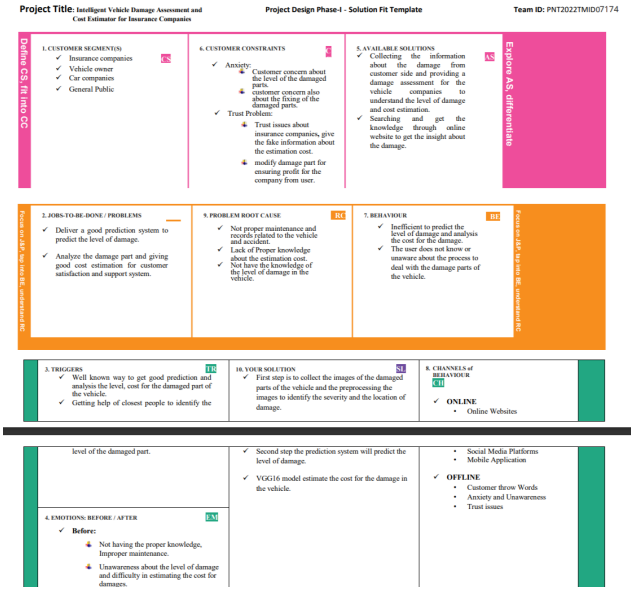
3.2. Ideation and Brainstorming



3.3. Proposed Solution

S.No.	Parameter	Description
1.	Problem Statement (Problem to be solved)	Insurance firms frequently suffer losses because they are unable to accurately estimate the cost of damaged automobiles and they are unable to calculate the cost of damaged cars precisely, insurance companies regularly incur losses.
2.	Idea / Solution description	We are creating an AI model to sense and detect the precise amount of automobiles damage.
3.	Novelty / Uniqueness	Automated calculator for the cost of filing an insurance claim.
4.	Social Impact / Customer Satisfaction	Determining the extent of vehicle damage and offering insurance accordingly.
5.	Business Model (Revenue Model)	Underwriting and investment income are the main sources of income for insurers. Financial investments, including listed shares, government bonds, commercial real estate, and corporate bonds, make up the majority of insurance firms' assets. By estimating the level of car damage using our AI model and providing insurance in accordingly, they are able to save more money and invest it in their own businesses.
6.	Scalability of the Solution	Our artificial intelligence (AI) has the capacity to operate at the scale, speed, and complexity required for the aim. Our model's accuracy will improve with more testing and training using real-time data.

3.4. Proposed Solution Fit



4. REQUIREMENT ANALYSIS

4.1. Functional Requirement

The functional Requirements of this projects invloves the better understanding of Deep Learning, Image Pre-processing, Application designing using HTML & CSS and IBM Waston Cloud. TensorFlow provides the deep learning architecture for learning the feature maps, Image generator module is used in generating the mulitple images based on the shape and agumentation mentioned in the passing parameters. IBM Watson provides the services such as Database, deployment etc.

