

Project Name:	Project - Early Detection of Chronic Kidney Disease using Machine Learning
Team ID:	PNT2022TMID13778

FINAL CODE PDF

Collecting , Visualizing, and Preprocessing the Dataset

1.Importing the packages

```
import numpy as np

import pandas as pd

import matplotlib.pyplot as plt

from collections import Counter as c

import seaborn as sns

import missingno as msng

from sklearn.metrics import accuracy_score, confusion_matrix

from sklearn.model_selection import train_test_split

from sklearn.preprocessing import LabelEncoder

from sklearn.linear_model import LogisticRegression
```

#Data Collections

```
data=pd.read_csv("/content/drive/MyDrive/chronickidneydisease.csv")

data.head()
```

	id	age	bp	sg	al	su	rbc	pc	pcc	ba	...	pcv	wc	rc	htn	dm	cad	appet	pe	ane	classification
0	0	48.0	80.0	1.020	1.0	0.0	NaN	normal	notpresent	notpresent	...	44	7800	5.2	yes	yes	no	good	no	no	ckd
1	1	7.0	50.0	1.020	4.0	0.0	NaN	normal	notpresent	notpresent	...	38	6000	NaN	no	no	no	good	no	no	ckd
2	2	62.0	80.0	1.010	2.0	3.0	normal	normal	notpresent	notpresent	...	31	7500	NaN	no	yes	no	poor	no	yes	ckd
3	3	48.0	70.0	1.005	4.0	0.0	normal	abnormal	present	notpresent	...	32	6700	3.9	yes	no	no	poor	yes	yes	ckd
4	4	51.0	80.0	1.010	2.0	0.0	normal	normal	notpresent	notpresent	...	35	7300	4.6	no	no	no	good	no	no	ckd

5 rows × 26 columns

```
data.drop(['id'],axis=1,inplace=True)
```

```
data.columns
```

```
data.columns=['age', 'bp', 'sg', 'al', 'su', 'rbc', 'pc', 'pcc', 'ba', 'bgr', 'bu',
              'sc', 'sod', 'pot', 'hemo', 'pcv', 'wc', 'rc', 'htn', 'dm', 'cad',
              'appet', 'pe', 'ane', 'classification']
```

```
data.columns
```

```
data['classification'].unique()
```

```
data.info()
```

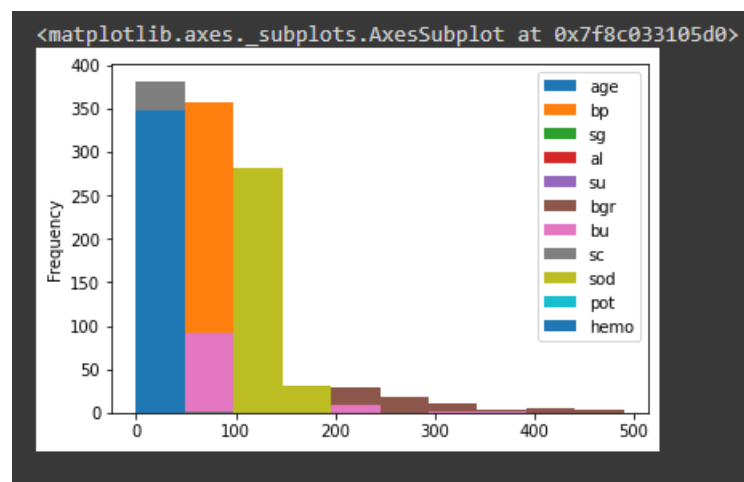
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 25 columns):
#   Column                Non-Null Count  Dtype
---  -
0   age                   391 non-null   float64
1   bp                    388 non-null   float64
2   sg                    353 non-null   float64
3   al                    354 non-null   float64
4   su                    351 non-null   float64
5   rbc                   248 non-null   object
6   pc                    335 non-null   object
7   pcc                   396 non-null   object
8   ba                    396 non-null   object
9   bgr                   356 non-null   float64
10  bu                    381 non-null   float64
11  sc                    383 non-null   float64
12  sod                   313 non-null   float64
13  pot                   312 non-null   float64
14  hemo                  348 non-null   float64
15  pcv                   330 non-null   object
16  wc                    295 non-null   object
17  rc                    270 non-null   object
18  htn                   398 non-null   object
19  dm                    398 non-null   object
20  cad                   398 non-null   object
21  appet                399 non-null   object
22  pe                    399 non-null   object
23  ane                   399 non-null   object
24  classification        400 non-null   object
dtypes: float64(11), object(14)
memory usage: 78.2+ KB
```

2. Data visualization

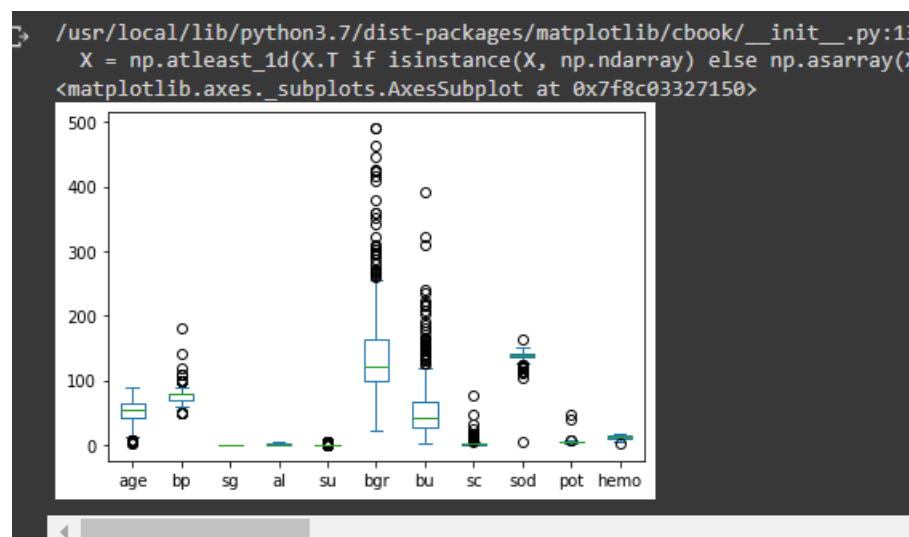
```
from matplotlib import pyplot
```

