

A REVIEW OF LIVER PATIENT ANALYSIS

TEAM ID:NM2023TMID19450

TEAM SIZE :4

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PROJECT REPORT

1. INTRODUCTION

Overview

A Review Of Liver Patient Analysis Methods Using Machine Learning

Project Description:

Liver diseases averts the normal function of the liver. This disease is caused by an assortment of elements that harm the liver. Diagnosis of liver infection at the preliminary stage is important for better treatment. In today's scenario devices like sensors are used for detection of infections. Accurate classification techniques are required for automatic identification of disease samples. This disease diagnosis is very costly and complicated. Therefore, the goal of this work is to evaluate the performance of different Machine Learning algorithms in order to reduce the high

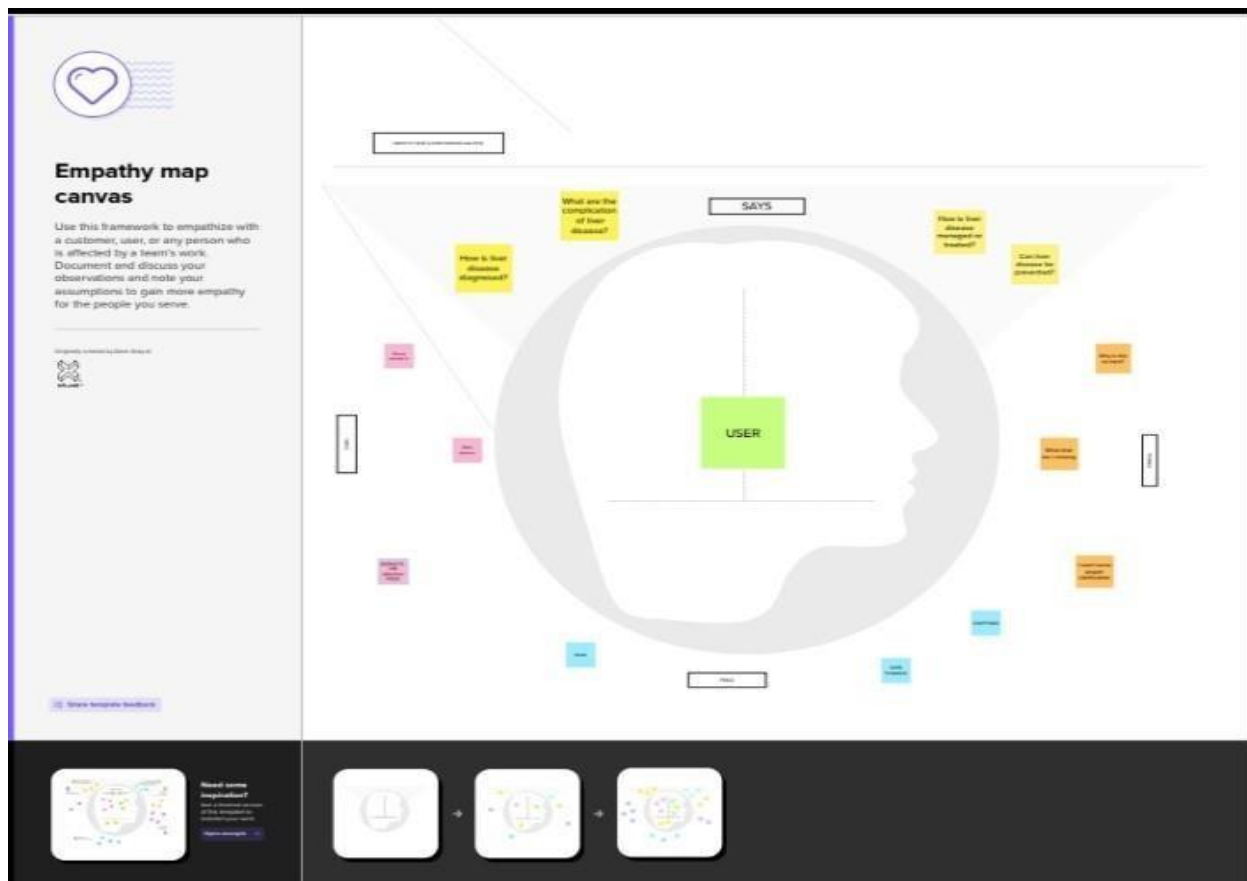
cost of liver disease diagnosis. Early prediction of liver disease using classification algorithms is an efficacious task that can help the doctors to diagnose the disease within a short duration of time. In this project we will analyse the parameters of various classification algorithms and compare their predictive accuracies so as to find out the best classifier for determining the liver disease. This project compares various classification algorithms such as Random Forest, Logistic Regression, KNN and ANN Algorithm with an aim to identify the best technique. Based on this study, Random Forest with the highest accuracy outperformed the other algorithms and can be further utilised in the prediction of liver disease and can be recommended to the user.

