# **Project On**

# Intelligent Vehicle Damage Assessment & Cost Estimator for Insurance Companies

powered By IBM India

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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 designing 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

S.No	Author/ Publication Year	Title	Methodology	Advantages	Disadvantage s
1.	Phyu Mar Kyu, Kuntpong Woraratpany a, 2021	Car Damage Assessme nt Based on VGG Models.	CNN models on ImageNet dataset to perform different tasks of localization and detection. YOLO object detection model to train and detect damage region as their important pipeline to improve their performanc e of damage detection.	Transfer learning and regularizatio n can work better than those of fine tuning. Pretrained models assess its location and security which help insurance companies to solve claim leakage problems.	CNN cannot accurately calculate the level of damage part. Sometimes overfitting occurs.

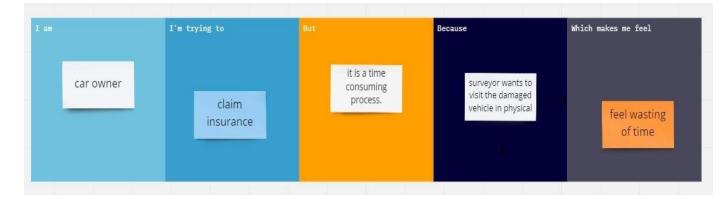
2	NT.: 1 1'	A 37-	C1. ' '	ТС	A .
2.	<u>Najmeddin</u>	A Very	Combination	Transfer	A
	<u>e Dhieb,</u>	Deep	of deep	learning	traditional
	<u>Hakim</u>	Transfer	learning,	significantl	CNN model
	<u>Ghazzai</u> ,	Learning	instance	y reduce the	can be very
	<u>Hichem</u>	Model for	segmentation	training	time
	<u>Besbes</u>	Vehicle	, and transfer	times when	consuming
	<u>Yehia</u>	Damage	learning	it uses the	to perform
	Massoud,	Detection	techniques	weight of	image
	2019	and	for features	pretrained	classificatio
		Localizatio	extraction	VGG	n tasks and
		n.	and damage	models. It	identify the
			identification	has	correct
			•	progress on	weights for
				how to	the network
				classify	by multiple
				problems	forward and
				when the	backward
				small	iterations.
				dataset was	
				not enough	
				to train a	
				CNN	
				model.	
3	U. Waqas,	Damage	Deep	It is a	The main
	N.	Assessment	learning	hybrid	drawback
	Akram, S.	of a vehicle	techniques,	approach	was that
	Kim, D.	and	Moire effect	which	Images in
	Lee and J.	Insurance	Detection,	provide	bad
	Jeon, t,	Reclaim.	Mobile Net	only	lighting,
	2012	210011111	model is	authentic	awkward
			proposed	images to	angles, and
			with transfer	algorithm	vehicle
			learning for	for damage	models in a
			classification	classificatio	small
			Classification	n as input.	dataset to
			•	moiré effect	achieve
				detection	automation
				and	is difficult
				metadata	but still the
				analysis are	range is
				performed	broad.
				to detect	
				fraudulent	
				images	

4.	Li Ying & Dorai Chitra, 2012	Applyin g image analysis to auto insuranc e Triage	Image analysis and pattern recognition are applied to automaticall y identify and characterize automobile damage.	Because of the advancement of image analysis and pattern recognition technologies, the auto insurance industry could significantly benefit.	The drawback is that the automobile damaged can be analyzed only having white background otherwise it will be not able to give the desired results.
5.	Srimal Jayewarde ne', 2013	Image based automatic vehicle damage detection	This approach requires 3D computer aided design (CAD) modes of the considered vehicle to identify how it would look if it were undamaged .	Automatically detecting the damage of the vehicle using photographs clicked at the accident site is extremely functional as it can greatly decrease the rate of processing insurance claims, and it will also provide greater conveniences for customers who are making the best use of this functionalit y.	Vehicles have very reflective metallic bodies the photograph s taken in such an uncontrolle d environmen t can be expected to have a certain amount of inter object reflection. Application of standard computer vision techniques is a very challenging task