```
data.plot
```

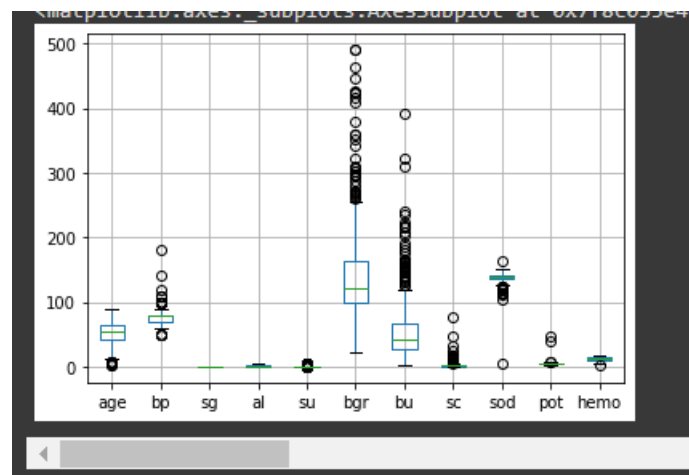
```
data.plot.hist()
```



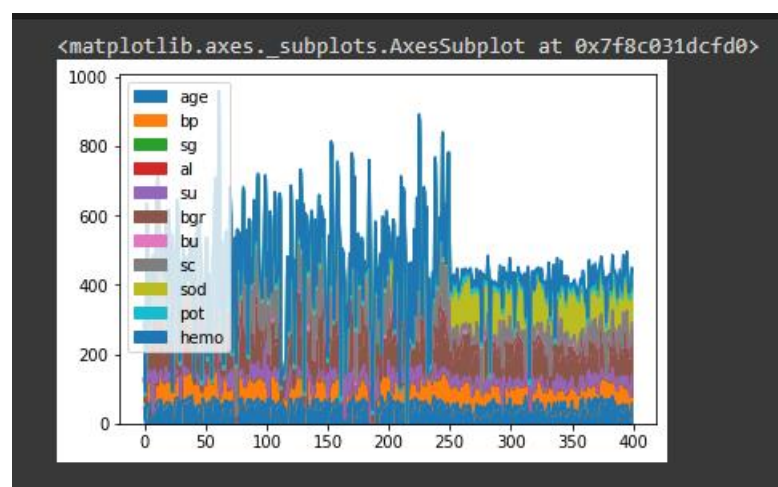
data.plot.box()



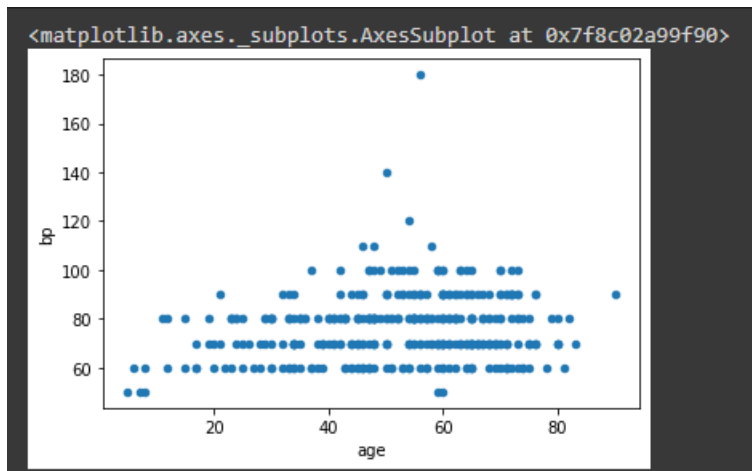
data.boxplot()



data.plot.area()



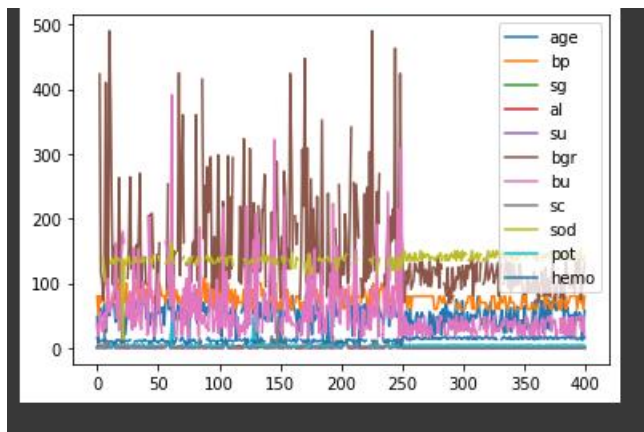
