

```

#from tabulate import tabulate

import pandas as pd

a=pd.read_excel(r"C:\Users\vennela\Documents\miniproject2/traindata1.xlsx",header=
0)

a["Gender of the patient"]=a["Gender of the patient"].replace("Female",0)

a["Gender of the patient"]=a["Gender of the patient"].replace("Male",1)

print(a.isnull().sum())

a=a.dropna()

#col=a.columns

#print(col)

#a["Albumin_and_Globulin_Ratio"]=a["Albumin_and_Globulin_Ratio"].fillna(0)

print(a)

a.hist(figsize=(14,11))

print(a.isnull().sum())


import numpy as np

#from tabulate import tabulate

import matplotlib.pyplot as plt

from sklearn import preprocessing

x=a.iloc[:,2:-1]

print(x)

y=np.array(a.iloc[:,10:])

y=np.ravel(y)

print(y)

```

```

from sklearn.linear_model import LogisticRegression

from sklearn.metrics import confusion_matrix, accuracy_score, classification_report

classifier = LogisticRegression(solver='lbfgs', random_state=0, max_iter=10000)

#print(classifier)

classifier.fit(X_train, Y_train)

predicted_y = classifier.predict(X_test)

print(predicted_y)

y_test_h=np.array([0.7,0.1,187,16,18,6.8,3.3,0.9])

y_test_h=y_test_h.reshape(1,8)

print(y_test_h)

h=classifier.predict(y_test_h)

print(h)

cm = confusion_matrix(Y_test, predicted_y)

print(cm)

print(accuracy_score(predicted_y,Y_test))

print(classification_report(predicted_y,Y_test))

```

output

```
[1 1 1 ... 1 1 1]
```

```
[[7.00e-01 1.00e-01 1.87e+02 1.60e+01 1.80e+01 6.80e+00 3.30e+00 9.00e-01]]
```

```
[1]
```

```
[[4595 231]
```

```
[1667 297]]
```

```
0.7204712812960236
```

	precision	recall	f1-score	support
--	-----------	--------	----------	---------

1	0.95	0.73	0.83	6262
---	------	------	------	------

2	0.15	0.56	0.24	528
---	------	------	------	-----

accuracy		0.72		6790
----------	--	------	--	------

macro avg	0.55	0.65	0.53	6790
-----------	------	------	------	------

weighted avg	0.89	0.72	0.78	6790
--------------	------	------	------	------

Decision Tree

from sklearn import tree

from sklearn.tree import DecisionTreeClassifier

clf_d = DecisionTreeClassifier()

clf_d.fit(X_train, Y_train)

y_predict=clf_d.predict(X_test)

print(confusion_matrix(Y_test,y_predict))

print(accuracy_score(y_predict,Y_test))

print(classification_report(y_predict,Y_test))

[[4824 2]

[1 1963]]

0.9995581737849779

	precision	recall	f1-score	support
--	-----------	--------	----------	---------

1	1.00	1.00	1.00	4825
---	------	------	------	------

2	1.00	1.00	1.00	1965
---	------	------	------	------

accuracy			1.00	6790
----------	--	--	------	------

macro avg	1.00	1.00	1.00	6790
-----------	------	------	------	------

weighted avg	1.00	1.00	1.00	6790
--------------	------	------	------	------

```
from sklearn.ensemble import RandomForestClassifier
```

```
from sklearn.metrics import confusion_matrix,accuracy_score,classification_report
```

```
clf_r=RandomForestClassifier(n_estimators=500)
```

```
clf_r.fit(X_train,Y_train)
```

```
y_pred_r=clf_r.predict(X_test)
```

```
y_test_h=np.array([0.7,0.1,187,16,18,6.8,3.3,0.9])
```

```
y_test_h=y_test_h.reshape(1,8)
```

```
h=clf_r.predict(y_test_h)
```

```
print(h)
```

```
print(accuracy_score(y_pred_r,Y_test))
```

```
print(classification_report(y_pred_r,Y_test))
```

```
[1]
```

```
0.9998527245949926
```

	precision	recall	f1-score	support
--	-----------	--------	----------	---------

1	1.00	1.00	1.00	4827
---	------	------	------	------

2	1.00	1.00	1.00	1963
---	------	------	------	------

accuracy			1.00	6790
----------	--	--	------	------

macro avg	1.00	1.00	1.00	6790
-----------	------	------	------	------

weighted avg	1.00	1.00	1.00	6790
--------------	------	------	------	------

```
from sklearn.svm import SVC
```

```
from sklearn.metrics import confusion_matrix,accuracy_score,classification_report
```

```
clf_s = SVC(random_state=7,gamma="auto")
```

```
clf_s.fit(X_train,Y_train)
```

```
y_pred_s=clf_s.predict(X_test)
```

```
print(accuracy_score(y_pred_s,Y_test))
```

```
print(classification_report(y_pred_s,Y_test))
```

0.9998527245949926

	precision	recall	f1-score	support
--	-----------	--------	----------	---------

1	1.00	1.00	1.00	4827
---	------	------	------	------

2	1.00	1.00	1.00	1963
---	------	------	------	------

accuracy			1.00	6790
----------	--	--	------	------

macro avg	1.00	1.00	1.00	6790
-----------	------	------	------	------

weighted avg	1.00	1.00	1.00	6790
--------------	------	------	------	------