	D1 3.4	C 1	CNDI 11'	D	TD C
6.	Phyu Mar Kyu,Kuntp ong Woraratpan ya ,2020	Car damage detection and classification	CNN model is trained on ImageNet dataset. After fine tuning the dataset, transfer learning with L2 regularization is applied	Pre- trained VGG model not only detect damage d part of a car but also assess its location and severity.	Transfer learning and regularizatio n can work better than those of fine tuning.
7.	M.W assel, 2019	A Secure AI- driven Architecture for Automated Insurance Systems: FraudDetect i on and Risk Measuremen t	Blockchain, data analysis, machine learning, AI for damage identification .	Proposed classifier s ensure not only the best accuracy in detecting fraudule nt claims but also can classify different types of fraud for insuranc e unlike the existing solutions.	The major drawback of the proposed model is that it only identifies the physical visible damage and not of the internal or the interior damage.

# 2.3. Problem Statement Definition

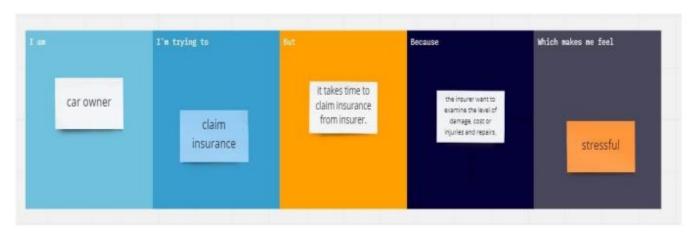
# **Problem Statement - 1**



# **Problem Statement - 2**

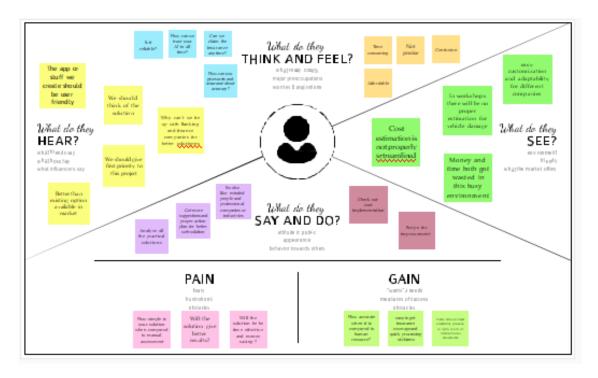


# **Problem Statement - 3**

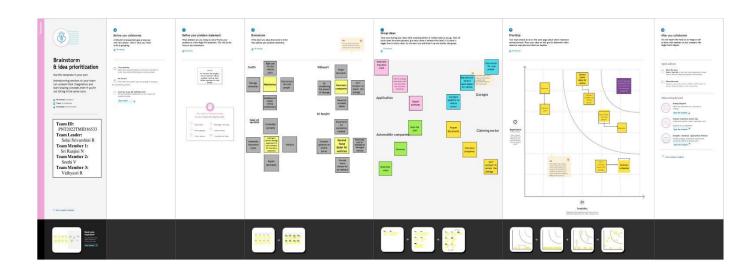


# **3.Ideation and Proposed Solution**

# 3.1. Empathy Map



# 3.2. Ideation and Brainstorming



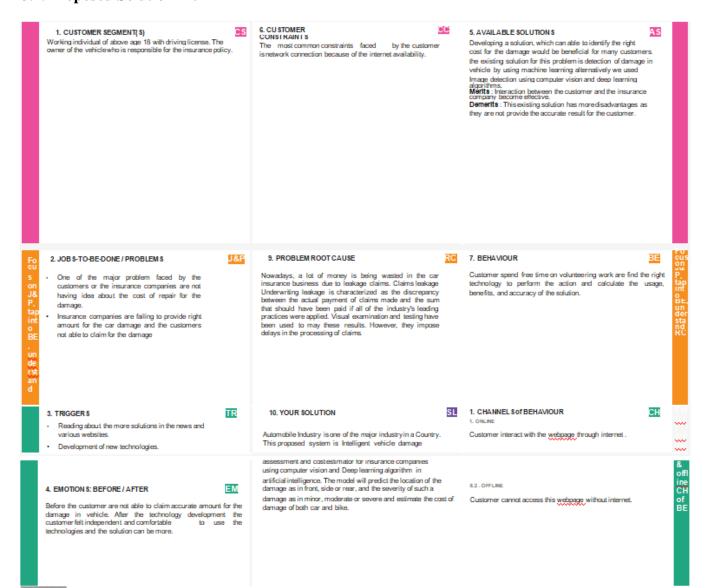
# 3.3. Proposed Solution

S.No.	Parameter	Description
1.	Problem Statement (Problem to be solved)	Every asset has a value attached to it that is primarily economic in nature. There is always a risk of these assets being destroyed due to incidents beyond human control. They also may not work due to such events. Depending on the asset class, the type and weight of risk also vary. This is where insurance policies are useful. The problem that might arise is that the claimant maynot know the amount of coverage that he/she has.
2.	Idea / Solution description	<ol> <li>To develop an optimized and accuratedeep learning architecture to detect the damage percentage and location of the damage with respect to the vehicle</li> <li>Implementing classification algorithms to classify damaged regions and implementing the model in web based application</li> <li>Create a user accessible portal and securely store the data provided by the user</li> <li>Compare the obtained damage percentage with the statistical cost estimation value to predict the cost.</li> </ol>