data.plot.scatter(x='age',y='bp')



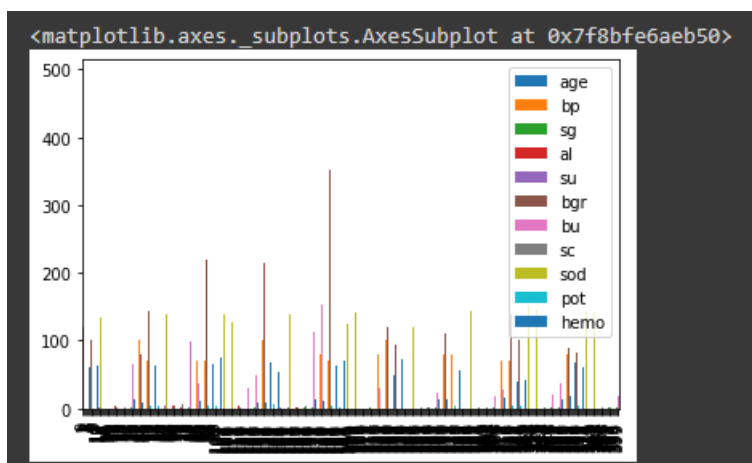
pie = data

pie

pie.plot();



data.plot.bar()

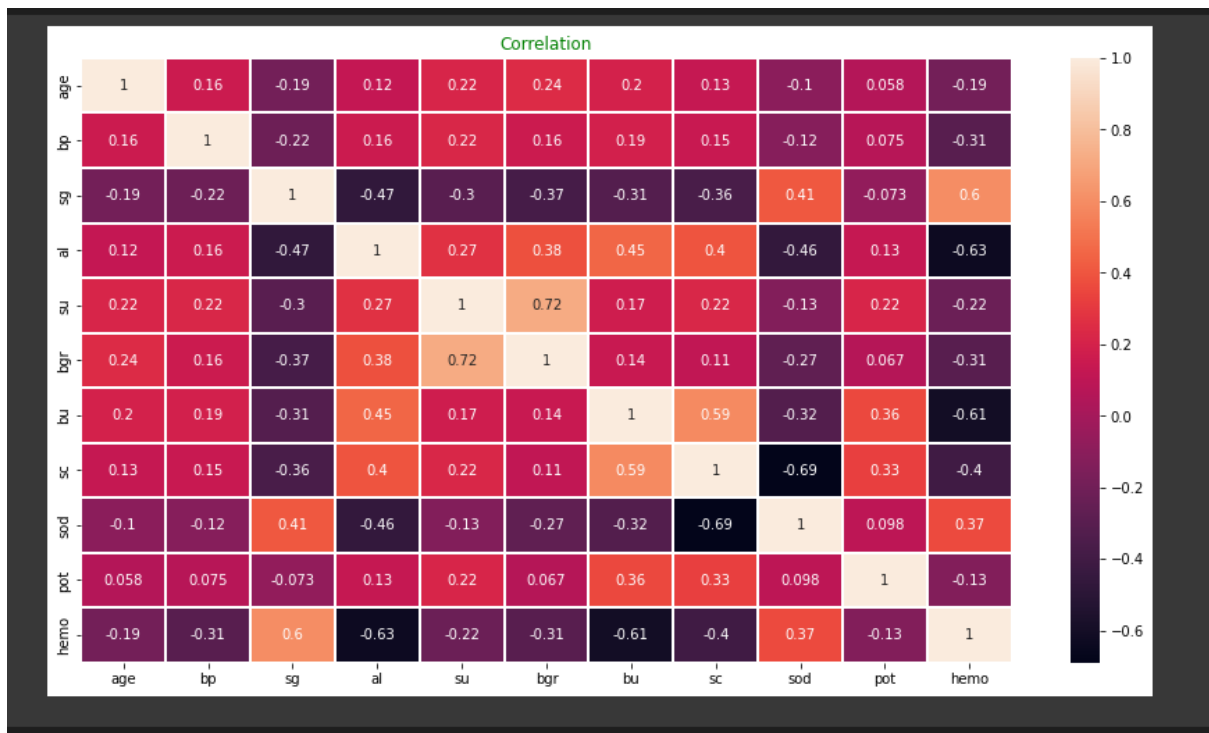


data.corr()

plt.figure(figsize=(15,8));

plt.title("Correlation",color="green")