PURPOSE

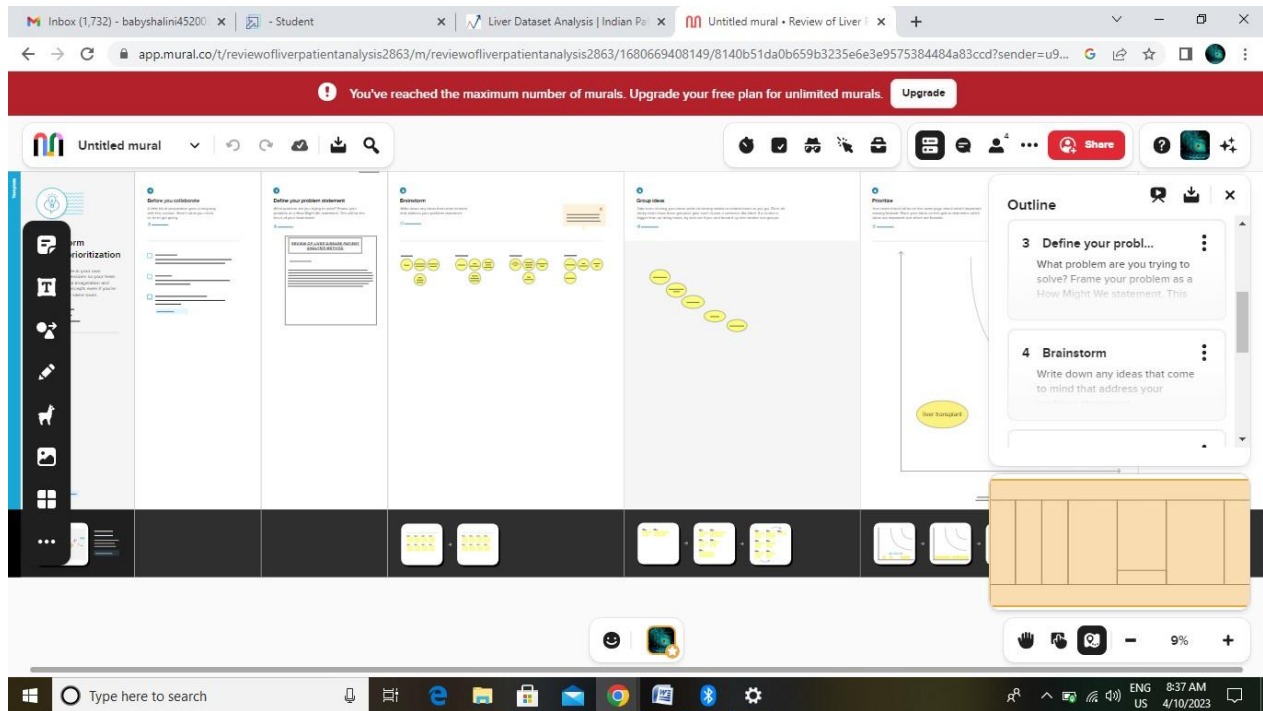
- Liver cirrhosis is the biggest health problem posed by alcohol use, with 1.4 lakh deaths every year.
- Sadly, no. In fact, it is getting more common in younger people than ever before. Dr. Amrish said that liver disease can set in childhood too as it can pass through genes.
- Cirrhosis isn't curable, but it's treatable. Alcohol abuse, hepatitis, and fatty liver disease are some of the main causes.