3.	Novelty / Uniqueness	<ol> <li>The deep learning algorithm will analyze images in real time and identifies the presence of any damage.</li> <li>Even in the presence of minute damages, artificial intelligence can detect the dents and marks on the car's body.</li> <li>With a lot of training, Artificial intelligence will able to distinguish simple stain from ascratch and effectively estimate the respective damage cost</li> </ol>
4.	Social Impact / Customer Satisfaction	<ol> <li>All the features of this project will bemade easily accessible to the customers.</li> <li>The webapp is intuitive, easy to use, simple and that the customer can rely on the product. It is easy to start with the appand understand how to use it, high complexity is not valuable for the user.</li> <li>All the uploaded images will be and the personal information of the customerwill be secured in cloud data security.</li> <li>The cost estimation for damages that the webapp provides to the customer will be legitimate and exactto what a normal insurance company offers.</li> </ol>

5.	Business Model (Revenue Model)	<ol> <li>The business model will be a freemium model providing the prediction of damage intensity which will be useful for the vehicle owners to keep track oftheir vehicle damage and the credentials toaccess the webpage can be provided on the purchase of thevehicle insurance.</li> <li>The add-on subscription model can be initiated forthe user where the damagecost is evaluated and provided to the users.</li> <li>The further revenue can be generated by tying upwith the automobile parts manufacturers and distributors by promoting their products to the vehicle that has specified parts damaged.</li> </ol>
6.	Scalability of the Solution	<ol> <li>The damage detection can be provided to all the insured clients to reach the stable base and then extend the service of cost estimation to the insurers.</li> <li>Make use of advanced machine learning techniques to analyze the damaged vehicle with high accuracy levels and keep on improving the learning ability of the model.</li> <li>In addition to the webpage a mobile application canbe created where the real time images and videos of the vehicle canbe extracted and insurance cost can be estimated.</li> </ol>