```
sns.heatmap(data.corr(),linewidth=1,annot=True);
```



```
sns.set_theme(style="white")
```

```
fig, ((ax1, ax2,ax3,ax4,ax5), (ax6, ax7,ax8,ax9,ax10))= plt.subplots(nrows=2, ncols=5, figsize=(18,14))
```

```
sns.boxplot(data=data,x="age",ax=ax1)
```

```
sns.boxplot(data=data,x="bp",ax=ax2)
```

```
sns.boxplot(data=data,x="sg",ax=ax3)
```

```
sns.boxplot(data=data,x="al",ax=ax4)
```

```
sns.boxplot(data=data,x="bgr",ax=ax5)
```

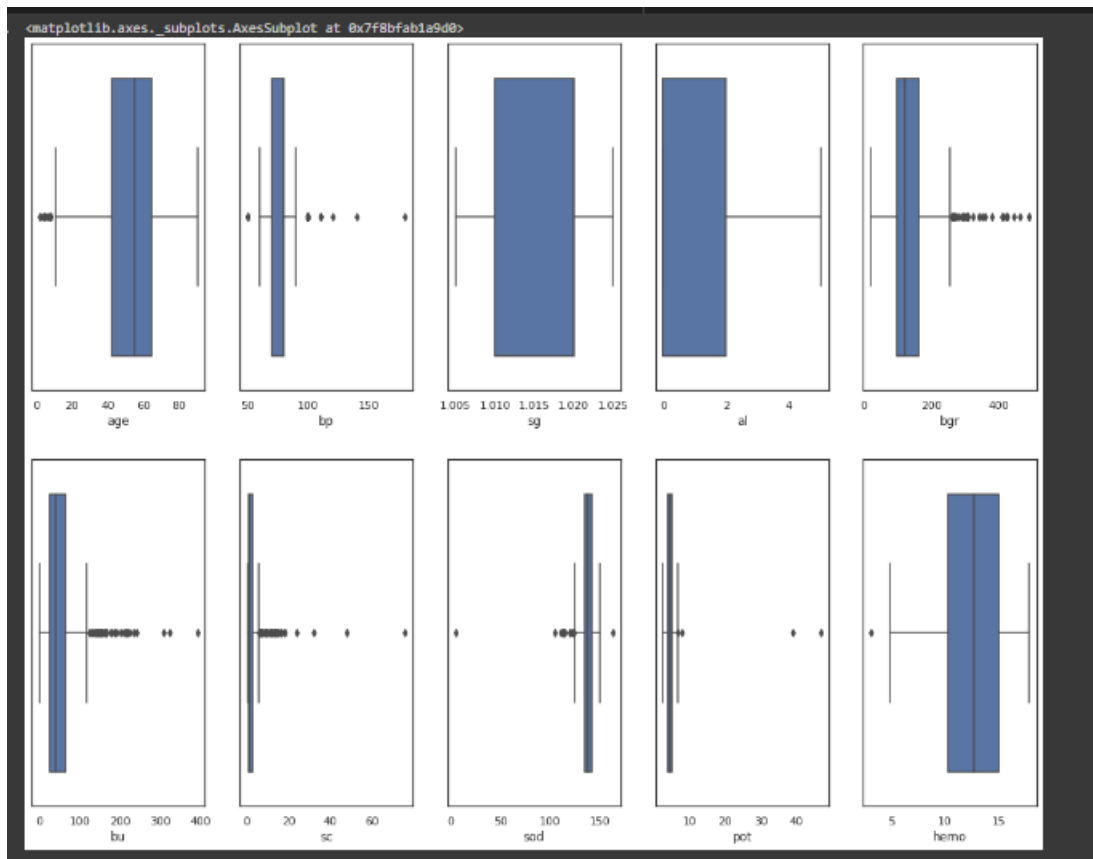
```
sns.boxplot(data=data,x="bu",ax=ax6)
```