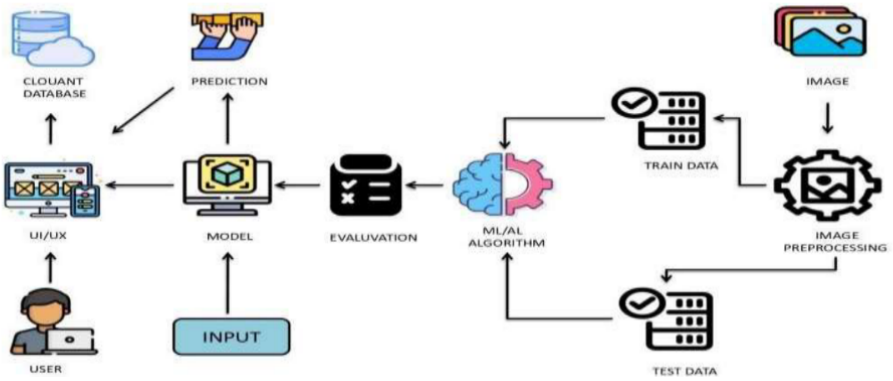
4.2. Non - Functional Requirements

The Non - Functional Requirements of this project are,

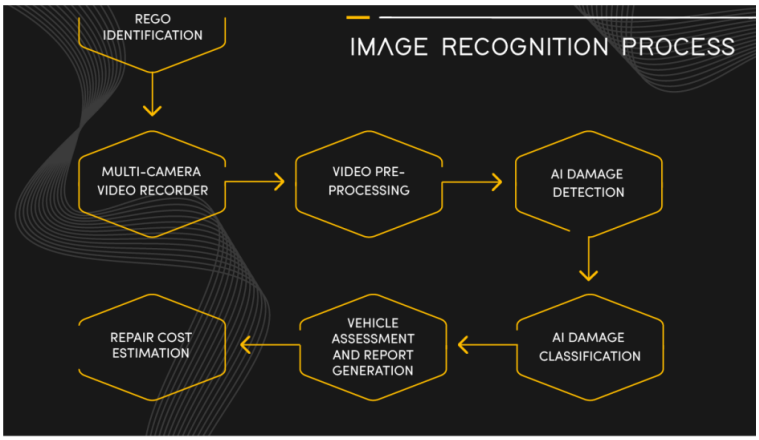
1. Highly accurate Image Predictive model
2. better user responsive web application
3. Cloud database for storing the informations

5. PROJECT DESIGN

5.1. DataFlow Diagram



5.2. Solution and Technical Architecture



5.3. User Stories

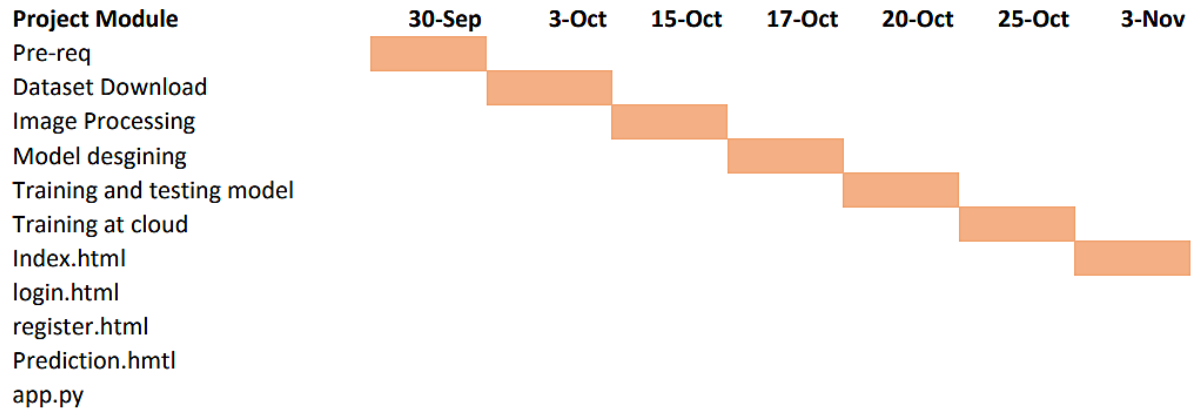
Use Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer (Mobile user)	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming my password.	I can access my account / dashboard by entering valid credentials	High	Sprint-1
Customer Details	Login	USN-2	As a user, I will receive confirmation email once I have registered for the application	I can receive confirmation email & click confirm	High	Sprint-1
Customer Uses	Dashboard	USN-3	As a user, I can register for the application through Facebook	I can register & access the dashboard with Facebook Login	Low	Sprint-4
Customer Options	Details about insurance companies	USN-4	As a user, I can register for the application through Gmail	I can register & access the dashboard with Facebook Gmail	Medium	Sprint-1
Customer usage	Login	USN-5	As a user, I can log into the application by entering email & password	I can log in and view my dashboard at my demand on any time	High	Sprint-1
Customer needs to do	Dashboard	USN-6	As a user I must capture images of my vehicle and upload it into the web portal	I can capture the entire vehicle and upload	High	Sprint-2
Customer (Web user)	Details about estimated cost based on damage	USN-7	As a user I must receive a detailed report of the damages present in the vehicle and the cost estimated	I can get the estimated insurance cost	High	Sprint-3
Customer Care Executive	Details about Estimated cost Based on damage	USN-8	As a user, I need to get support from developers in case of queries and failure of service provided	I can have smooth user experiences and all the issues raised is sorted	Medium	Sprint-4
Administrator	Details about Estimated cost Based on damage	USN-9	We need to satisfy the customer needs in an efficient way and make sure any sort of errors are fixed	I can finish the work without any problems	High	Sprint-4