- Then you people will get answers like these as I mentioned above, So the purpose and inspiration of this project clearly simplifies the devastating answers from the data available with Google. We do need a system that in some stage reduces the burden on doctors, and today in this article I'll try to frame a practical logic that will help our healthcare system in a long run.

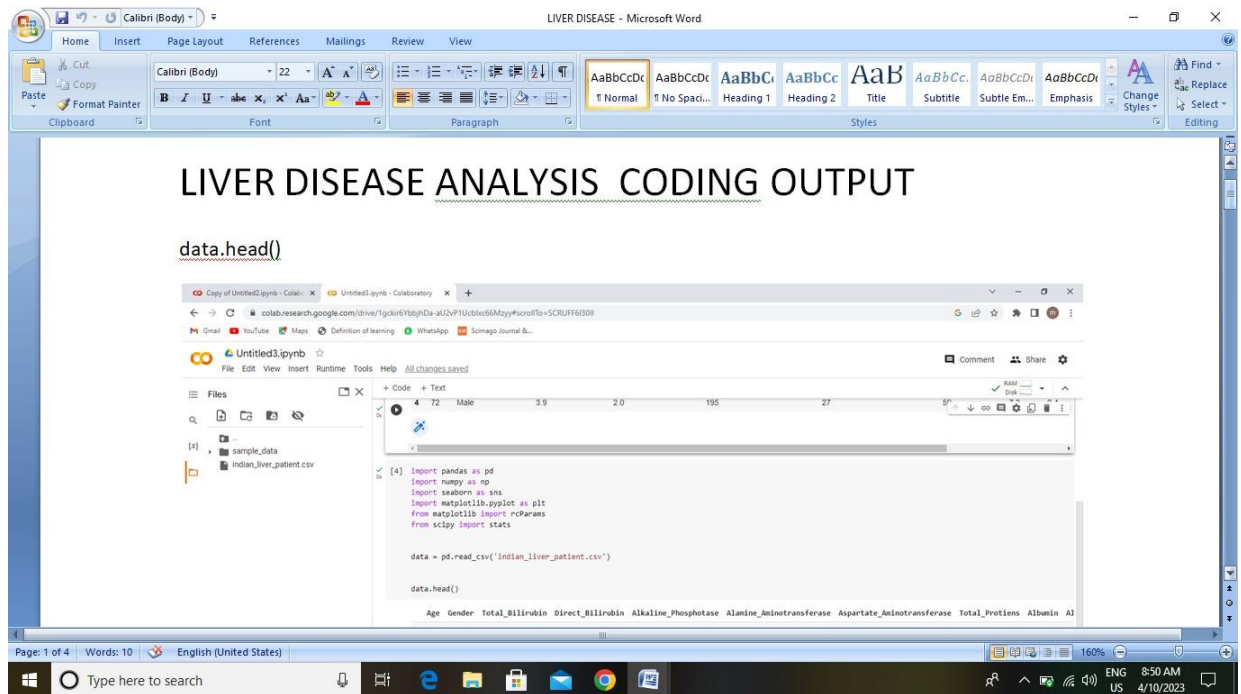
2. PROBLEM DEFINITION & DESIGN THINKING



IDEATION & BRAINSTORMING MAP



3. RESULT



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data.isnull().sum()

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Files

sample_data indian_liver_patient.csv

Code + Text

```
data.isnull().sum()
Age 0
Gender 0
Total_Bilirubin 0
Direct_Bilirubin 0
Alkaline_Phosphatase 0
Alamine_Aminotransferase 0
Aspartate_Aminotransferase 0
Total_Proteins 0
Albumin 0
Albumin_and_Globulin_Ratio 0
Dataset 0
dtype: int64
```

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data.describe()