```
sns.boxplot(data=data,x="sc",ax=ax7)
```

```
sns.boxplot(data=data,x="sod",ax=ax8)
```

```
sns.boxplot(data=data,x="pot",ax=ax9)
```

```
sns.boxplot(data=data,x="hemo",ax=ax10)
```



3. Data Preprocessing

```
data['classification']=data['classification'].replace("ckd\t",'ckd')
```

```
catcols=set(data.dtypes[data.dtypes=='O'].index.values)
```

```
print(catcols)
```

```
for i in catcols:
```

```
    print("columns:",i)
```

```
    print(c(data[i]))
```

```
    print('*'*120+'\n')
```

```
catcols.remove('rbc')
```

```
catcols.remove('pcv')
```

```
catcols.remove('wc')
```

```
catcols
```

```
→ {'ane',
    'appet',
    'ba',
    'cad',
    'classification',
    'dm',
    'htn',
    'pc',
    'pcc',
    'pe',
    'rc'}
```

```
contcols=set(data.dtypes[data.dtypes!='O'].index.values)
```

```
contcols
```

```
for i in catcols:
```

```
    print("continuous columns :",i)
```

```
    print(c(data[i]))
```

```
    print('*'*120+'\n')
```

```
contcols.remove('sg')
```

```
contcols.remove('al')
```

```
contcols.remove('su')
```

```
print(contcols)
```

```
contcols.add('rbc')
```

```
contcols.add('pc')
```

```
contcols.add('wc')
```

```
print(contcols)
```

```
catcols.add('sg')
```

```
catcols.add('al')
```

```
catcols.add('su')
```

```
print(catcols)
```

```
data['cad']=data.cad.replace("\tno",'no')
```

```
c(data['cad'])
```

```
data['dm']=data.dm.replace(to_replace={"\tno":'no','\tyes':'yes',' yes':'yes'})
```

```
c(data['dm'])
```

```
data.isna().any()
```

```
age      True
bp       True
sg       True
al       True
su       True
rbc      True
pc       True
pcc      True
ba       True
bgr      True
bu       True
sc       True
sod      True
pot      True
hemo     True
pcv      True
wc       True
rc       True
htn      True
dm       True
cad      True
appet    True
pe       True
ane      True
classification  False
dtype: bool
```

```
data.isna().sum()
```



```

age          9
bp           12
sg           47
al           46
su           49
rbc          152
pc           65
pcc          4
ba           4
bgr          44
bu           19
sc           17
sod          87
pot          88
hemo         52
pcv          70
wc           105
rc           130
htn          2
dm           2
cad          2
appet        1
pe           1
ane          1
classification 0
dtype: int64

```

```

data.pcv=pd.to_numeric(data.pcv,errors='coerce')
data.wc=pd.to_numeric(data.wc,errors='coerce')
data.rc=pd.to_numeric(data.rc,errors='coerce')
data['bgr'].fillna(data['bgr'].mean(),inplace=True)
data['bp'].fillna(data['bp'].mean(),inplace=True)
data['bu'].fillna(data['bu'].mean(),inplace=True)
data['hemo'].fillna(data['hemo'].mean(),inplace=True)
data['pcv'].fillna(data['pcv'].mean(),inplace=True)
data['pot'].fillna(data['pot'].mean(),inplace=True)
data['rc'].fillna(data['rc'].mean(),inplace=True)
data['sc'].fillna(data['sc'].mean(),inplace=True)
data['sod'].fillna(data['sod'].mean(),inplace=True)
data['wc'].fillna(data['wc'].mean(),inplace=True)
data['age'].fillna(data['age'].mode()[0],inplace=True)
data['htn'].fillna(data['htn'].mode()[0],inplace=True)

```

```

data['pcc'].fillna(data['pcc'].mode()[0],inplace=True)
data['appet'].fillna(data['appet'].mode()[0],inplace=True)
data['al'].fillna(data['al'].mode()[0],inplace=True)
data['pc'].fillna(data['pc'].mode()[0],inplace=True)
data['rbc'].fillna(data['rbc'].mode()[0],inplace=True)
data['cad'].fillna(data['cad'].mode()[0],inplace=True)
data['ba'].fillna(data['ba'].mode()[0],inplace=True)
data['ane'].fillna(data['ane'].mode()[0],inplace=True)
data['su'].fillna(data['su'].mode()[0],inplace=True)
data['dm'].fillna(data['dm'].mode()[0],inplace=True)
data['pe'].fillna(data['pe'].mode()[0],inplace=True)
data['sg'].fillna(data['sg'].mode()[0],inplace=True)

```

ML MODEL CREATION

Importing the packages