## 3.4. Proposed Solution Fit



# 4. REQUIREMENT ANALYSIS

4.1. Functional Requirement

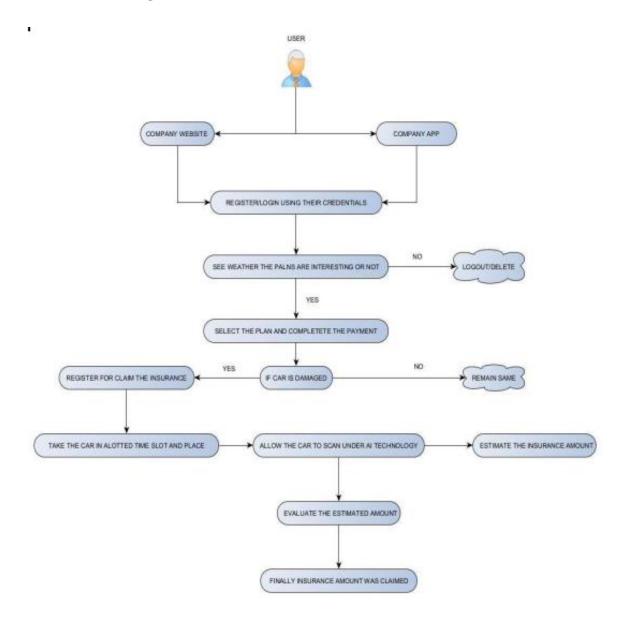
FR	Functional	Sub Requirement (Story / Sub-Task)
No.	Requirement (Epic)	-
FR-1	User Registration	Registration
		through Form
		Registration
		through Gmail
		Registration through LinkedIN
FR-2	User Confirmation	Confirmation
		via Email
		Confirmation
		via OTP
FR-3	User Login	Login via
		Gmail
		Login via
		credentials
FR-4	Purchase	Purchase the insurance plan from list
FR-5	Apply	Apply when car got damaged
FR-6	Allocation	Allot a place and time to user for Scanning and
		estimatethe insurance amount
FR-7	Result	The estimated amount was showed and provided to
		user

**4.2. Non - Functional Requirements** 

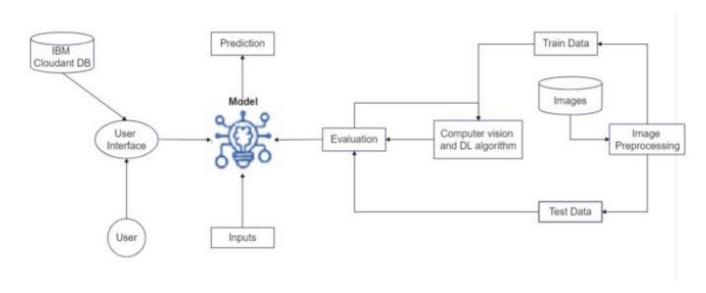
FR	Non-Functional	Description
No.	Requirement	
NFR-1	Usability	The Final data should be easily
		understandable .
NFR-2	Security	The model are designed in a secured manner inorder to maintain the privacy
NFR-3	Reliability	Even if there is a firmware issues (failures) the lastupdated Data's are stored in a Default manner.
NFR-4	Performance	High quality sensors are used to ease the customer's work.
NFR-5	Availability	The model are designed in such a way that areavailable ,usable and can be modified anytime.
NFR-6	Scalability	The size of the system didn't varies it is made tousable for all types of cars

