PROJECT REPORT

1. INTRODUCTION

Overview

A Review Of Liver Patient Analysis Methods Using Machine Learning

Project Description:

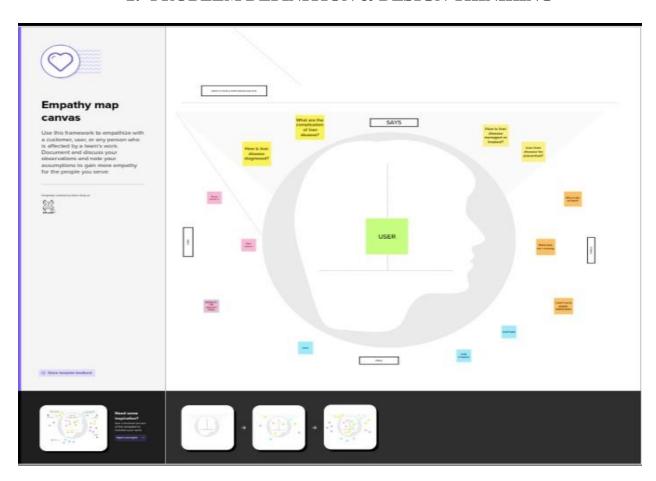
Liver diseases averts the normal function of the liver. This disease is caused by an assortment of elements that harm the liver. Diagnosis of liver infection at the preliminary stage is important for better treatment. In today's scenario devices like sensors are used for detection of infections. Accurate 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

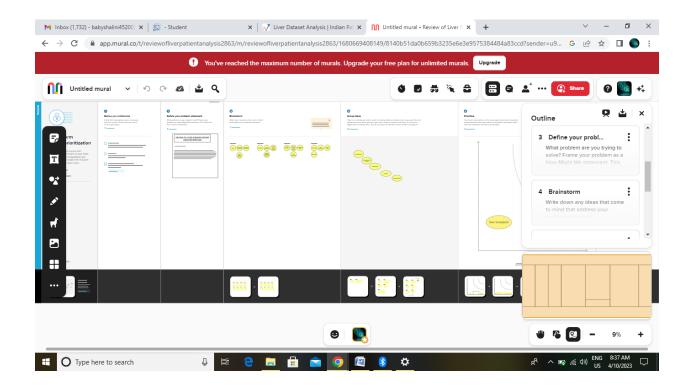
- Liver cirrhosis is the biggest health problem posed by alcohol use, with 1.4 lakh deaths every year.
- > Sadly, no. In fact, it is getting more common in younger people than ever before. Dr. Amrish said that liver disease can set in childhood too as it can pass through genes.
- ➤ Cirrhosis isn't curable, 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.

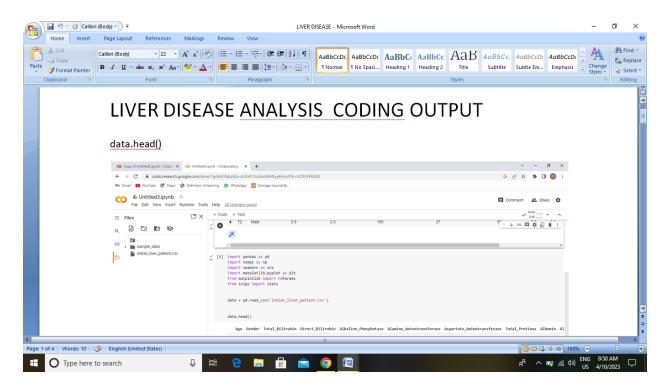
2. PROBLEM DEFINITION & DESIGN THINIKING