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Files

sample_data indian_liver_patient.csv

Code + Text

```
data.describe()
Age Total_Bilirubin Direct_Bilirubin Alkaline_Phosphatase Alamine_Aminotransferase Aspartate_Aminotransferase Total_Proteins Albumin
count 563.000000 563.000000 563.000000 563.000000 563.000000 563.000000 563.000000 563.000000
mean 44.746141 3.298799 1.486106 290.576329 80.713561 108.919806 6.483190 3.1416
std 16.189633 6.209522 2.806496 242.937969 162.620356 288.918529 1.085451 0.7965
min 4.000000 0.400000 0.100000 63.000000 10.000000 10.000000 2.700000 0.900000
25% 33.000000 0.800000 0.200000 175.000000 23.000000 25.000000 5.800000 2.600000
50% 45.000000 1.000000 0.300000 208.000000 35.000000 42.000000 6.600000 3.100000
75% 58.000000 2.600000 1.300000 298.000000 60.500000 87.000000 7.200000 3.800000
max 90.000000 75.000000 19.700000 2110.000000 2000.000000 4929.000000 9.600000 5.500000
```

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data.describe()

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Files

sample_data indian_liver_patient.csv

Code + Text

data.describe()

	Age	Total_Bilirubin	Direct_Bilirubin	Alkaline_Phosphatase	Alanine_Aminotransferase	Aspartate_Aminotransferase	Total_Proteins	Album
count	583.000000	583.000000	583.000000	583.000000	583.000000	583.000000	583.000000	583.000000
mean	44.746141	3.296799	1.486106	290.576329	80.713551	109.910806	6.483190	3.1418
std	16.189633	6.209522	2.806498	242.937989	182.620356	288.918529	1.085451	0.7955
min	4.000000	0.400000	0.100000	63.000000	10.000000	10.000000	2.700000	0.9000
25%	33.000000	0.800000	0.200000	175.500000	23.000000	25.000000	5.800000	2.6000
50%	45.000000	1.000000	0.300000	208.000000	35.000000	42.000000	6.600000	3.1000
75%	58.000000	2.600000	1.300000	296.000000	60.500000	87.000000	7.200000	3.8000
max	90.000000	75.000000	19.700000	2110.000000	2000.000000	4929.000000	9.600000	5.5000

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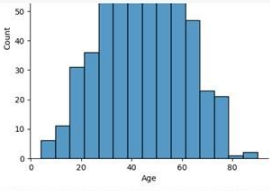
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Code + Text

```
sns.heatmap(df.corr(),annot=True)
from sklearn.preprocessing import scale
x=data.iloc[:,1:]
y=data[Y]
x_scaled=pd.DataFrame(scale(x),columns=x.columns)
plt.figure(figsize=(10,7))
sns.heatmap(df.corr(),annot=True)
```



```
-----
NameError                                Traceback (most recent call last)
<ipython-input-47-c2a82aa3438b> in <cell line: 1>():
     11 from sklearn.preprocessing import scale
     12 x=data.iloc[:,1:]
--> 13 y=data[Y]
     14 x_scaled=pd.DataFrame(scale(x),columns=x.columns)
```

Page: 3 of 4 Words: 10 English (United States) 160% ENG US 8:55 AM 4/10/2023

Microsoft Word window titled "LIVER DISEASE - Microsoft Word" showing a Jupyter Notebook interface. The notebook displays a bar chart titled "Gender" with two categories: "Female" (blue bar, count ~150) and "Male" (orange bar, count ~350). The x-axis is labeled "count" and ranges from 0 to 400. The y-axis is labeled "Gender". The notebook also shows a code cell with the following Python code:

```
13 ydata[V]
14 x_scaled=pd.DataFrame(scale(x),column=x.columns)
15 plt.figure(figsize=(10,7))
```

The code cell is followed by an error message: "NameError: name 'y' is not defined". The notebook interface includes a toolbar with options like "File", "Edit", "View", "Insert", "Runtime", "Tools", and "Help". The status bar at the bottom indicates "Page: 4 of 4", "Words: 10", and "English (United States)".