# 5. PROJECT DESIGN

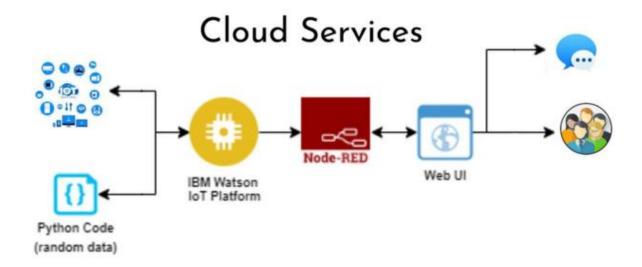
# 5.1. Data Flow Diagram



# 5.2. Solution and Technical Architecture



# **Technical Architecture**



# **5.3. User Stories**

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer (Mobile and Web 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	High	Sprint-1
		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
		USN-3	As a user, I can register for the application through Google account	I can register & access the dashboard with google Login	Low	Sprint-2
	Login	USN-4	As a user, I can log into the application by entering email & password	I can login my account	Medium	Sprint-1
	Dashboard	USN-5	As a user now I can see the required information	I can know the required information	High	Sprint-1
	All plans	USN-6	As a user now I can choose the best plan	I can choose a best plan among the list	High	Sprint-1
	Payment	USN-7	As a user now I can pay the insurance amount through online via credit/debit card/upi	I can pay the insurance amount which I was selected	High	Sprint-2
	Acknowledgement	USN-8	As a user now I will receive the Acknowledgement in mail	I can receive the bill via mail	Medium	Sprint-1
	Apply	USN-9	As a user now I can apply for claiming the insurance if my car is damaged	I can register for claiming the insurance	High	Sprint-1
Customer Care Executive	Slot Allocation	USN-10	As a User the Customer Care Executive will allocate the slot for me to Scan my car	I can take my car to the Allocated place and time	Medium	Sprint-1
	Estimation	USN-11	As a user now I can know the estimated insurance amount for my car by Scanning	I can know the estimated amount by Scanning	High	Sprint-2
Administrator	Officer	USN-12	As a user know I received the estimated insurance amount by the insurance officer	I can receive the estimated insurance amount	Very High	Sprint-2
	Receipt	USN-13	As a user now I can receive the Receipt for the received insurance amount	I can receive the bill for claiming the insurance	Medium	Sprint-1
Mechanical job	Repairing	USN-14	As a user now I can repair my car by using the insurance amount	I can repair my damaged car	low	Sprint-1

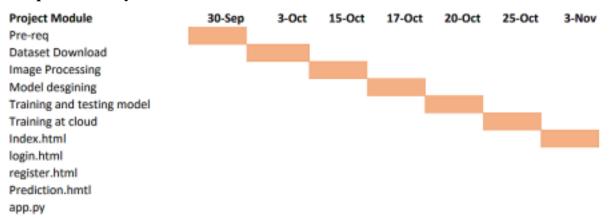
# 6. PROJECT PLANNING

# **6.1. Sprint Planning and Estimation**

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Registration	USN-1	As an owner of a particular vehicle, I can log into the application by entering email & password.		High	Solai srivarshini R Sri Ranjini N Sruthi V Vidhyasri R
Sprint-1	User Confirmation	USN-2	As an owner of a particular Vehicle, I will receive confirmation email once I have registered for the application.		Medium	Solai srivarshini R Sri Ranjini N Sruthi V Vidhyasri R
Sprint-1	Login	USN-3	As an owner of a particular vehicle, I can log into the application by entering email & password.	2	High	Solai srivarshini R Sri Ranjini N Sruthi V Vidhyasri R
Sprint-2	Data Collection	USN-1	Download the dataset used in intelligent vehicle damage assessment & cost estimator for insurance companies.		High	Solai srivarshini R Sri Ranjini N Sruthi V Vidhyasri R

Sprint-2	Image Pre Processing	USN-1	Improve the image data that suppresses unwilling distortions or enhances some image features important for further processing, although performing some geometric transformations of images like rotation, scaling, etc.	2	High	Solai srivarshini R Sri Ranjini N Sruthi V Vidhyasri R
Sprint-3	Model Building	USN-1	Define the model architecture and adding CNN layer and testing , saving the model.	2	High	Solai srivarshini R Sri Ranjini N Sruthi V Vidhyasri R
Sprint-3	Cloud DB	USN-1	Below are steps that need to follow for creating and using cloud ant service.  Register & login to IBM cloud Create service instance Creating service credentials Launch cloud ant DB Create database		High	Solai srivarshini R Sri Ranjini N Sruthi V Vidhyasri R
Sprint-4	Application Building	USN-1	Building a web application that is integrated into the model we built. A UI is provided to the user where he has uploaded the image. Based on the saved model, the uploaded image will be analyzed and prediction is showcased on the UI.		High	Solai srivarshini R Sri Ranjini N Sruthi V Vidhyasri R
Sprint-4	Train The Model On IBM	USN-1	Build Deep learning model and computer vision Using the IBM cloud.	2	High	Solai srivarshini R Sri Ranjini N Sruthi V Vidhyasri R

# **6.2. Sprint Delivery Schedule**



#### 7. CODING & SOLUTION

#### **7.1. Feature 1**

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**

```
1 img = load_img(filepath,target_size=(224,224))
2 x = img_to_array(img)
3 x = np.expand_dims(x,axis=0)
4 img_data = preprocess_input(x)
5
6    prediction1 = np.argmax(model1.predict(img_data)) 7    prediction2 = np.argmax(model2.predict(img_data)) 8
9 index1 = ['front','near','side']
10 index2 = ['minor','moderate','severe']
11
12 result1 = index1[prediction1]
13 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-7224-1658850233

#### VideoLink:

https://drive.google.com/file/d/1wZ\_BgP7G8zyOnZ5q5xBCwM7FwBfBapm1/view

## App.py