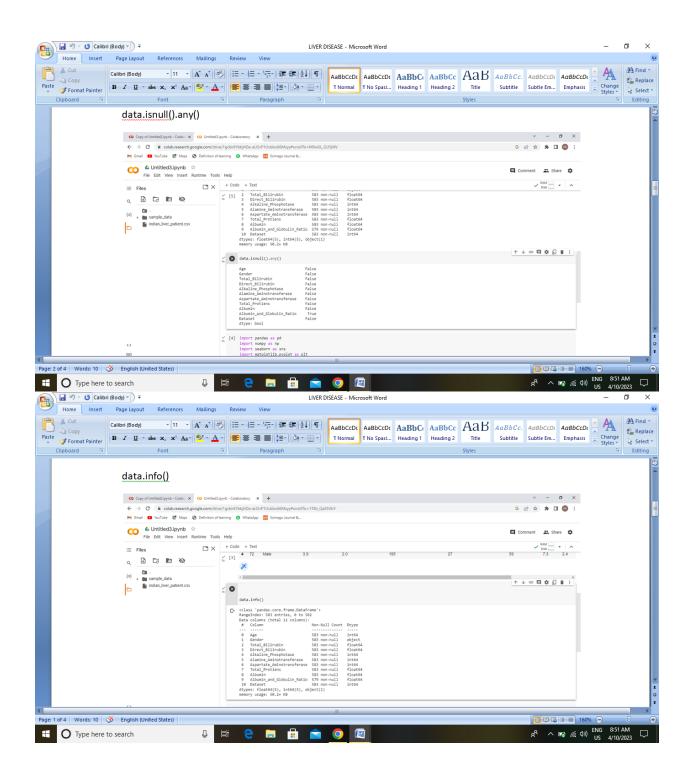


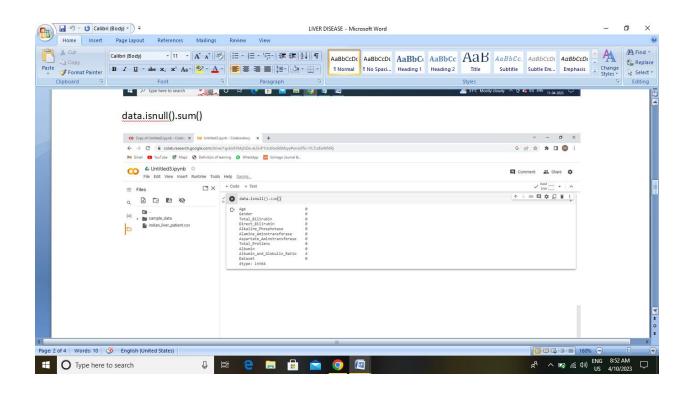
IDEATION & BRAINSTORMING MAP

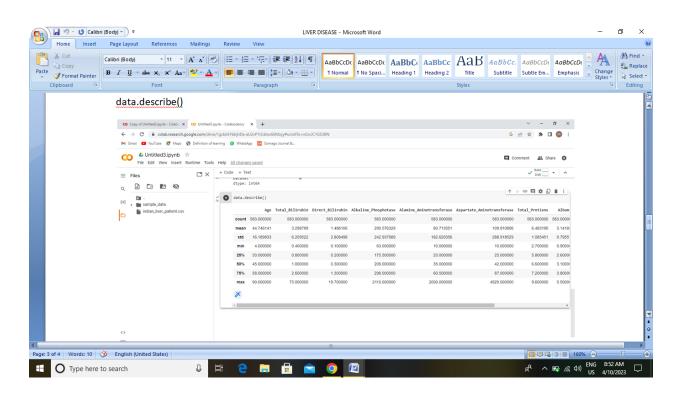


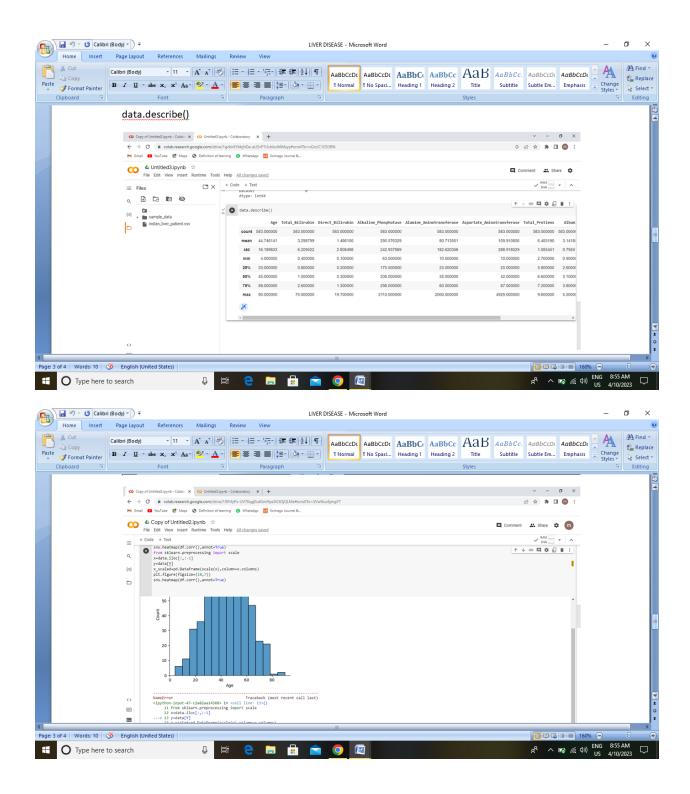
3. RESULT

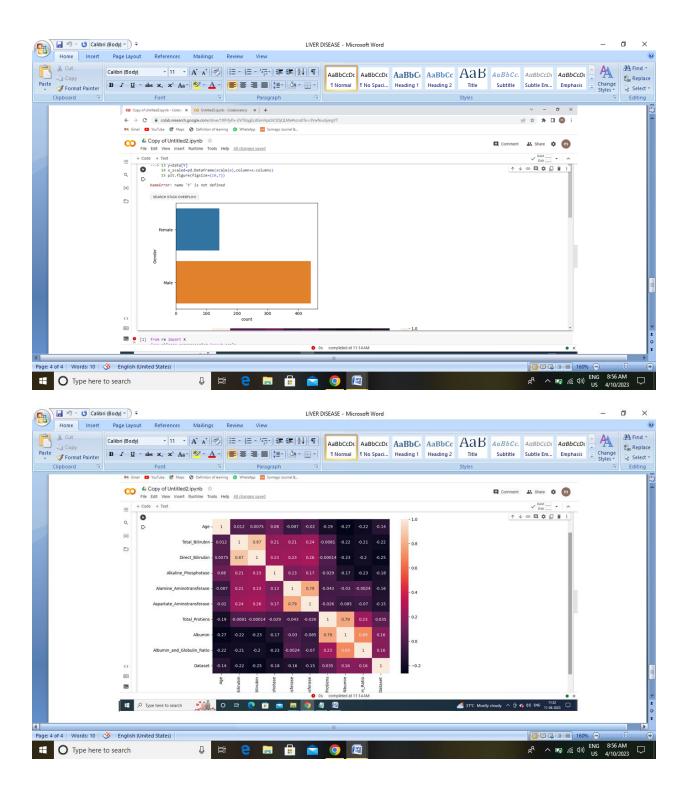












4. ADVANTAGES & DISADVANTAGES

Liver biopsy

Benefits	Disadvantages
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_Radio'].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.proprocessing 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_samplig 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)
dtc1=accuracy_score(y_test,y_predict)
dtc1
pd.crosstab(y_test,y_predict)
print(classification_report(y_test, y_predict))
```

```
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_crossentopy', 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
```

import tensorflow.keras

from tensorflow.keras.models import sequential

```
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', logi1]]
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,ascendi
ng=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), flot(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 desease
problem,you must and:
else:
return render_template('Chance.html', prediction='you dont have a liver desease
problem')

if__name__ == '__main__':
app.run()
```