6. PROJECT PLANNING

6.1. Sprint Planning and Estimation

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Prediction model	USN-1	How can I estimate the cost for vehicle assessment	2	High	Vimal
Sprint-1	Creating the Database in the cloud	USN-2	How can I store the information	1	High	Vimal
Sprint-2	Index.html	USN-3	As user, how can I access the prediction model in the fancy way	2	Low	Sachin
Sprint-3	Login.html	USN-4	As a user, I can log into the application by entering email & password	2	Medium	Suriya
Sprint-3	Register.html	USN-5	As a user, I can register for the application through Gmail	1	Medium	Suriya
Sprint-4	Prediction.html	USN-6	As the user, I can upload the image and make the estimation	1	High	Barath
Sprint-5	App.py	USN-7	How to connect the front end with the back end	1	High	Vimal

6.2. Sprint Delivery Schedule



7. CODING & SOLUTIONING

7.1. Feature 1

```
client = Cloudant.iam("1c6f917d-87ac-491b-90a0-6e3ae5b5daca-bluemix","tYJcUyVJYs3WrxF_1absTN4RXrbdQ_RDWBRUy9BX-28c",connect=True)
database = client.create_database("bath4_database")

#load model
model1 = load_model('V:\\WorkSpace\\IBM-Project-23426-1659882722\\Final Deliverables\\model\\body.h5')
model2 = load_model('V:\\WorkSpace\\IBM-Project-23426-1659882722\\Final Deliverables\\model\\level.h5')
```

The feature 1 gives access to the trained deep learning models for predicting mulitple damages in various areas in the vechile and connected with the IBM Waston Database for storing the user data.

7.2. Feature 2

```
img = load_img(filepath,target_size=(224,224))
        x = img_to_array(img)
        x = np.expand_dims(x,axis=0)
        img_data = preprocess_input(x)

        prediction1 = np.argmax(model1.predict(img_data))
        prediction2 = np.argmax(model2.predict(img_data))

        index1 = ['front','near','side']
        index2 = ['minor','moderate','severe']

        result1 = index1[prediction1]
        result2 = index2[prediction2]
```

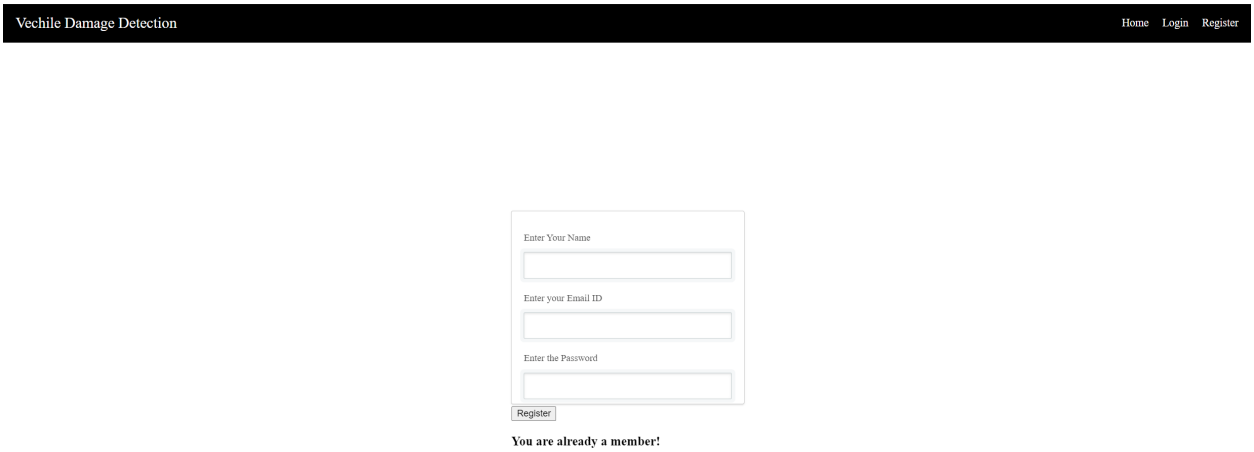
feature 2 enables the web application to predict the incomming image from the user into the given labels. The code gets the image, convert into pixcels and load into the model. Based on the predicted results, the algorithm will returns the value as the estimated cost.