```

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import missingno as msng
from sklearn.metrics import accuracy_score,confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from sklearn.linear_model import LogisticRegression

```

1.splitting the dataset

```

for i in catcols:
    print("LABEL ENCODING OF :",i)
    le=LabelEncoder()
    print(c(data[i]))
    data[i]=le.fit_transform(data[i])
    print(c(data[i]))

```

```
print('*'*100)
```

```
data['rbc']=le.fit_transform(data['rbc'])
```

```
selcols=['rbc','pc','bgr','bu','pe','ane','dm','cad']
```

```
x=pd.DataFrame(data,columns=selcols)
```

```
y=pd.DataFrame(data,columns=['classification'])
```

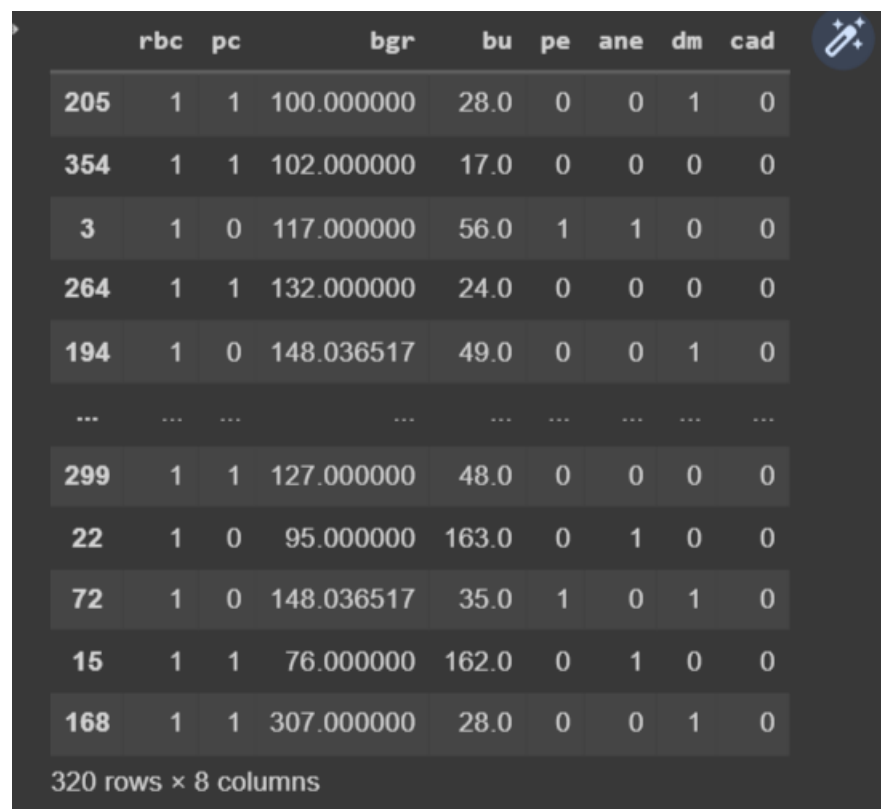
```
print(x.shape)
```

```
print(y.shape)
```

```
(400, 8)
(400, 1)
```

```
xtrain,xtest,ytrain,ytest=train_test_split(x,y,test_size=0.2,random_state=2)
```

```
xtrain
```



	rbc	pc	bgr	bu	pe	ane	dm	cad
205	1	1	100.000000	28.0	0	0	1	0
354	1	1	102.000000	17.0	0	0	0	0
3	1	0	117.000000	56.0	1	1	0	0
264	1	1	132.000000	24.0	0	0	0	0
194	1	0	148.036517	49.0	0	0	1	0
...
299	1	1	127.000000	48.0	0	0	0	0
22	1	0	95.000000	163.0	0	1	0	0
72	1	0	148.036517	35.0	1	0	1	0
15	1	1	76.000000	162.0	0	1	0	0
168	1	1	307.000000	28.0	0	0	1	0

320 rows × 8 columns

2. Model creation

```
lgr=LogisticRegression()
```

```
lgr.fit(xtrain.values,ytrain.values)
```

```
ypred=lgr.predict(xtest)
ypred1=lgr.predict([[129,99,1,0,0,1,0,1]])
print(ypred1)
c(ypred)
```

3. Accuracy , Confusion Matrix , Classification Report