Microsoft Word window titled "LIVER DISEASE - Microsoft Word" showing a Jupyter Notebook interface. The notebook displays a heatmap titled "Dataset" showing correlations between various features. The features listed are: Age, Total_Bilirubin, Direct_Bilirubin, Alkaline_Phosphatase, Alamine_Aminotransferase, Aspartate_Aminotransferase, Total_Proteins, Albumin, Albumin_and_Globulin_Ratio, and Dataset. The heatmap uses a color scale from -0.2 (dark purple) to 1.0 (dark red). The diagonal elements are all 1.0. The correlations are as follows:

	Age	Total_Bilirubin	Direct_Bilirubin	Alkaline_Phosphatase	Alamine_Aminotransferase	Aspartate_Aminotransferase	Total_Proteins	Albumin	Albumin_and_Globulin_Ratio	Dataset
Age	1	0.012	0.0075	0.08	-0.087	-0.02	-0.19	-0.27	-0.22	-0.14
Total_Bilirubin	0.012	1	0.87	0.21	0.21	0.24	-0.0081	-0.22	-0.21	-0.22
Direct_Bilirubin	0.0075	0.87	1	0.23	0.23	0.26	-0.00014	-0.23	-0.2	-0.25
Alkaline_Phosphatase	0.08	0.21	0.23	1	0.13	0.17	-0.029	-0.17	-0.23	-0.18
Alamine_Aminotransferase	-0.087	0.21	0.23	0.13	1	0.79	-0.043	-0.03	-0.0024	-0.16
Aspartate_Aminotransferase	-0.02	0.24	0.26	0.17	0.79	1	-0.026	-0.085	-0.07	-0.15
Total_Proteins	-0.19	-0.0081	-0.00014	-0.029	-0.043	-0.026	1	0.78	0.23	0.035
Albumin	-0.27	-0.22	-0.23	-0.17	-0.03	-0.085	0.78	1	0.69	0.16
Albumin_and_Globulin_Ratio	-0.22	-0.21	-0.2	-0.23	-0.0024	-0.07	0.23	0.69	1	0.16
Dataset	-0.14	-0.22	-0.25	-0.18	-0.16	-0.15	0.035	0.16	0.16	1

The notebook interface includes a toolbar with options like "File", "Edit", "View", "Insert", "Runtime", "Tools", and "Help". The status bar at the bottom indicates "Page: 4 of 4", "Words: 10", and "English (United States)".

4. ADVANTAGES & DISADVANTAGES

Liver biopsy	
<i>Benefits</i>	<i>Disadvantages</i>
Clear diagnostic criteria	Major invasive test
Diagnostic value confirmed	Complications include death
May suggest the etiology	Significant sampling errors
Can perform differential diagnosis	High cost
Assess the degree and stage of liver damage	Inter-observer variability
It can decide the therapy	

5. APPLICATIONS

The liver filters all of the blood in the body and breaks down poisonous substances, such as alcohol and drugs. The liver also produces bile, a fluid that helps digest fats and carry away waste.

- Hospitals.
- Specialty Clinics.
- Medical Research Department.

6.

CONCLUSION

The main roles of the liver include removing toxins, processing food nutrients and regulating body metabolism. Important causes of liver disorders are fatty liver, hepatitis virus infections and alcohol. Cirrhosis (liver scarring), the end-result of many liver disorders, can lead to liver failure.

7.FUTURE SCOPE

- Hospitals.
- Specialty Clinics.
- Medical Research Department.
- Patient(Body)

8. APPENDIX

Source Code

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from matplotlib import rcParams
from scipy import stats

data = pd.read_csv('indian_liver_patient.csv')

data.head()

data.info()

data.isnull().any()

data.isnull().sum()

data['Albumin_and_Globulin_Ratio'] =
data.fillna(data['Albumin_and_Globulin_Ratio'].mode()[0])
data.isnull().sum()
```

```
from sklearn.preprocessing import LabelEncoder
lc = LabelEncoder()
data['gender']= lc.fit_transform(data['gender'])
```

```
data.describe()
```

```
sns.displot(data['age'])
plt.title('Age Distribution Graph')
plt.show()
```

```
sns.countplot(data['outcome'], hue=data['gender'])
```

```
plt.figure(figsize=(10,7))
sns.heatmap(df.corr(),annot=True)
```

```
from sklearn.preprocessing import scale
X_scaled=pd.DataFrame (scale(X), column=X.columns)
```

```
X_scaled.head()
```

```
x=data.iloc[:, :-1]
y=data.outcome
```

```
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x_scaled,y,
test_size=0.2,random_state=42)
```