8. TESTING

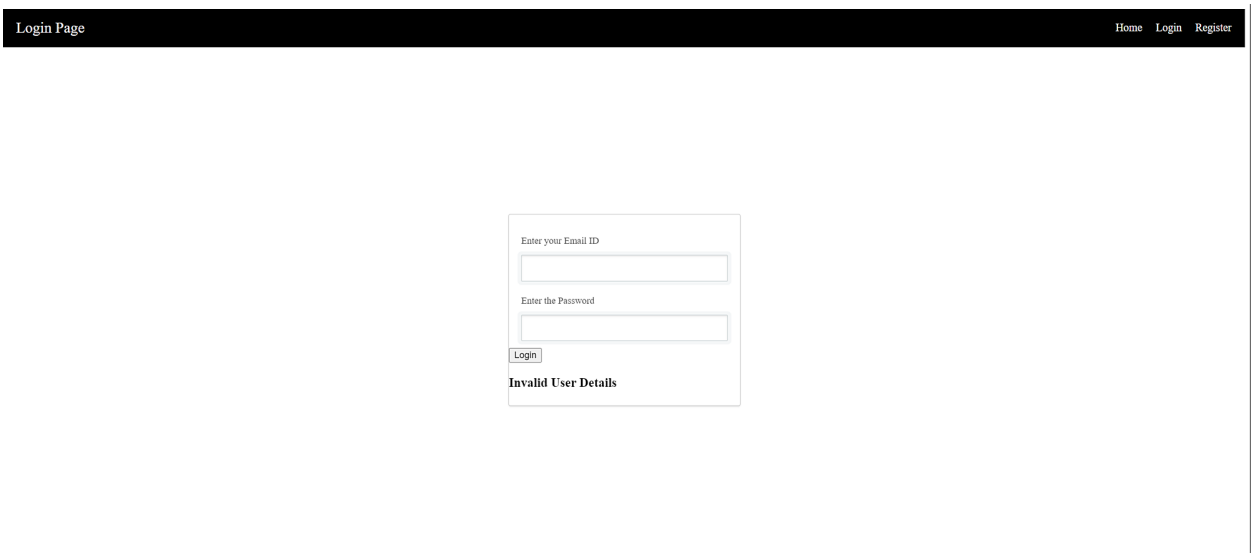
8.1. Test Cases

1. User Login and Registration test
2. Database Update test
3. Prediction test

8.2. User Acceptance Testing



The registeration web page is tested with the already registered user information and hence it shows a message "You are already a member" by which the repeation of user infromation at database is prevented.



The login web page is tested with the invalid user information to check the invalid login testing into the webpage.

The Estimated Cost of the Damage:6000 - 8000 Inr

The prediction page is given with the test image of a damaged car to check the accuracy of the models.

9. PERFORMANCE

The performance of the Cost estimator for insurance companies is tested and assested with the latency check, which is run over the prediction page. The time taken to load the image and predict the cost based on the damages in the vechile is checked. The results show that the web application took less than 10s to provide the estimated cost of the given vechile image. The model is tested with the various damaged car images which is not used during the training and validation of the model which also shows that the model works with the accuracy of about 98% in the overall performance.

10. ADVANTAGES AND DISADVANTAGES

1. The Advantage of having an Intelligent Cost Estimator based on the damages can save the time and resource of the user in automatically evaluating the images with the damages using the Deep Learning models trained with the various car images.
2. The Disadvantage of the project is expensive coding and time to develop the front end and back end of the web application

11. CONCLUSION

We conclude by suggesting this web application for damage assessment and cost estimation for the insurance companies. The web application is supported by the Deep Learning and IBM waston cloud which stands for the complex image prediction and user information storage. The web application takes the user registration and login, The user can login into the prediction page using their ID and password. The prediction

takes the image input and the model can predict the input based on the perviour knowledge about the damages.

12. FUTURE SCOPE

In future, The User Interface of the web application can be improved by updating the HTML and CSS codings. The improvement in UI can gives the better user exprience in future, The model's accuracy over various images can increased by trainning with various damaged images. The Image processing methods can be improved to achive higher performance of the model in the future.