```
[1]
Counter({0: 48, 1: 32})
```

```
print(accuracy_score(ytest,ypred)*100)
```

```
92.5
```

```
confmat=confusion_matrix(ytest,ypred)
```

```
confmat
```

```
array([[48,  6],
       [ 0, 26]])
```

```
from sklearn.metrics import classification_report
```

```
print(classification_report(ytest, ypred))
```

	precision	recall	f1-score	support
0	1.00	0.89	0.94	54
1	0.81	1.00	0.90	26
accuracy			0.93	80
macro avg	0.91	0.94	0.92	80
weighted avg	0.94	0.93	0.93	80

```
from sklearn.model_selection import cross_val_score
```

```
scores = cross_val_score(lgr, xtrain, ytrain, cv=50)
```

```
print('Cross-Validation Accuracy Scores', scores)
```

```
Cross-Validation Accuracy Scores [0.85714286 0.85714286 0.85714286 0.71428571 1. 1.
0.85714286 1. 0.85714286 0.71428571 1. 0.85714286
0.85714286 0.85714286 0.85714286 1. 1. 1.
0.85714286 1. 1. 1. 1. 1.
0.83333333 0.83333333 1. 1. 1. 1.
0.83333333 0.83333333 0.66666667 0.83333333 1. 0.83333333
1. 1. 0.83333333 0.83333333 0.83333333 1.
1. 0.83333333 1. 0.83333333 0.66666667 0.83333333
1. 1. ]
```

1.FrontEnd Development

Frontend consists of 3 pages

1. Index page
2. Prediction page
3. Output page

Technology used in Frontend

HTML

CSS

JS

1.Index.html

```
<!DOCTYPE html>
```

```
<html lang="en">
```

```
<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>Document</title>
```

```
<style>
```

```
* {
```

```
padding: 0;
```

```
margin: 0;
```

```
}
```

```
.background {
```

```
background-image: url("/images/bg4.jpeg");
```

```
background-repeat: no-repeat;
background-size: cover;
}
.header {
display: flex;
flex: 100%;
flex-direction: row;
justify-content: flex-end;
height: 60px;
padding-right: 30px;
background-color: #2e6e82;
align-items: center;
}
.btn{
color: #fff;
font-size: large;
text-decoration: none;
}
.titleWrapper{
height: 500px;
display: flex;
justify-content: center;
align-items: center;
}
.title{
background-color: #2e6e82;
border-radius: 5px;
padding: 20px 90px;
}
</style>
</head>
```



```
<style>

.header {
  display: flex;
  justify-content: center;
  align-items: center;
}

.title {
  background-color: #2e6e82;
  border-radius: 5px;
  padding: 20px 90px;
  color: white;
}

.inputs {
  display: flex;
  flex-direction: column;
  justify-content: center;
  align-items: center;
  margin-top: 10px;
  background-color: #2e6e82;
  padding: 50px;
  border-radius: 50px;
}

input {
  width: 300px;
  height: 25px;
  text-align: center;
  margin-bottom: 5px;
  font-size: large;
}

select {
  width: 310px;
```



```
height: 25px;
text-align: center;
margin-bottom: 5px;
font-size: large;
}
.btn {
display: flex;
justify-content: center;
align-items: center;
margin-top: 30px;
}
button {
position: relative;
font-size: 14px;
letter-spacing: 3px;
height: 3em;
padding: 0 3em;
border: none;
background-color: #2e6e82;
color: #fff;
text-transform: uppercase;
overflow: hidden;
border-radius: 5px;
}
```

```
button::before {
content: "";
display: block;
position: absolute;
z-index: 0;
bottom: 0;
```

```
left: 0;
height: 0px;
width: 100%;
background: rgb(46, 110, 130);
background: linear-gradient(
  90deg,
  rgba(46, 110, 130, 1) 20%,
  rgba(46, 110, 130, 1) 100%
);
transition: 0.2s;
}
```

```
button .label {
  position: relative;
}
```

```
button .icon {
  display: flex;
  align-items: center;
  justify-content: center;
  height: 3em;
  width: 3em;
  position: absolute;
  top: 3em;
  right: 0;
  opacity: 0;
  transition: 0.4s;
}
```

```
button:hover::before {
  height: 100%;
```

```
}
```

```
button:hover .icon {
```

```
  top: 0;
```

```
  opacity: 1;
```

```
}
```

```
</style>
```

```
</head>
```

```
<body>
```

```
<div class="header">
```

```
  <h1 class="title">Chronic Kidney disease prediction</h1>
```

```
</div>
```

```
<div class="inputs">
```

```
  <input type="number" placeholder="Blood Urea" />
```

```
  <input type="number" placeholder="Blood Glucose Random" />
```

```
  <select name="Coronary Artery Disease" id="">
```

```
    <option value="Coronary Artery Disease">Coronary Artery Disease</option>
```