```
pip install imblearn
```

```
from imblearn.over_sampling import SMOTE  
smote = SMOTE()
```

```
y_train.value_counts()
```

```
x_train_smote, y_train_smote = smote.fit_resample(x_train, y_train)
```

```
y_train_smote.value_counts()
```

```
from sklearn.ensemble import RandomForestClassifier  
model1=RandomForestClassifier()  
model1.fit(x_train_smote, y_train_smote)  
y_predict=model1.predict(x_test)  
rfc1=accuracy_score(y_test,y_predict)  
rfc1  
pd.crosstab(y_test, y_predict)  
print(classification_report(y_test, y_predict))
```

```
from sklearn.tree import DecisionTreeClassifier  
model4=DecisionTreeClassifier()  
model4.fit(x_train_smote, y_train_smote)  
y_predict=model4.predict(x_test)  
dct1=accuracy_score(y_test,y_predict)  
dct1  
pd.crosstab(y_test,y_predict)  
print(classification_report(y_test, y_predict))
```

```
import tensorflow.keras
from tensorflow.keras.models import sequential
from tensorflow.keras.layers import Dense

classifier = Sequential()

classifier.add(Dense(units=100, activation='relu', input_dim=10))

classifier.add(Dense(units=50, activation='relu'))

classifier.add(Dense(units=1, activation='sigmoid'))

classifier.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])

model_history = classifier.fit(x_train, y_train, batch_size=100, validation_split=0.2,
epochs=100)

model4.predict([[50,1,1.2,0.8,150,70,80,7.2,3.4,0.8]])

model1.predict([[50,1,1.2,0.8,150,70,80,7.2,3.4,0.8]])

classifier.save("liver.hs")

y_pred = classifier.predict(x_test)
y_pred
```

```
y_pred = (y_pred > 0.5)
y_pred
```

```
def predict_exit(sample_value):
    sample_value = np.array(sample_value)
    sample_value = sample_value.reshape(1,-1)
    sample_value = scale(sample_value)
    return classifier.predict(sample_value)
```

```
sample_value = [[50,1,1.2,0.8,150,70,80,7.2,3.4,0.8]]
if predict_exit(sample_value)>0.5:
    print('Prediction: Liver patient')
else:
    print('Prediction: Healthy')
```

```
acc_smote= [['KNN Classifier', Knn1], ['RandomForestClassifier', rfc1],
            ['DecisionTreeClassifier', dtc1], ['LogisticRegression', log1]]
Liverpatient_pred= pd.DataFrame(acc_smote, columns = ['classification models',
            'accuracy_score'])
Liverpatient_pred
```

```
plt.figure(figsize=(7,5))
plt.xticks(rotation=90)
plt.title('Classification models & accuracy scores after SMOTE',fontsize=18)
sns.barplot(x="classification models", y="accuracy_score",
            data=Liverpatient_pred,palette = "Set2")
```

```
from sklearn.ensemble import ExtraTreesClassifier
model=ExtraTreesClassifier()
model.fit(x,y)
```

```
ExtraTreesClassifier()
```

```
model.feature_importances_  
dd=pd.DataFrame(model.feature_importances_,index=X.columns).sort_values(0,ascending=False)  
dd
```

```
dd.plot(kind='barch', figsize=(7,6))  
plt.title("FEATURE IMPORTANCE",fontsize=14)
```

```
import joblib  
joblib.dump(model1, 'ETC.pkl')
```

```
from flask import Flask, render_template, request  
import numpy as np  
import pickle
```

```
app=Flask(__name_)  
@app.route('/')  
def home():  
    return render_template('home.html')  
@app.route('/predict')  
def index():  
    return render_template("index.html")
```

```
@app.route('/data_predict', methods=['POST'])  
def predict():
```

```
data = [[float(age), float(gender), float(tb), float(db), float(ap), float(aa1), float(aa2),  
float(tp),
```



```
model=pickle.load(open('liver_analysis.pkl', 'rb'))
```

```
prediction= model.predict(data)[0]
```

```
if (prediction == 1):
```

```
    return render_template('noChance.html', prediction='you have a liver disease  
    problem,you must and:
```

```
    else:
```

```
    return render_template('Chance.html', prediction='you dont have a liver disease  
    problem')
```

```
if __name__ == '__main__':
```

```
    app.run()
```