13. APPENDIX

Github Repo:
https://github.com/IBM-EPBL/IBM-Project-23426-1659882722

VideoLink:
https://drive.google.com/file/d/1wZ_BgP7G8zyOnZ5q5xBCwM7FwBfBapm1/view

App.py

```
from cloudant.client import Cloudant
import os
import tensorflow
from keras.utils import load_img, img_to_array
from werkzeug.utils import secure_filename
import numpy as np
from keras.models import load_model
from tensorflow.python.ops.gen_array_ops import concat
from keras.applications.inception_v3 import preprocess_input

#creating the Cloudant Database
client = Cloudant.iam("1c6f917d-87ac-491b-90a0-6e3ae5b5daca-bluemix","tYJcUyVJYs3WrxF_1absTN4RXrbdQ_RDWBRUy9BX-28c",connect=True)
database = client.create_database("bath4_database")

#load model
model1     =     load_model('V:\\WorkSpace\\IBM-Project-23426-
```

```
1659882722\\Final Deliverables\\model\\body.h5')
model2 = load_model('V:\\WorkSpace\\IBM-Project-23426-1659882722\\Final Deliverables\\model\\level.h5')

from flask import Flask,render_template,request,redirect,url_for

app = Flask(__name__)

@app.route('/')
def home():
    return render_template('index.html')

#login page setting

@app.route('/login')
def login():
    return render_template('login.html')

@app.route('/afterLogin',methods=['POST','GET'])
def afterlogin():
    user = request.form['_id']
    passw = request.form['psw']
    print(user,passw)

    query = {'_id':{'$eq':user}}

    docs = database.get_query_result(query)
    print(docs)
    print(len(docs.all()))

    if(len(docs.all())==0):
            return render_template('login.html',message='The username is not found')
    else:
                        if((user==docs[0][0]['_id'] and passw==docs[0][0]['psw'])):
```

```
49              return redirect(url_for('prediction'))
50          else:
51                                                      return
  render_template("login.html",message="Invalid User Details")
52
53
54 #Register page setting
55
56 @app.route('/register')
57 def register():
58     return render_template('register.html')
59
60 @app.route('/afterRegister',methods=['POST'])
61 def afterregister():
62     x = [x for x in request.form.values()]
63     print(x)
64     data = {
65         '_id':x[1],
66         'name':x[0],
67         'psw' : x[2]
68     }
69     print(data)
70
71     query = {'_id':{'$eq' : data['_id']}}
72     docs = database.get_query_result(query)
73
74     if(len(docs.all())==0):
75         url = database.create_document(data)
76                         return   render_template('register.html',
  message="Registration is Successfully Completed")
77     else:
78          return render_template("register.html", message="You
  are already a member!")
79
80 #prediction
81
```

```
@app.route('/prediction')
def prediction():
    return render_template('prediction.html')

#logout page

@app.route('/logout')
def logout():
    return render_template('logout.html')

#results

@app.route('/result', methods = ['GET', 'POST'])
def upload_file():
   if request.method == 'POST':
        f = request.files['_file']
        basepath = os.path.dirname(__name__)
                                                filepath      =
  os.path.join(basepath,'uploads',f.filename)
          f.save(filepath)

          img = load_img(filepath,target_size=(224,224))
          x = img_to_array(img)
          x = np.expand_dims(x,axis=0)
          img_data = preprocess_input(x)

          prediction1 = np.argmax(model1.predict(img_data))
          prediction2 = np.argmax(model2.predict(img_data))

          index1 = ['front','near','side']
          index2 = ['minor','moderate','severe']

          result1 = index1[prediction1]
          result2 = index2[prediction2]

          if(result1=="front" and result2=="minor"):
```

```
            value= "3000 - 5000 Inr"
        elif(result1=="front" and result2=="moderate"):
            value ="6000 - 8000 Inr"
        elif(result1=="front" and result2=="severe"):
            value="9000 - 11000 Inr"
        elif(result1=="near" and result2=="minor"):
            value="4000 to 6000 Inr"
        elif(result1=="near" and result2=="moderate"):
            value="7000 - 9000 Inr"
        elif(result1=="near" and result2=="severe"):
            value="11000 - 13000 Inr"
        elif(result1=="side" and result2=="minor"):
            value="6000 - 8000 Inr"
        elif(result1=="side" and result2=="moderate"):
            value="9000 - 11000Inr"
        elif(result1=="side" and result2=="severe"):
            value="12000 - 15000 Inr"
        else:
            value = "16000 - 50000 Inr"

                                                    return
  render_template("prediction.html",prediction=value)


 if (__name__ == '__main__'):
     app.run(debug=True)
```