```
    <option value="yes">Yes</option>
```

```
    <option value="no">No</option>
```

```
</select>
```

```
<select name="anemia" id="">
```

```
  <option value="anemia">Anemia</option>
```

```
  <option value="yes">Yes</option>
```

```
  <option value="no">No</option>
```

```
</select>
```

```
<select name="pus cell" id="">
```

```
  <option value="pus cell">Pus Cell</option>
```

```
  <option value="normal">Normal</option>
```

```
  <option value="abnormal">Abnormal</option>
```

```
</select>
```

```
<input type="number" placeholder="Red Blood Cell Count" />
```

```

<select name="diabetes mellitus" id="">
  <option value="diabetes mellitus">Diabetes Mellitus</option>
  <option value="yes">Yes</option>
  <option value="no">No</option>
</select>

<select name="pedal edema" id="">
  <option value="pedal edema">Pedal Edema</option>
  <option value="yes">Yes</option>
  <option value="no">No</option>
</select>

</div>

<div class="btn">
  <a href="results.html">
    <button>
      <span class="label">Predict</span>
      <span class="icon">
        <svg
          xmlns="http://www.w3.org/2000/svg"
          viewBox="0 0 24 24"
          width="24"
          height="24"
          >
          <path fill="none" d="M0 0h24v24H0z"></path>
          <path
            fill="currentColor"
            d="M16.172 11l-5.364-5.364 1.414-1.414L20 12l-7.778 7.778-1.414-1.414L16.172
13H4v-2z"
          ></path>
        </svg>
      </span>
    </button>
  </a>
</div>

```

```

</a>
</div>
</body>
</html>

```

Inputs page Or Prediction page

Chronic Kidney disease predication

Blood Urea

Blood Glucose Random

Coronary Artery Disease

Anemia

Pus Cell

Red Blood Cell Count

Diabetes Mellitus

Pedal Edema

PREDICT

FrontEnd and Backend connection

1.FrontEnd Development

OUTPUT PAGE

Result.html

```

<!DOCTYPE html>
<html lang="en">
<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>Document</title>

```

```
<style>

.header {
  display: flex;
  justify-content: center;
  align-items: center;
}

.title {
  background-color: #2e6e82;
  border-radius: 5px;
  padding: 20px 90px;
  color: white;
}

.resultWrapper {
  display: flex;
  height: 200px;
  justify-content: center;
  align-items: center;
}

.result {
  border-radius: 10px;
  padding: 10px 30px;
}

.result-positive {
  color: red;
  font-size: larger;
}

.result-negative {
  color: blue;
  font-size: larger;
}

h2 {
```

```
        color: #2e6e82;
    }
</style>
</head>
<body>
    <div class="header">
        <h1 class="title">Chronic Kidney disease predication</h1>
    </div>
    <div class="resultWrapper">
        <div class="result">
            <h2>
                Prediction:
                <samp class="result-positive">You have Chronic Kidney Disease</samp>
            </h2>
            <!-- <h2>
                Prediction:
                <samp class="result-negative"> You Don't Chronic Kidney Disease</samp>
            </h2> -->
        </div>
    </div>
</body>
</html>
```

Results Page

Chronic Kidney disease prediction

Prediction: You have Chronic Kidney Disease

2.Backend development

Flask

```
import pandas as pd
from flask import Flask, request, render_template
import pickle

app = Flask(__name__) # initializing a flask app
model = pickle.load(open('CKD.pkl', 'rb')) #loading the model

@app.route('/')# route to display the home page
def home():
    return render_template('home.html') #rendering the home page
@app.route('/Prediction',methods=['POST','GET'])
def prediction(): # route to display prediction page
    return render_template('indexnew.html')
@app.route('/Home',methods=['POST','GET'])
def my_home():
    return render_template('home.html')

@app.route('/predict',methods=['POST'])# route to show the predictions in a web UI
def predict():
    #reading the inputs given by the user
    input_features = [float(x) for x in request.form.values()]
    features_value = [np.array(input_features)]
```



```
features_name = ['blood_urea', 'blood glucose random', 'coronary_artery_disease',  
                'anemia', 'pus_cell', 'red_blood_cells', 'diabetesmellitus', 'pedal_edema']  
  
df = pd.DataFrame(features_value, columns=features_name)  
  
output = model.predict(df) # predictions using the loaded model file  
  
# showing the prediction results in a UI# showing the prediction results in a UI  
return render_template('result.html', prediction_text=output)  
  
if __name__ == '__main__':  
    app.run(debug=True) # running the app
```