```
1 from cloudant.client import Cloudant
2 import os
3 import tensorflow
4 from keras.utils import load_img, img_to_array
5 from werkzeug.utils import secure filename
6 import numpy as np
7 from keras.models import load model
8 from tensorflow.python.ops.gen_array_ops import concat 9 from
keras.applications.inception_v3 import preprocess_input 10
11 #creating the Cloudant Database
12 client = Cloudant.iam("1c6f917d-87ac-491b-90a0-6e3ae5b5daca
   bluemix","tYJcUyVJYs3WrxF_1absTN4RXrbdQ_RDWBRUy9BX 28c",connect=True)
13 database = client.create_database("database_")
14
15 #load model
16 model1 = load_model('C:\IBM Folder\Vehicle Damage\model\body.h5')
17 model2 = load_model('C:\IBM Folder\Vehicle Damage\model\level.h5')
18
19 from flask import Flask,render_template,request,redirect,url_for
21 \text{ app} = \text{Flask}(\underline{\text{name}})
```

```
22
23 @app.route('/')
24 def home():
25 return render_template('index.html')
26
27 #login page setting
28
29 @app.route('/login')
30 def login():
31 return render_template('login.html')
32
33 @app.route('/afterLogin',methods=['POST','GET']) 34 def afterlogin():
35 user = request.form[' id']
36 passw = request.form['psw']
37 print(user,passw)
38
39 query = \{'_id': \{'\}eq':user\}\}
40
41 docs = database.get_query_result(query)
42 print(docs)
43 print(len(docs.all()))
44
45 \text{ if}(\text{len}(\text{docs.all}())==0):
46 return render_template('login.html',message='The username is not found')
47 else:
48 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 \text{ for } x \text{ 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
82 @app.route('/prediction')
83 def prediction():
84 return render_template('prediction.html')
85
86 #logout page
87
88 @app.route('/logout')
89 def logout():
90 return render_template('logout.html')
91
92 #results
93
94 @app.route('/result', methods = ['GET', 'POST']) 95 def upload_file():
96 if request.method == 'POST':
97 f = request.files['_file']
98 basepath = os.path.dirname(__name__)
99 filepath = os.path.join(basepath, 'uploads', f. filename)
100 f.save(filepath)
101
102 img = load img(filepath,target size=(224,224))
```

```
103 \text{ x} = \text{img to array(img)}
104 \text{ x} = \text{np.expand\_dims}(x, axis=0)
105 img_data = preprocess_input(x)
106
107 prediction1 = np.argmax(model1.predict(img_data))
108 prediction2 = np.argmax(model2.predict(img_data))
109
110 index1 = ['front', 'near', 'side']
111 index2 = ['minor', 'moderate', 'severe']
112
113 result1 = index1[prediction1]
114 result2 = index2[prediction2]
115
116 if(result1=="front" and result2=="minor"): 15
117 value= "3000 - 5000 Inr"
118 elif(result1=="front" and result2=="moderate"):
119 value = "6000 - 8000 Inr"
120 elif(result1=="front" and result2=="severe"):
121 value="9000 - 11000 Inr"
122 elif(result1=="near" and result2=="minor"):
123 value="4000 to 6000 Inr"
124 elif(result1=="near" and result2=="moderate"):
125 value="7000 - 9000 Inr"
126 elif(result1=="near" and result2=="severe"):
127 value="11000 - 13000 Inr"
128 elif(result1=="side" and result2=="minor"):
129 value="6000 - 8000 Inr"
130 elif(result1=="side" and result2=="moderate"):
131 value="9000 - 11000Inr"
132 elif(result1=="side" and result2=="severe"):
133 value="12000 - 15000 Inr"
134 else:
135 value = "16000 - 50000 Inr"
136
137 return render_template("prediction.html",prediction=value) 138
139
```

```
140
141 if (__name__ == '__main__'):
142 app.run(debug=True)
```